# The Multi-Run Simulation Environment SimEnv

**User's Guide for Version V1.11** 

by M. Flechsig, U. Böhm, T. Nocke & C. Rachimow



#### **Disclaimer of Warranty**

The authors make no warranties, expressed or implied, that the programs and data contained in the software package and the formulas given in this document are free of error, or are consistent with any particular standard of merchantability, or that they will meet your requirements for any particular application. They should not be relied for solving a problem whose incorrect solution could result in injury to a person or loss of property. If you do use the programs or data or formulas in such a manner, it is on your own risk. The authors disclaim all liability for direct or consequential damages from your use of the programs and data.



# The Multi-Run Simulation Environment SimEnv

# User's Guide for Version V1.11 (16-Jun-2004)

by

Michael Flechsig Potsdam Institute for Climate Impact Research

Department Data & Computation, flechsig@pik-potsdam.de

Uwe Böhm University of Potsdam

Institute for Physics, boehm@pik-potsdam.de

Thomas Nocke University of Rostock

Institute of Computer Graphics, nocke@informatik.uni-rostock.de

Claus Rachimow Potsdam Institute for Climate Impact Research

Department Data & Computation, rachimow@pik-potsdam.de

# SimEnv in the Internet: http://www.pik-potsdam.de/topik/pikuliar/simenv/home



Potsdam Institute for Climate Impact Research

Telegrafenberg

14473 Potsdam, Germany

Phone ++49 - 331 - 288 2604 Fax ++49 - 331 - 288 2600 WWW http://www.pik-potsdam.de



University of Potsdam Institute for Physics Am Neuen Palais 10 14469 Potsdam, Germany

WWW http://www.physics.uni-potsdam.de



University of Rostock Institute of Computer Graphics Albert-Einstein-Str. 21 18059 Rostock, Germany

WWW http://wwwicg.informatik.uni-rostock.de



# Contents

	EXECUTIVE SUMMARY	1
1	ABOUT THIS DOCUMENT	3
1.1	Document Conventions	
1.2	Example Layout	3
2	GETTING STARTED	5
2	VERSION 1.11	
3 3.1	What is New?	
3.2	Known Bugs and Their Workarounds	
3.3	Limitations	
4	EXPERIMENT TYPES	ç
4.1	General Approach	
4.2	Behavioural Analysis	
4.3	Monte Carlo Analysis	
4.4 4.5	Local Sensitivity Analysis	
-	·	
5	MODEL INTERFACE	
5.1 5.2	Coordinate and Grid Assignments to Variables	
5.3	Model Interface for Fortran and C/C++ Models	
5.4	Model Interface for Python Models	
5.4.1	Standard User-Defined Files for Python Models	27
5.5	Model Interface at Shell Script Level	
5.6	Model Interface for GAMS Models	29
5.6.1 5.6.2	GAMS Description File <model>.gdf</model>	
5.6.3	Files Created during GAMS Model Performance	
5.7	Distributed Models	34
5.8	Running Interfaced Models Outside SimEnv	34
6	EXPERIMENT PREPARATION	35
6.1	Experiment Description File <model>.edf</model>	
6.2	Behavioural Analysis	
6.2.1 6.2.2	AdjustmentsThe Combination	
6.2.3	Example	
6.2.4	Experiment Performance	38
6.3	Monte Carlo Analysis	
6.3.1 6.3.2	Adjustments  Distribution Functions and their Parameters	40
6.3.3	Example	
6.3.4	Experiment Performance	
6.4	Local Sensitivity Analysis	42
6.4.1	Adjustments	
6.4.2 6.4.3	Example Experiment Performance	
6.5	Optimization	
6.5.1	Adjustments	
6.5.2	Example	
6.5.3	Experiment Performance	45
7	EXPERIMENT PERFORMANCE	47
7.1	Experiment Start	
7.2	Experiment Restart	
7.3 7.4	Experiment Partial Performance	
7. <del>4</del> 7.5	Experiment-Related User Scripts and Files	
7.6	Saving Experiments	
8	EXPERIMENT POST-PROCESSING	
8.1	Operands	
8.2	Model Output Variables	



8.3	Operators	57
8.3.1	Operands and Coordinate Checking	
8.4	Built-in Elemental, Basic, and Advanced Operators	
8.4.1	Built-in Elemental Operators	
8.4.2	Built-in Basic and Advanced Operators	
8.5	Experiment-Specific Operators	
8.5.1	Multi-Run Standard Aggregation / Moment Operators	69
8.5.2	Behavioural Analysis	70
8.5.3	Monte Carlo Analysis	73
8.5.4	Local Sensitivity Analysis	76
8.5.5	Optimization	
8.6	User-Defined Operators	
8.6.1	Declaration of User-Defined Operator Dynamics	
8.6.2	Operator Definition File <model>.odf</model>	
8.6.3	Handling Undefined Results	
8.7	Undefined Results	83
8.8	Macros and Macro Definition File <model>.mac</model>	
8.9	Handling the Post-Processor	84
9	VISUAL EXPERIMENT EVALUATION	85
10	GENERAL CONTROL, SERVICES, USER FILES, AND SETTINGS	87
10.1	General Configuration File <model>.cfg</model>	
10.1	Main and Auxiliary Services	
10.3	User Scripts and Files	
10.4	Built-in Items, Reserved Names	
10.5	Case Sensitivity	
10.6	Nodata Representation	
10.7	Environment Variables	
11	STRUCTURE OF USER-DEFINED FILES	
11.1	General Structure	
11.2	Value Lists	
11.3	Coordinate Transformation File	102
12	MODEL AND POST-PROCESSOR OUTPUT DATA STRUCTURES	105
12.1	NetCDF Model and Post-Processor Output	
12.1.1	Global Attributes	
12.1.2	Variable Labelling and Variable Attributes	
12.2	IEEE Compliant Binary Model Output	108
12.3	IEEE Compliant Binary and ASCII Post-Processor Output	108
13	PROSPECTS	111
14	REFERENCES	113
15	ADDENDICES	115
15 15.1	APPENDICES	
15.1.1	Version Implementation Linking User Models and User-Defined Operators	
15.1.1	Example Models and User Files	
15.1.2	Example User-Defined Operators	
15.1.3	Technical Limitations	
15.1.4	Examples for Model Interfaces, User-Defined Operators, and Result Import Interfaces	
15.2.1	Examples for Model interfaces, Oser-Defined Operators, and Result import interfaces	120
15.2.1	Fortran Model	
15.2.3	C Model	
15.2.4	C++ Model	
15.2.4	Python Model	
15.2.6	Model Interface at Shell Script Level	
15.2.7	GAMS Model	
15.2.7	Post-Processor User-Defined Operator	
15.2.6	Post-Processor Result Import Interface	
15.2.9	Compilation of Post-Processor Built-in Operators and Operator Arguments	
15.3.1	Post-Processor Built-in Operators (in Thematic Order)	
15.3.1	Post-Processor Built-in Operators (in Thematic Order)	
15.3.2	Character Arguments of Post-Processor Built-in Operators	
15.3.4	Constant Arguments of Post-Processor Built-in Operators	
15.3.4	GlossaryGlossary	
		1 7 1



# **Tables**

Tab. Tab.		Document conventions	
Tab.	3.1	SimEnv changes in version 1.11	7
Tab.	-	User actions to upgrade to version 1.11	7
Tab.	3.3	Known bugs and their workarounds	
Tab.	4 1	Statistical measures	13
Tab.		Probability density functions	
Tab.		Local sensitivity, linearity, and symmetry measures	15
Tab.	5 1	Elements of a model output description file <model>.mdf</model>	20
Tab.		SimEnv data types	
Tab.	_	Model coupler functions at language level	
Tab.		Model coupler functions for Python models	
Tab.	_	Model coupler functions at shell script level	
Tab.		Elements of a GAMS description file <model>.gdf</model>	
Tab.	6.1	Elements of an experiment description file <model>.edf</model>	35
Tab.	6.2	Adjustment types in experiment preparation	
Tab.	6.3	Experiment-specific elements of an edf-file for behavioural analysis	36
Tab.	6.4	Experiment-specific elements of an edf-file for Monte Carlo analysis	
Tab.	6.5	Probability density functions and their parameters	
Tab.	6.6	Probability density functions: Distribution parameters - conditions and adaptation	41
Tab.	6.7	Experiment-specific elements of an edf-file for local sensitivity analysis	
Tab.	6.8	Experiment-specific elements of an edf-file for an optimization experiment	
Tab.	7.1	Experiment-related user scripts and files	51
Tab.	8.1	Additional coordinates	
Tab.	8.2	Built-in elemental operators	
Tab.	8.3	Classified argument restriction(s) / result description	
Tab.	8.4	Built-in advanced operators (without standard aggregation / moments operators)	
Tab.	8.5	Built-in generic standard aggregation / moment operators	65
Tab.	8.6	Built-in standard aggregation / moment operators without suffix	
Tab.	8.7	Built-in standard aggregation / moment operators with suffix _n	
Tab.	8.8	Built-in standard aggregation / moment operators with suffix _I	
Tab.	8.9	Multi-run standard aggregation / moment operators	
Tab.	8.10	Experiment-specific operators for behavioural analysis	
	8.11	Syntax of the filter argument 1 for operator behav	
Tab.	8.12	Experiment-specific operators for Monte Carlo analysis	
Tab.	8.13	Experiment-specific operators for local sensitivity analysis	
Tab.	8.14	Syntax of the filter argument 1 for the local sensitivity operators	
Tab.	8.15	Operator functions: Declarative and computational part	78
	8.16	Operator functions to get and put structural information	
	8.17	Operator function to get / check / put arguments and results	
Tab.	8.18	Elements of an operator description file <model>.odf</model>	
Tab.	8.19	Elements of an macro description file <model>.mac</model>	84
Tab.	10.1	Elements of a general configuration file <model>.cfg</model>	87
Tab.	10.2	<info> value defaults for the general configuration file</info>	88
Tab.	10.3	Service commands	
Tab.	10.4	User scripts and files	91
	10.5	User files generated during SimEnv performance	93
	10.6	Built-in model variables	
Tab.	10.7	Built-in coordinates	
	10.8	Built-in shell script variables in \$SE_HOME/simenv_*_sh	
	10.9	Reserved names and file names in user-defined files and for models	
	10.10	Case sensitivity of SimEnv entities	
	10.11	Data type related nodata values	
Tab.	10.12	Environment variables	



Tab.	11.1	User-defined files	99
Tab.	11.2	Constraints in user-defined files	
Tab.	11.3	Line types in user-defined files	. 100
Tab.	11.4	Syntax rules for value lists	
Tab.	11.5	Elements of an coordinate transformation file	
Tab.	12.1	NetCDF data types	
Tab.	12.2	Additional global NetCDF attributes	. 106
Tab.	12.3	Variable NetCDF attributes	
Tab.	12.4	Variable NetCDF attributes for visualization	. 107
Tab.	15.1	Implemented models for current version	
Tab.	15.2	Implemented model-related user files for current version	. 118
Tab.	15.3	Available user-defined operators	. 118
Tab.	15.4	Current SimEnv technical limitations	. 119
Tab.	15.5	Parameters for the model world	. 120
Tab.	15.6	Post-processor built-in operators (in thematic order)	. 135
Tab.	15.7	Post-processor built-in operators (in alphabetical order)	. 138
Tab.	15.8	Character arguments of post-processor built-in operators	
Tab.	15.9	Constant arguments of post-processor built-in operators	. 140

# **Figures**

Fig. 0.1	SimEnv system design	2
Fig. 4.1	Target space	10
Fig. 4.2		
Fig. 4.3	Behavioural analysis: Scanning multi-dimensional target spaces	12
Fig. 4.4	Sample for a Monte Carlo analysis	
Fig. 4.5	Sample for a sensitivity analysis	
Fig. 4.6	Part of a sample for an optimization experiment, generated during the experiment	
Fig. 5.1	Grid types	19
Fig. 5.2	Model variable definition: Grid assignment	23
Fig. 6.1	Monte Carlo analysis: Latin hypercube sampling	40
Fig. 7.1	Flowcharts for performing simenv.run and simenv.rst	52
Fia. 10.1	1 SimEnv user scripts and files	94

# Examples

Example 1.1	Example layout	4
Example 5.1 Example 5.2	Model output description file <model>.mdf</model>	
Example 5.3	Model output description file for a GAMS model	
Example 5.4	GAMS description file <model>.gdf</model>	
Example 5.5	GAMS description file for coupled GAMS models	
Example 6.1	Experiment description file <model>.edf for behavioural analysis</model>	38
Example 6.2	Experiment description file <model>.edf for Monte Carlo analysis</model>	41
Example 6.3	Experiment description file <model>.edf for local sensitivity analysis</model>	
Example 6.4	Experiment description file <model>.edf for an optimization experiment</model>	45
Example 7.1	Shell script <model>.ini for user-model specific experiment preparation</model>	
Example 7.2	Shell script <model>.run to wrap the user model</model>	
Example 7.3	Shell script <model>.end for user-model specific experiment wrap-up</model>	
Example 7.4	Shell script <model>.rst to prepare model performance during experiment restart</model>	50
Example 8.1	Addressing model output variables in model output post-processing	
Example 8.2	Checking rules for coordinates	
Example 8.3	Post-processing with advanced operators	
Example 8.4	Post-processing operator behav for behavioural analysis	
Example 8.5	Post-processing operators for Monte Carlo analysis	
Example 8.6	Post-processing operators for local sensitivity analysis	
Example 8.7 Example 8.8	User-defined operator description file <model>.odf User-defined macro definition file <model>.mac</model></model>	62 84
Example 10.1	User-defined general configuration file <model>.cfg</model>	
Example 11.1	Structure of a user-defined file	
Example 11.2	Examples of value lists	
Example 11.3	Coordinate transformations by a transformation file	
Example 12.1	IEEE compliant model output data structure	108
Example 15.1	Model interface for Fortran models - model world f.f	121
Example 15.2	Model interface for C models – model world_c.c	123
Example 15.3	Model interface for C++ models – model world_cpp.cpp	
Example 15.4	Model interface for Python models – model world_py.py	
Example 15.5	Model interface at shell script level – model shell script world_sh.run	127
Example 15.6	Model interface for GAMS models – model gams_model.gms	129
Example 15.7	Post-processor user-defined operator module – operator mat_mul	
Example 15.8	ASCII compliant post-processor result import interface	132



# **Executive Summary**

SimEnv is a multi-run simulation environment that focuses on model evaluation and usage mainly for quality assurance matters and scenario analyses using sampling techniques. Interfacing models to the simulation environment is supported for a number of programming languages by minimal source code modifications and in general at the shell script level. Pre-defined experiment types are the backbone of SimEnv, enabling experimenting with numerical parameter, initial value, or driving forces adjustments of the model. The resulting multi-run experiment can be performed sequentially or in parallel. Interactive experiment post-processing makes use of built-in operator definitions, optionally supplemented by user-defined operators and applies operator chains on model output and reference data. Result output functions generated during post-processing can be evaluated within SimEnv with advanced visualization techniques.

Simulation is one of the cornerstones for research in Global Change. The aim of the SimEnv project is to develop a toolbox oriented simulation environment that enables the modeller to handle model related quality assurance matters (Saltelli *et al.*, 2000, Saltelli *et al.*, 2004) and scenario analyses. Both research foci require complex simulation experiments for model inspection, validation and control design without changing the model in general.

SimEnv (Flechsig *et.al*, 2004) aims at model evaluation by performing simulation runs with a model in a coordinated manner and running the model several times. Co-ordination is achieved by pre-defined experiment types representing multi-run simulations.

According to the strategy of a selected experiment type for a set of so-called targets  ${\bf t}$  which represent drivers, parameters, boundary and initial values of the model  ${\bf M}$  a sample is generated before simulation and the targets t are re-adjusted numerically before each single simulation run during the experiment. Each experiment results in a sequence of model outputs over the single runs for selected state variables  ${\bf z}$  dependent on the target adjustments of the model M. Model outputs can be processed and evaluated across the run ensemble specifically after simulation.

The following experiment types form the base of the SimEnv multi-run facility:

#### Behavioural analysis

Inspection of the model's behaviour in a space spanned from targets t with discrete numerical adjustments and a flexible inspection strategy for the whole space.

For model verification, numerical validation, deterministic error analysis, deterministic control design, scenario analysis and spatial patch model applications.

#### Monte Carlo analysis

Perturbations of targets t according to probability density functions. Determination of moments, confidence intervals and heuristic probability density functions for z in the course of post-processing.

For error analysis, uncertainty analysis, verification and validation of deterministic models.

#### Local sensitivity analysis

Determination of model (state variable's z) local sensitivity to targets t. Is performed by finite difference derivative approximations from M.

For numerical validation purposes, model analysis, sub-model sensitivity.

#### Optimization

Iterative determination of optimal targets t for a cost functions derived from z by a simulated annealing methods.

For model validation (system - model comparison), control design, decision making.

SimEnv makes use of modern IT concepts. Model preparation for interfacing them to SimEnv is based on minimal source code manipulations by implementing function calls into Fortran-, C/C++-, Python- or GAMS-model source code for target adjustments and model output. Additionally, an interface at shell script level is available.

In experiment preparation an experiment type is selected and equipped numerically. Experiment performance supports local, remote, and parallel architectures.

Experiment-specific model output post-processing enables navigation in the experiment - model output space and interactive filtering of model output and reference data by application of built-in and user-defined post-processing operator chains.

Result evaluation is dominated by application of pre-formed visualization modules.

SimEnv model output as well as model output post-processing offer data interfaces for NetCDF, IEEE compliant binary and ASCII format for a more detailed post-processing outside SimEnv.

#### SimEnv key features:

- Support of key working techniques in experimenting with models:
   SimEnv enables model evaluation, uncertainty and scenario analyses in a structured, methodologically sound and pre-formed manner applying sampling techniques.
- Run ensembles instead of single model runs:
   Model evaluation by multi-run simulation experiments
- Availability of pre-defined multi-run simulation experiment types:
   To perform an experiment only the targets (parameters, drivers, initial values, ...) to experiment with and a strategy how to sample the target space have to be specified.
- Simple model interface to the simulation environment:
  There are model interface functions mainly to re-adjust an experiment target and to output model results for later post-processing. Model interfacing and finally communication between the model and SimEnv can be done at the model language level by incorporating coupling functions into model source code (C/C++, Fortran and Python) or can be done at the shell script level within shell-scripts. Additionally, there is a special interface for GAMS models.
- Support of distributed models: Independently on the kind distributed models are coupled they can be interfaced to SimEnv.
- Parallelization of the experiment: This is a prerequisite for a lot of simulation tasks.
  - Operator-based experiment post-processing:

    Chains of built-in and user-defined operators enable interactive experiment post-processing based on experiment model output and reference data including general purpose and experiment-specific operators.
- Graphical experiment evaluation: For post-processed model output
- Support of standard data formats:
   Output from the model as well from the post-processor can be stored in NetCDF or IEEE compliant binary format.

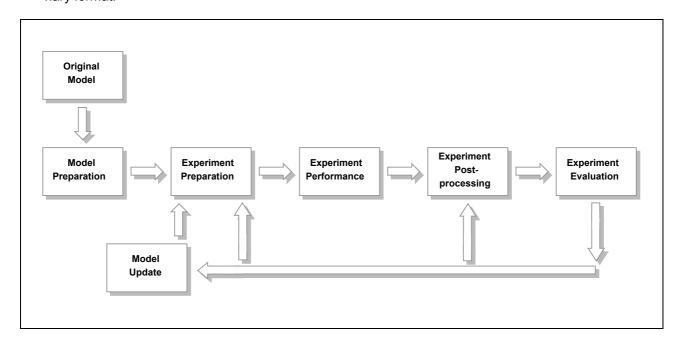


Fig. 0.1 SimEnv system design

#### 1 About this Document

In this chapter document conventions are explained. Within the whole document one reference example model is used to explain application of SimEnv. Examples are always located in grey boxes.

#### 1.1 Document Conventions

Character / string Meaning		
< >	angle brackets enclose a placeholder for a string	
{ } braces enclose an optional element		
[     ] square brackets enclose a list of choices, separated by a vertical b		
nil	stands for the empty string (nothing)	
monospace	indicates SimEnv example code	

Tab. 1.1 Document conventions

Tab. 1.2 summarizes the main placeholders used in this document.

Placeholder Description			
<file_name></file_name>	name of a data file		
<gams_model></gams_model>	name of a GAMS model		
<model></model>	model name to start a SimEnv service with		
<nil></nil>	the empty string		
<path></path>	path to a file		
<res></res>	integer post-processor output file number 1, 2,, 99		
<res_char></res_char>	character post-processor output file number 01, 02,, 99		
<run></run>	integer single run number 0, 1, within an experiment		
<pre><run_char></run_char></pre>			
<sep></sep>	sequence of white spaces as item separators in user-defined files		
<string></string>	any string		
<target_def_val></target_def_val>	default value of a target according to <model>.edf</model>		
<target_name> name of a target to experiment with</target_name>			
<pre><value_list></value_list></pre> list of values in explicit or implicit notation according to Tab. 11.4			

Tab. 1.2Placeholders in this document

# 1.2 Example Layout

All examples in this document refer to a hypothetical global simulation model world. It is to describe dynamics of atmosphere and biosphere at the global scale over 200 years. Lateral (latitudinal and longitudinal) model resolution differs for different model implementations (see below), temporal resolution is at decadal time steps. Additionally, atmosphere is structured vertically into levels.

The model world is assumed to map lateral and vertical (level) fluxes and demanding that's why for computing state variables for the whole globe.

The model world is a generic model. Model implementation in several programming languages results in models world\_<lng> where <lng> is an identifier for the programming language (and the lateral model resolution).

In the model pixel\_f state variables are calculated for one pixel (for one single latitude - lon-gitude constellation) without consideration of lateral fluxes.

Model state variable	Description	Defined on	Data type
atmo	aggregated atmospheric state	lat x lon x level x time	float
bios	aggregated biospheric state at land masses (defined between 83°N and 60°S latitude at land masses, i. e., without Antarctic)	lat x lon x time	float
atmo_g (not for model pixel_f)	aggregated global state derived from atmo for level 1	time	int
bios_g (not for model pixel_f)	aggregated global state derived from bios	-	int

Dynamics of all model variables depend on model parameters p1, p2, p3 and p4.

With the SimEnv release, documented in this User's Guide, the following model implementations are distributed:

	Model	Resolution			
Model	Model interface example for	lateral: lat x lon	vertical: number of levels	temporal: number of time steps	
world_f	Fortran	4 x 4	4: 1, 7, 11, 16	20	
world_c	С	4 x 4	4: 1, 7, 11, 16	20	
world_cpp	C++	4 x 4	4: 1, 7, 11, 16	20	
world_py	Python	4 x 4	4: 1, 7, 11, 16	20	
world_sh	script level	4 x 4	4: 1, 7, 11, 16	20	
pixel_f	Fortran	without, implicitly by experiment as 4 x 4	4: 1, 7, 11, 16	20	
world_f_1x1	Fortran	1 x 1	16: 1 - 16	20	

The only example that does not refer to the above model type is that for the GAMS model interface to SimEnv (see chapter 5.4 at page 26).

Examples are generally placed in grey-shaded boxes.

Examples that are available in the corresponding examples directory of the SimEnv release are marked as such in the lower right corner of an example box.

**Example 1.1** Example layout



# 2 Getting Started

In this chapter a quick start tour is described. Without going into details the user can get an impression how to apply SimEnv and which user files are essential to use the simulation environment.

- SimEnv is implemented under AIX at IBM's RS6000.
- Set the operating system environment variable **SE\_HOME** to /usr/local/simenv/bin in your .profile file and export it for the Korn shell.
- For interfacing Python and GAMS models to SimEnv extend your operating system environment variable **PYTHONPATH** by \$SE\_HOME, include it in your .profile file and export it for the Korn shell.
- Change to a working directory you have full access rights.
- Get basic information on SimEnv by entering

#### **\$SE\_HOME/simenv.hlp**

at the operating system prompt.

• Select a model implementation language < Ing> you want to check SimEnv with the model from Example 1.1 at page 4:

For a GAMS model example check chapter 5.6 at page 27.

Start from the working directory the shell script

#### \$SE\_HOME/simenv.cpy world\_<Ing>

to copy model world\_<lng> model and experiment related files to this working directory.

Copy the file world.edf\_c to world\_<lng>.edf

Check		for
<ul> <li>The SimEnv configuration file</li> </ul>	world_ <ing>.cfg</ing>	general configurations of SimEnv
<ul> <li>The model output description file</li> </ul>	ile world_ <ing>.mdf</ing>	available model variables
The model	world_ <ing>.<ing></ing></ing>	implementation of the model
<ul> <li>The model shell script</li> </ul>	world_ <ing>.run</ing>	wrapping the model executable
<ul> <li>The experiment description file</li> </ul>	e world_ <ing>.edf</ing>	experiment definition
<ul> <li>The post-processing input file</li> </ul>	world.post_c	post-processor expression sequence
<ul> <li>The macro description file</li> </ul>	world_ <ing>.mac</ing>	macros for the post-processor
<ul> <li>The operator description file</li> </ul>	world_ <ing>.opr</ing>	description of user-defined operators
The user-defined operators	usr opr <opr>.f</opr>	code of user-defined operator <opr></opr>

Start a complete SimEnv session by

**\$SE HOME/simenv.cpl** world < Ing> -1 world.post c

- · SimEnv files will be checked
- The experiment will be prepared
- The experiment will be performed machine (select the login machine on request)
- Model output post-processing will be started for this experiment
  - With the post-processing input file world\_post\_c and following
  - Interactively: Enter any expression and finish post-processing by entering a single <return>
- Visualization of post-processed results will be started

Start \$SE\_HOME/simenv.chk world\_<lng> to check model and experiment files.

- Start \$SE\_HOME/simenv.run world\_<lng> to prepare and perform a simulation experiment.
- Start \$SE\_HOME/simenv.rst world\_<lng> to restart a simulation experiment.
- Start \$SE\_HOME/simenv.res world <lng> {[ new | append | replace ]} {<run>}



(\*)

to post-process the last simulation experiment over the whole run ensemble or for run number <run> and to create a new / append to / replace the result file <model>.res<res\_char>.[ nc | ieee | ascii ] with the highest two-digit number <res\_char>. <res\_char> (can range from 01 to 99).

- Start \$SE\_HOME/simenv.vis world\_<lng> {[ latest | <res\_char> ]} (\*) to visualize output from the latest post-processing output file world\_<lng>.res<res\_char>.nc or that with number <res\_char> with the highest two-digit number <res\_char>. <res\_char> can range from 01 to 99.
- Check in the working directory the model interface and experiment performance log-files world <lng>.mlog and world <lng>.elog.
- Start **\$SE\_HOME/simenv.dmp** world\_<lng> | more to dump a SimEnv model or post-processor output file.
- Start **\$SE\_HOME/simenv.cln** world\_<lng> to wrap up a simulation experiment.
- Get the usage of all commands by entering a command without arguments.
- To run other simulation experiments and/or output in other data formats modify
  - world <lng>.cfg
  - world\_<lng>.edf
  - world <lng>.mdf
  - world <lng>.<lng> and/or
  - world <lng>.run
- To experiment with other models replace world\_<lng> by <model> as a placeholder for the name of any
  other model.

(\*): To get access rights for the visualization server check the SimEnv service \$SE\_HOME/simenv.key <user\_name>

in chapter 10.2 at page 89.



# **3 Version 1.11**

This chapter summarizes differences between the current and the previous SimEnv release, limitations, and bugs and their workarounds.

#### 3.1 What is New?

Туре	Check / see chapter	At page	Description	
new	4.5	16	New experiment type Optimization:	
	6.5	44	Minimize a cost function that is composed from model output variables	
	8.5.5	78	and consequently depends on experiment targets over a bounded tar-	
			get space by a simulates annealing method.	
update	9	85		
			- fast visualization track	
			- 3D visualization of 2D data	
			- additional histogram visualization of any n-dimensional result	
new	11.3	102	Result coordinate transformation file to incorporate external results from	
			the operator get_experiment into post-processing	
new	5.6.2	31	New sub-keyword time_limit for the gdf-file of GAMS models	
			Bug fixes	

Tab. 3.1SimEnv changes in version 1.11

Upgrade type	Upgrade action
mandatory	Re-link all models

Tab. 3.2User actions to upgrade to version 1.11

# 3.2 Known Bugs and Their Workarounds

Where / Bug / Workaround	Description			
Where	Experiment performance: Optimization / model output to a common NetCDF file for the whole experiment			
Bug	Write error			
Workaround	Specify IEEE model output or single NetCDF file output in <model>.cfg</model>			
Where	Experiment performance: Experiment restart / model output to a common NetCDF file for the whole experiment			
Bug	Read error in experiment post-processing			
Workaround	Specify IEEE model output or single NetCDF file output in <model>.cfg</model>			

Where / Bug / Workaround	Description			
Where	Experiment post-processing: Behavioural analysis / result output to NetCDF			
Bug	When applying operator behav non-monotonic target adjustments are transferred to NetCDF output in a wrong manner.			
Workaround	Specify only monotonic target adjustments in <model>.edf</model>			

Tab. 3.3Known bugs and their workarounds

### 3.3 Limitations

- Only accessible under Unix/AIX
- No C-interface to write user-defined operators for simenv.res
- Graphical user interface only for simenv.vis
- Experiment type optimization can not performed in parallel



# 4 Experiment Types

SimEnv supplies a set of pre-defined multi-run experiment types. Each experiment type addresses a special experiment class for performing a simulation model several times in a co-ordinated manner. In this chapter an overview on the available experiment types is given from the viewpoint of system's theory.

#### 4.1 General Approach

SimEnv supplies a set of pre-defined multi-run experiment types, where each type addresses a special multi-run experiment class for performing a simulation model or any algorithm with an input - output transition behaviour.

In the following, the general SimEnv approach will be described for time dynamic simulation models, because this class forms the majority of SimEnv applications. All information can be transformed easily to any other algorithm.

Based on systems' theory, each time dynamic model M can be formulated - without limitation of generality - for the time dependent, time discrete, and state deterministic case as

	M:	$Z(t)$ = ST ( $Z(t\text{-}\Delta t)$ ,, $Z(t\text{-}n^*\Delta t)$ , P , $X(t)$ , $Z_0$ , B )
with	ST Z P X Z <sub>0</sub> B t	state transition description state variables' vector parameter vector input (driving forces) vector initial value vector boundary value vector time
	$\Deltat$	time increment
	n	time delay

The output vector Y is a function of the state vector Z, parameters P, drivers X, and initial values Z<sub>0</sub>:

$$Y(t) = OU (Z(t), P, X(t), Z_0).$$

Model behaviour Z is determined for fixed n and  $\Delta t$  by state transition description ST, parameters P, driving forces X, initial values  $Z_0$ , and boundary values B. Manipulating and exploring model behaviour in any sense means changing these four model components. While state transition description ST reflects mainly model structure and is quite complex to change, each component of the driving forces vector X normally is a time-dependent vector.

Introduction of additional technical parameters  $P_{\text{tech}}$  can reduce the complexity of handling a model with respect to the five model components, described above: Changes in state transition description ST can be predetermined in the model by assigning values of a technical parameter  $p_{\text{tech}}$  to alternative submodel versions, which are switched on or off by these values. Additionally, each component of the driving forces vector X can be combined with technical parameters in different ways:

- By selecting special driving forces dependent on the technical value
- By manipulating the driving forces with the parameter value (e.g., as an additive or multiplicative adjustment)
- · By parametrizing the shape of a driving force

When this has been done, the model behaviour finally depends only on the parameters P, the initial values  $Z_0$ , and the boundary values P. From the methodical point of view there is no difference between parameters, initial values and boundary values, because all are considered as constant during one model run. That

is why in the following the term target stands as a placeholder for all the four model components parameters, drivers, initial values and boundary values. All targets form the target set T:

$$T = \{ P, X, Z_0, B \}$$

and

$$Z = ST(T)$$
.

In the following,

$$T_m = (t_1, ..., t_m)$$
  $m > 0$ 

stands for a subset of the target set T that spans up an m-dimensional sub-space of T by selected model targets (t<sub>1</sub>,..., t<sub>m</sub>) from T and

$$T_{mn} = \begin{pmatrix} t_{11} & \dots & t_{1m} \\ \dots & & \dots \\ t_{n1} & \dots & t_{nm} \end{pmatrix} \qquad m > 0, \ n > 1$$

stands for a numerical sample for T<sub>m</sub> of size n and finally for m\*n values representing in any sense T<sub>m</sub>. In the set of all  $T_{mi}$  (i > 1) one extraordinary sample  $T_{m1}$  exists that matches the nominal (default) numerical target constellation for the model M.

If  $\{ \}_n$  denotes the dynamics of the model M over a sample of size n then it yields:

$$\{Z\}_n = \{ST(T_{mn})\}_n$$
.

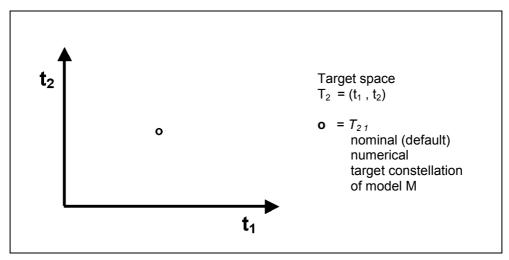


Fig. 4.1 Target space

SimEnv supports different sampling stategies and performance of multi-run experiments where m targets are readjusted numerically for each of n single simulation runs. Central goal is to study dependency of the model dynamics on target adjustments. For simulation purposes in SimEnv experimentation with the model M over T<sub>mn</sub> is based on the assumption that dynamics of M for each representative from the sample is indepent from all other representatives, which is fulfilled in general. This results in the possibility to form a run ensemble for performing the model M with n single model runs from the sample T<sub>mn</sub>.

SimEnv experiment types differ in the way T<sub>m</sub> is sampled to get T<sub>mn</sub>. There are deterministic and nondeterministic sampling strategies that offer a broad range of techniques for

- Experimentation with models
- Post-processing model output results
- Interpreting results with respect to uncertainty and sensitivity matters of models.

The experiment types are described in detail in the following.

# 4.2 Behavioural Analysis

Behavioural analysis uses a deterministic strategy to sample  $T_m$ . It is the inspection of the model in the target space  $T_m$  where inspection points are set in a regular and well structured manner.

Behavioural analysis can be interpreted and used in different ways:

- For scenario analysis: to show how model behaviour changes with changes of target values
- For numerical validation purposes: to determine target values in such a way that the output vector matches with measurement results of the real system
- For deterministic error analysis: to analyse how the model error is dependent on target errors
- For a simulation-based control design: to determine target values in such a way that a goal function becomes an extreme

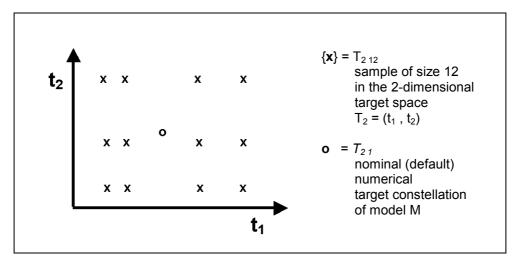


Fig. 4.2 Sample for a behavioural analysis

SimEnv behavioural analysis sampling strategy is a generalization of the one-dimensional case for  $T_1$ , where the model behaviour is scanned in dependence on deterministic adjustments of one target  $t_1$ . The general case for  $T_m$  demands a strategy for scanning m-dimensional spaces in a flexible manner. Based on the predecessors of SimEnv (Wenzel *et al.*, 1990, Wenzel *et al.*, 1995, Flechsig, 1998) subspaces of the m-dimensional target space can be scanned on the subspace diagonal (parallel in a one-dimensional hyperspace) or completely for all dimensions (combinatorially on a grid) and both techniques can be combined. Besides this regular scanning method an irregular technique is possible.

The resulting number of single simulation runs for the experiment depends on the number of target samples per dimension of the scanned target space and from the selected scanning method. An experiment is described by the names of the involved targets, their numerical adjustments and their combination (scanning method). Model output post-processing resolves the scanning method again and outputs results as projections on multi-dimensional target subspaces.

Fig. 4.3 describes the regular scanning technique by an example. In the left scheme (a) the two-dimensional target space  $T_2 = (p_1, p_2)$  is scanned combinatorially, resulting in 4\*4 = 16 model runs, while the middle scheme (b) represents a parallel scanning of these two targets at the diagonal by 1+1+1+1 = 4 model runs. The scheme (c) at the right side shows a complex scanning strategy of the 3-dimensional target space  $T_3 = (p_1, p_2, p_3)$  with (1+1+1+1)\*3 = 12 model runs. Each filled dot represents a single model run.

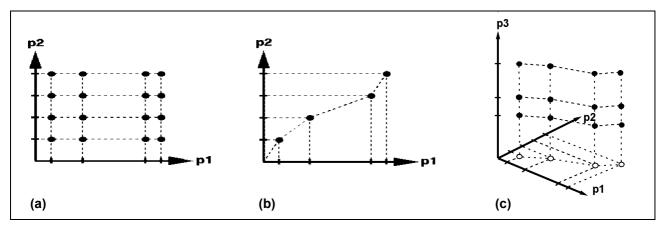


Fig. 4.3 Behavioural analysis: Scanning multi-dimensional target spaces

### 4.3 Monte Carlo Analysis

Monte Carlo analysis uses a non-deterministic strategy to sample  $T_{mn}$ . A Monte Carlo experiment in SimEnv is a perturbation analysis with pre-single run target perturbations.

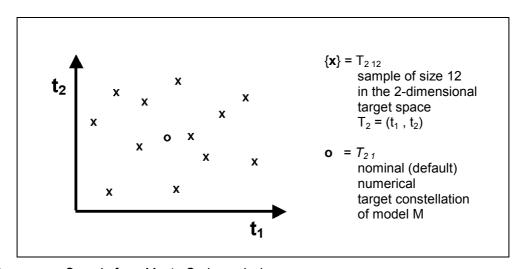


Fig. 4.4 Sample for a Monte Carlo analysis

Theoretically, with a Monte Carlo analysis moments of a state variable z can be computed as

$$M^{(k)}\{z\} = \int_{-\infty}^{\infty} z(T_m)^k \cdot pdf(T_m) \ dT_m$$
 with 
$$M^{(k)}\{z\} \qquad \qquad k\text{-th moment of the state variable } z \text{ with respect to the probability density function pdf}$$
 
$$z(T_m) \qquad \qquad \text{state variable } z \text{ as a function of } T_m \\ pdf(T_m) \qquad \qquad \text{probability density function of } T_m$$

By interpreting the probability density function  $pdf(T_m)$  as the error distribution in the target space  $T_m$  it is possible to study error propagation in the model. On the other hand Monte Carlo analysis can be interpreted as a stochastic error analysis, if there are measurements of the real system for z.

For a numerical experiment in SimEnv it is assumed that the probability density function  $pdf(T_m)$  can be decomposed into independent probability density functions  $pdf_i$  for all targets  $t_i$  of  $T_m$ :

$$pdf(T_m) = \prod_{i=1}^{m} pdf_i(t_i)$$

and the m-dimensional integral is approximated by a sequence of n single simulation runs of the model where the numerical target values  $t_{ij}$  of  $t_i$  ( $1 \le i \le m$ ,  $1 \le j \le n$ ) are sampled according to the probability density function  $pdf_i$ .

On the basis of these assumptions, the statistical measures in Tab. 4.1 can be computed during performance of a post-processing session from a Monte Carlo analysis with n simulation runs resulting in n realizations  $z_1, ..., z_n$  of the model's state variables  $z, z_1$  and  $z_2$ :

Statistical measure		Definition (*)
minimum	min(z)	= min (z <sub>i</sub> )
maximum	max(z)	= max (z <sub>i</sub> )
sum	sum(z)	$= \sum z_i$
arithmetic mean	avg(z)	$= \sum z_i / n$
variance	var(z)	= $\sum (z_i - avg(z))^2 / (n - 1)$
skewness	z <sup>(3)</sup>	= $\sum (z_i - avg(z))^3 / n * (\sum (z_i - avg(z))^2 / (n - 1))^{3/2}$
kurtosis	z <sup>(4)</sup>	= $(\sum (z_i - avg(z))^4 / n * (\sum (z_i - avg(z))^2 / (n - 1))^2) - 3$
range	rng(z)	= max(z) - min(z)
geometric mean	avgg(z)	$= (\prod z_i)^{1/n}$
harmonic mean	agvh(z)	= $n / \sum (1 / z_i)$
weighted mean	avgw(z)	= $\sum z_i * w_i / \sum w_i$ w: weight
correlation	cor(z1,z2)	$= \sum_{i=1}^{n} (z1_{i} - avg(z1)) * (z2_{i} - avg(z2)) /$
covariance	cov(z1,z2)	$\sqrt{\sum (z1_i - avg(z1))^2 * \sum (z2_i - avg(z2))^2}$ = \(\sum (z1_i - avg(z1)) * (z2_i - avg(z2)) / (n - 1)
linear regression coefficient		= $(\Sigma(z1_i - avg(z1))*(z2_i - avg(z2)))/$ $(\Sigma(z1_i - avg(z1))^2)$
median	med(z)	= middle value from increasingly ordered { z <sub>i</sub> } (n = odd) mean of the two middle values from { z <sub>i</sub> } (n = even)
quantile	qnt <sup>(p)</sup> (z)	= that value from increasingly ordered { z <sub>i</sub> } which corresponds to a cumulative frequency of n*p qnt <sup>(0.5)</sup> (z) = med(z)
confidence interval boundaries	$cnf^{(lpha)}(z)$	= $avg(z) \pm t_{\alpha,n-1} \sqrt{var(z)/n}$ $\alpha$ : level of error $t_{\alpha,n}$ : significance boundaries of Student distribution
heuristic probability density function	hgr <sup>(class)</sup> (z)	= number of $z_i$ with $class_{min} \le z_i < class_{max}$ $class_{min}$ , $class_{max}$ : boundaries of equidistant classes

Tab. 4.1Statistical measures

(\*): indices for sums  $\Sigma$ , products  $\Pi$  and extremes run from 1 to n:  $\sum_{i=1}^{n} \prod_{j=1}^{n} \min_{i=1}^{n} \max_{j=1}^{n} \sum_{i=1}^{n} \min_{j=1}^{n} \max_{i=1}^{n} \sum_{j=1}^{n} \min_{i=1}^{n} \max_{j=1}^{n} \sum_{i=1}^{n} \sum_{j=1}^{n} \sum_{j=1}^{n} \sum_{i=1}^{n} \sum_{j=1}^{n} \sum_{j=1}^{n} \sum_{j=1}^{n} \sum_{i=1}^{n} \sum_{j=1}^{n} \sum_{j=1$ 

Tab. 4.2 summarizes these probability density functions (Bohr, 1998) that are pre-defined in SimEnv for targets to be perturbed. Additionally, SimEnv offers to import random number samples in the course of experiment preparation.

Distribution	Short- cut	Probability density function	Distr	ibution parameters	
uniform	U(a,b)	$pdf(x) = \frac{1}{b-a}$ $pdf(x) = 0$	if $x \in [a,b]$	a b	lower boundary upper boundary > a
			otherwise	it is:	mean = $(a+b) / 2$ standard deviation = $\sqrt{(b-a)^2 / 12}$
normal	$N(\mu, \sigma^2)$	$pdf(x) = \frac{1}{\sigma\sqrt{2\pi}} exp\left(-\frac{(x-\mu)^2}{2\sigma^2}\right)$		$\mu$ $\sigma$	mean standard deviation > 0
lognormal	$L(\mu, \sigma^2)$	$pdf(x) = \frac{1}{x\sigma\sqrt{2\pi}} exp\left(-\frac{(\ln x - \mu)^2}{2\sigma^2}\right)$	if x > 0	μ σ	> 0
		pdf(x) = 0	otherwise	it is:	$ln(x) \sim N(\mu, \sigma^2)$
exponential	Ε(μ)	$pdf(x) = \frac{1}{\mu} exp\left(-\frac{x}{\mu}\right)$	if x > 0	μ	mean > 0
		pdf(x) = 0	otherwise	it is:	standard deviation = $\mu$

Tab. 4.2Probability density functions

The number of runs to be performed during a Monte Carlo analysis has to be specified. An experiment is described by the targets involved in the analysis, their distribution and the appropriate distribution parameters.

# 4.4 Local Sensitivity Analysis

Local sensitivity analysis uses a deterministic sampling stategy in  $\epsilon$ -neighbourhoods of the numerical default constellation  $T_{m1}$  of the model M. For each target  $t_i$  from the nominal target constallation  $T_{m1}$  and each  $\epsilon_j$  from the  $\epsilon$ -neighbourhoods ( $\epsilon_1$ ,...,  $\epsilon_k$ ) two members ( $t_1$ ,...,  $t_{i-1}$ ,  $t_i \pm \epsilon_j$ ,  $t_{i+1}$ ,...,  $t_m$ ) of the resulting sample are generated. The sample size n is given by 2\*m\*k. Running the model at this sampling set serves to determine sensitivity functions.

In classical systems' theory, model sensitivity of a model state variable z with respect to a target t is the partial derivative of z after t  $\delta z/\delta t$ . In the numerical simulation of complex systems a finite sensitivity function is preferred, because it can be obtained without model enlargements or re-formulations. It is a linear approximations of the classical model sensitivity measure (Wierzbicki, 1984).

Local sensitivity measures as well as measures which reflect model output linearity and/or symmetry nearby  $T_{\rm m1}$  can be used for localizing modification-relevant model parts as well as control-sensitive targets in control problems. On the other hand, identification of robust parts of a model or even complete robust models makes it possible to run a model under internal or external disturbances. Sensitivity analysis in SimEnv post-processing is based on finite sensitivity, linearity, and symmetry measures, which are defined as in Tab. 4.3.

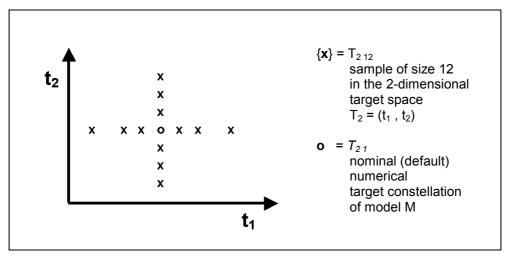


Fig. 4.5 Sample for a sensitivity analysis

Local	Definition					
measure	Absolute measure	Relative measure				
sensitivity measure	sens_abs(z,± $\epsilon$ ) = $\frac{z(t \pm \epsilon) - z(t)}{\pm \epsilon}$	$sens_rel(z,\pm\epsilon) = sens_abs(z,\pm\epsilon)\frac{t}{z(t)}$				
linearity measure	$lin\_abs(z,\varepsilon) = \frac{(z(t+\varepsilon)-z(t))+(z(t-\varepsilon)-z(t))}{\varepsilon}$	$\lim_{r \to \infty} rel(z, \varepsilon) = \lim_{r \to \infty} s(z, \varepsilon) \frac{t}{z(t)}$				
symmetry measure	$sym\_abs(z,\varepsilon) = \frac{z(t+\varepsilon)-z(t-\varepsilon)}{\varepsilon}$	$sym_rel(z,\varepsilon) = sym_abs(z,\varepsilon) \frac{t}{z(t)}$				

**Tab. 4.3** Local sensitivity, linearity, and symmetry measures for a selected target t from  $T_{m1}$  and a selected  $\varepsilon$  from  $(\varepsilon_1, ..., \varepsilon_n)$ 

Accordingly, local measures of the model with respect to a target are always expressed as a measure of a model's state variable z, usually at a selected time step within a surrounding neighborhood  $\epsilon$  of a target value t. That is why the conclusions drawn from a local sensitivity analysis are only valid locally at  $T_{m1}$  with respect to the whole target space  $T_m$ . Additionally, local measures only describe the influence of one target  $t_i$  from the whole vector  $T_m$  on the model's dynamics.

As stated above, the sensitivity measures reflect the classical sensitivity functions in a neighborhood of  $T_{m1}$ . The larger the absolute value of the measure the higher is the influence of an incremental change of the target t on the model output z. The linearity measures map the linear behaviour of z nearby  $T_{m1}$ . If the linear measure is zero z shows a linear behaviour with respect to t. The symmetry measures measures map the symmetric behaviour of the z nearby  $T_{m1}$ . If the symmetry measure is zero z shows a symmetric behaviour with respect to t. The larger the absolute values of the latter two measures the higher is the nonlinear / non-symmetric behaviour of z with respect to t.

The absolute measures are best suited to compare the influence of different targets {t} on the same state variable z while due to their normalization factor the relative measures enable comparison of the influence of one target t on different state variables {z}.

From the local measures of table Tab. 4.3 additional measures can be derived on demand, e.g.,  $abs(sym\_abs(z, \epsilon))$ .

A local sensitivity experiment is described by the names of the targets t to be involved and the increments  $\epsilon$ . The number of runs for the experiment results from the number of targets and increments: two runs per target for each increment plus one run with the default values of the targets. Local sensitivity functions are calculated during model output post-processing.

#### 4.5 Optimization

The optimization experiment in SimEnv uses a stochastic strategy to sample  $T_m$ . It is the only experiment type where the sample is generated during experiment performance and not at experiment preparation. The general approach of optimization is to find the global minimum of a cost function (synonym: objective function)

$$F(Z) = F(ST(T_m))$$

that depends on model's state variables Z and consequently on the experiment targets  $T_m = (t_1, ..., t_m)$ :

 $\begin{array}{ll} \text{minimize} & F(t_1 \;,..., \; t_m) \\ \text{subject to} & t_{i \; \text{min}} \leq t_i \leq t_{i \; \text{max}} \; \; \text{for i = 1 },..., \; m \\ \end{array}$ 

Often, F represents a distance measure in a specific metric between selected model state variables and reference data (measurement values of the real system or simulation results from an other model). Consequently, optimization can be used for model validation and control design to find optimal values of model targets in such a way that model state variables are close to reference data. In SimEnv the cost function is specified in result post-processing as an expression formed from model output (and reference data) where an operator chain is applied on (check chapters 6.5 and 8). The value of the cost function is calculated directly after the current single run has been performed.

SimEnv uses a gradient free optimization approach that is called "Simulated Annealing" and is a generalization of a Monte Carlo method for examining the equations of state of n-body systems. The concept is based on the manner in which metals recrystalize in the process of annealing. In an annealing process a melt, initially at high temperature Temp and disordered, is slowly cooled so that the system at any time is approximately in thermodynamic equilibrium. As cooling proceeds, the system becomes more ordered and approaches a "frozen" ground state at Temp = 0. Hence the process can be thought of as an adiabatic approach to the lowest energy state E. If the initial temperature of the system is too low or cooling is done insufficiently slowly the system may become quenched forming defects or freezing out in metastable states (i.e. trapped in a local minimum energy state).

The annealing scheme is that an initial state of a thermodynamic system is chosen at energy E and temperature Temp, holding Temp constant the initial configuration is perturbed and the change in energy dE is computed. If the change in energy is negative or zero the new configuration is accepted. If the change in energy is positive it is accepted with a probability given by

$$p = exp(-dE/(k_B*Temp))$$

where  $k_B$  denotes the Boltzmann constant. This process is then repeated sufficient times to give good sampling statistics for the current temperature, and then the temperature is decremented and the entire process repeated until a frozen state is achieved at Temp = 0.

By analogy the generalization of this Monte Carlo approach to optimization problems is straight forward:

- The current state of the thermodynamic system is analogous to the current solution to the optimization problem
- The energy equation for the thermodynamic system is analogous to the objective function F, and
- The ground state at Temp = 0 is analogous to the global minimum of F.

The major difficulty (art) in implementation of a simulated annealing algorithm is that there is no obvious analogy for the temperature Temp with respect to a free parameter in the optimization problem. Furthermore, avoidance of entrainment in local minima (quenching) is dependent on the "annealing schedule", that is, the choice of initial temperature, how many iterations are performed at each temperature, and how much the temperature is decremented at each step as cooling proceeds (after Gray *et al.*, 1997). Ideally, when local optimization methods are trapped in a poor local minimum, simulated annealing can 'climb' out.

The algorithm applied in SimEnv is a very fast simulated re-annealing method, named Adaptive Simulated Annealing ASA (Ingber 2004, Ingber 1989 and Ingber 1996). For the above stated probability p the term  $k_B$  \* Temp is chosen as



$$k_B$$
 \* Temp = Temp<sub>0</sub> \* exp(-ck<sup>1/m</sup>)

where k is the annealing time.

The ASA schedule is much faster than Boltzmann annealing, where  $k_B$  \* Temp = Temp $_0$  /ln(k) and faster than fast Cauchy annealing, where  $k_B$  \* Temp= Temp $_0$  / k. With the ASA method the global minimum of a nonlinear non-convex cost function F over an m-dimensional bounded target space  $T_m$  is determined.

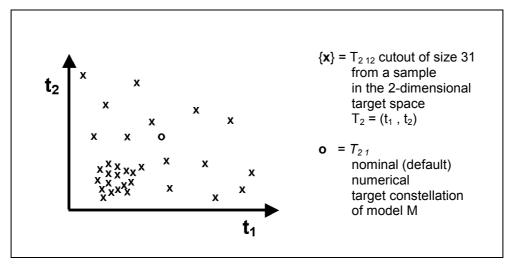


Fig. 4.6 Part of a sample for an optimization experiment, generated during the experiment

#### 5 Model Interface

To use any model within SimEnv it has to be coupled to the simulation environment. SimEnv offers easy coupling techniques at programming language and shell script level. While at language level SimEnv function calls have to be implemented into model source code to adjust experiment targets, i. e. model parameters, initial values or boundary values of the current single run out of the run ensemble numerically and to output simulation results, at the shell script level communication between the simulation environment and the model can be based on operating system information exchange methods. To plug the model into the simulation environment the variables of the model to be output during experiment performance and to be post-processed during model output processing have to be declared in the model output description file <model>.mdf. Additionally, the model itself has to be wrapped into a shell script <model>.run.

Model interfacing is related to transferring adjusted numerical values of model targets under investigation from the simulation environment to the model and to transferring model variables under investigation from the model to the simulation environment for later post-processing. Coupling is supported at the programming language level for C/C++, Fortran, Python, and GAMS programming languages, the model is implemented in and the shell script shell script level.

### 5.1 Coordinate and Grid Assignments to Variables

To each variable

Dimensionality dim(variable)

Extents ext(variable,i) with i=1,...,dim(variable)
 Coordinates coord(variable,i) with i=1,...,dim(variable)

are assigned to. The dimensionality is the number of dimensions, an extent is related to each dimension and represents the number of elements in that dimension. Extents are always greater than 1. To each dimension a coordinate is assigned to. Coordinates have a name and from all coordinate values the coordinate is defined for a subset is assigned to the extent of the dimension of the variable. Variables of dimensionality 0 do not have a coordinate assignment.

A variable of dimensionality n corresponds with an n-dimensional array, a variable of dimensionality 0 is a scalar.

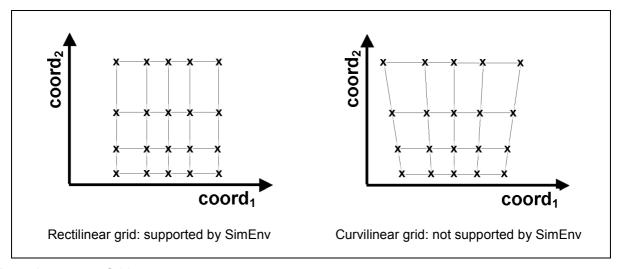


Fig. 5.1 Grid types

Additionally, coordinate axes are defined. Each coordinate axis a strictly monotonic sequence of coordinate values, a description and a unit is assigned to. For reasons of simplification in model output post-processing coordinate axes are assumed as curvilinear.

Each dimension of a variable with a dimensionality > 0 a complete coordinate axis or a part of a coordinate axis is assigned to. Consequently, each variable with a dimensionality > 0 is defined on a coordinate system formed from the assigned coordinates. For reasons of simplification in result evaluation with visualization techniques coordinate systems are assumed as rectilinear (orthogonal with variable distances between adjacent coordinate values). The model variable values then exist on the grid, spanned up from the coordinate values of the coordinate axes (see Fig. 5.1).

Since coordinate axes can be assigned to model variable dimensions in a flexible manner, model variables can exist on the same coordinate system or completely or partially disjoint coordinate systems.

## 5.2 Model Output Description File <model>.mdf

In the model output description file <model>.mdf the model variables are declared that are to be output by a SimEnv model coupling interface function in the model (code) and are to be post-processed after experiment performance. Additionally, coordinate axes are defined and flexibly assigned to model variables. Consequently, a model variable always is defined on a coordinate system, formed from the assigned coordinates to the variable.

<keyword></keyword>	<name></name>	<sub- keyword&gt;</sub- 	Line type	Max. line nmb.	<info></info>	Explanation
mdf	<nil></nil>	descr	0	any	<string></string>	model description
coordinate	coordinate_	descr	0	1	<string></string>	coordinate axis description
	name	unit	0	1	<string></string>	coordinate axis unit
	(co_name)	values	m	1	<value_list></value_list>	strictly monotonic sequence of coordinate values co_vals (for syntax see Tab. 11.4)
variable	variable_	descr	0	1	<string></string>	variable description
	name	unit	0	1	<string></string>	variable unit
		type	m	1	see Tab. 5.2	variable type in the simulation model
		coords	c1	1	co_name <sub>1</sub> , , co_name <sub>n</sub>	assigns a coordinate axis by its name to each dimension of the variable. Determines in this way implicitly the dimensionality n of the variable.
		coord_extents	c2	1	co_val <sub>11</sub> :co_val <sub>12</sub> , , co_val <sub>n1</sub> :co_val <sub>n2</sub>	assigns start and end coordinate value from each coordinate axis to the variable. If missing all coordinate values will be used from all assigned coordinates.
		var_extents	c1	1	vi_ext <sub>11</sub> :vi_ext <sub>12</sub> , , vi_ext <sub>n1</sub> :vi_ext <sub>n2</sub>	assigns start and end index for each dimension to the variable. Indices can be used to address the variable during post-processing.

 Tab. 5.1
 Elements of a model output description file <model>.mdf



Each model variable has a name, a dimensionality and assigned extents, a data type, a description and a unit. The name should correspond with the name of the variable in the simulation model code. Association between these two names is achieved by the SimEnv coupling function simenv\_put\_\* (see below).

<model>.mdf is an ASCII file that holds this information. It follows the coding rules in chapter 11 at page 99 with the keywords, names, sub-keywords, and info as in Tab. 5.1.

To Tab. 5.1 the following additional rules and explanations apply:

- For the description of line type check Tab. 11.3 at page 100.
- Coordinate and variable names must differ from target names in experiment description (see chapter 6.1) and from built-in and user-defined operator names for model output post-processing (see chapter 8.6.2).
- Assignment of coordinate axes to variable dimensions and consequently of a grid to a variables is only
  valid for model output post-processing. Normally, the simulation model itself will also exploit the same
  grid structure. Nevertheless, the grid structures of the model are defined autonomously in the model in a
  explicit or implicit manner and do only correspond with the grid structure in the model output description
  file symbolically.
- Model variables with dimensionality 0 are not assigned to a coordinate axis.
- The values of a coordinate have to be ordered in a strictly monotonic sequence. They may be non-equidistant and may be ordered in a decreasing sequence.
- With the sub-keyword coord\_extents only a portion of coordinate values of a coordinate axis can be assigned to a dimension of a variable. This portion is addressed by its begin and end value co\_val<sub>i1</sub> and/or co\_val<sub>i2</sub>. The number of coordinates values of the portion has to be greater than 1. co\_val<sub>i1</sub> > co\_val<sub>i2</sub> for strictly increasing values of coordinates co\_val<sub>i1</sub> < co\_val<sub>i2</sub> for strictly decreasing values of coordinates
- With the sub-keyword **var\_extents** portions of variables are made addressable during SimEnv post-processing. In the same way multi-dimensional variables are equipped with indices in the simulation model they also have an index description in the model output description file for purposes of model output post-processing. It is advisable, that these two descriptions coincide. The index range is described by a start and an end index vi ext<sub>i1</sub> and/or vi ext<sub>i2</sub>.

The index set is a strictly increasing, equidistant set of integer values with an index increment of 1,  $vi ext_{i1} < vi ext_{i2}$ ,

 $vi_ext_{i1} \le 0$  is possible.

- Coordinate values and index values are assigned in a one-to-one manner.
- For multi-dimensional variables that do not exist on an assigned grid completely or partially, simply assign formal coordinate axes to.
- Specify at least one model output variable in <model>.mdf.

SimEnv data type			Description		Restriction
byte	or	int*1	1 byte	integer	not for Python models
short	or	int*2	2 bytes	integer	not for Python models
int	or	int*4	4 bytes	integer	
float	or	real*4	4 bytes	real	
double	or	real*8	8 bytes	real	not for Python models

Tab. 5.2 SimEnv data types

For the following example of a model output description file and the assigned grid for model variable bios check Example 1.1 at page 4:

mdf		descr	World with a resolution of
mdf		descr	4° lat x 4° lon x
mdf		descr	4 levels x 20 time steps
mdf		descr	Data centred per lat-lon cell
			=
mdf		descr	This file is valid for all models
mdf			world_[ f   c   cpp   py   sh ]
	1 - +	1	
coordinate	lat	descr	geographic latitude
coordinate	lat	unit	deg
coordinate	lat	values	equidist_end 88(-4)-88
coordinate	lon	descr	geographic longitude
coordinate	lon	unit	deg
coordinate	lon	values	equidist_end -178(4)178
	1 1	daaaa	atmoonhouistisal lassl
coordinate	level	descr	atmospheric vertical level
	level	unit	level no
coordinate	level	values	list 1,7,11,16
coordinate	time	descr	time in decades
coordinate	time	unit	10 years
coordinate	time	values	equidist_nmb 1(1)20
		1	
variable	atmo	descr	aggregated atmospheric state
variable	atmo	unit	without
variable	atmo	type	float
variable	atmo	coords	lat , lon , level , time
variable	atmo	var_extents	1:45 , 1:90 , 1:4 , 1:20
variable	bios	descr	aggregated biospheric state
variable	bios	unit	g/m <sup>2</sup>
variable	bios	type	float
variable	bios	coords	lat , lon , time
variable	bios		84.:-56. , -178.:178. , 1:20
variable	bios	var extents	1:36 , 1:90 , 1:20
. 3223320	3200		, 2.35
variable	atmo_g	type	int
variable	atmo_g	coords	time
variable	atmo_g	var_extents	1:20
variable	bios g	type	int
			Example-file: world_[ f   c   cpp   py   sh ].mdf

**Example 5.1** Model output description file <model>.mdf

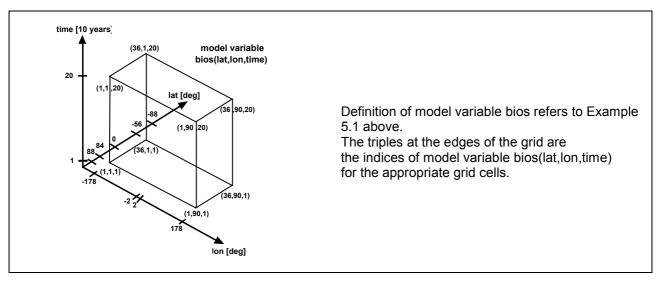


Fig. 5.2 Model variable definition: Grid assignment

#### 5.3 Model Interface for Fortran and C/C++ Models

Tab. 5.3 describes the functions that can be used in user models written in Fortran or C/C++ to adjust experiment targets for the current single run of the run ensemble and to output model results from the current single run. Two additional functions are responsible to initialize and/or finish SimEnv model coupling interactions.

Finally, two other model coupling functions are available: One function can be used to get the number of the current single run and an other to announce output of a slice of the data of a defined model variable. The latter is good for models with multi-dimensional variables where at least one dimension is omitted in the model's variable declaration because the dynamics for this dimension is calculated in place (e.g., time). The assigned variable then has a lower dimensionality than the corresponding variable in the model output description file. Nevertheless the slice-function ensures that model output over the omitted dimension can be handled in model output post-processing in common.

Model coupling functions are generic. To distinguish between the programming languages function names have a language suffix \_f (for Fortran) and \_c (for C/C++). All functions have a 4-byte integer function value (integer\*4 and/or int). Implementation of the functions for C/C++ is based on a call by reference for the function arguments.

In Tab. 5.3 input and output data types are documented for functions used in Fortran. For C/C++ the corresponding data types are valid.

Function name without suffix _f, _c	Function description	Arguments / function value	Argument / function value description
simenv_end ( )	close model coupling interface  Apply always after the last call of the other SimEnv functions in the model	integer*4 simenv_end (function value)	return code = 0 ok = 2 I/O error for model output file
simenv_get ( target_name, target_val_def, target_val_adj )	gets the numerical modification (adjustment) for the target (parameter / initial value / boundary value) to be experimented with in the current single	character*(*) target_name (input) real*4 target_val_def (input) real*4 target_val_adj (output)	name of the target in <model>.edf  nominal / default (non-adjusted) target value.  If target_name is not defined in <model>.edf then target_val_adj is set to target_val_def adjusted target value</model></model>
	run	integer*4 simenv_get (function value)	return code = 0 ok = 1 target_name undefined:     target_val_adj := target_val_def = 3 For Monte Carlo analysis:     warning w.r.t. distribution paramter adjustment     (check Tab. 6.6 at page 41)
simenv_get_run ( run_int, run_char	gets run number of current run as an integer value and a character string	character*6 run_char (output) integer*4 run_int (output) integer*4 simenv_get_run (function value)	current run number with leading zeros  current run number  return code = 0 ok
simenv_ini ( )	initialize model coupling interface  Apply always before the first call of the other SimEnv functions in the model	integer*4 simenv_ini (function value)	return code = 0 ok = 2 I/O error for model output file = 3 error memory allocation = 4 I/O error for <model>.edf_bin = 5 I/O error for <model>.mdf_bin = 6 I/O error for <model>.edf_adj = 7 wrong single run number</model></model></model>
simenv_put ( var_name, field )	outputs model results to native SimEnv output file(s)	character*(*) var_name (input) dimension field(), type according to <model>.mdf (input) integer*4 simenv_put (function value)</model>	name of the variable in <model>.mdf to be output  data of variable var_name to be stored as simulation results  return code = 0 ok = 1 var_name undefined = 2 I/O error model output file</model>



Function name without suffix _f, _c	Function description	Arguments / function value	Argument / function value description
simenv_slice ( var_name,	announces to out- put at next simenv_put call	character*(*) var_name (input)	name of the variable in <model>.mdf to be sliced</model>
idim, ifrom, ito	only a slice of variable var_name. This announce-	integer*4 idim (input)	dimension to be sliced
)	ment becomes inactive after performance of the	integer*4 ifrom (input)	slice to start at position ifrom
	appropriate simenv_put	integer*4 ito (input)	slice to end at position ito
		integer*4 simenv_slice (function value)	return code = 0 ok = 1 var_name undefined
			<ul> <li>= 3 inconsistency between variable and idim, ifrom, ito</li> <li>= 4 slice storage exceeded</li> <li>= 5 warning: slice overwritten</li> </ul>

Tab. 5.3Model coupler functions at language level

- Make sure consistency of type and dimension declarations between the model variables in model source code and the corresponding variable declarations in the model output description file <model>.mdf.
- Model variables that are not output completely or partially within the user model are handled in result-post-processing as their corresponding nodata-values (see Tab. 10.11).
- Application of simenv\_slice\_\* for NetCDF model output may result in a higher consumption of computing time for each single run of the experiment compared with NetCDF model output without simenv\_slice\_\*.
   For this case, keep in mind the trade-off between the demand for computing time and the demand for main memory.

In Example 15.1 at page 121 the model world\_f.f and in Example 15.2 at page 123 the model world\_c.c are explained.

# 5.4 Model Interface for Python Models

Due to the special features of Python the coupling interface to SimEnv differs from that for Fortran and C/C++ in chapter 5.3. Additionally, Python supports only some data types (check Tab. 5.2). Tab. 5.4 summarizes the coupling functions to use for a Python model.

Function name	Function description	Arguments / function value	Argument / function value description
simenv_ end_py (	close model cou- pling interface		
simenv_ get_py ( target_name, target_def_val)	gets the numerical modification (adjustment) for the target (parameter / initial value / boundary value) to be experimented with in the current single	string target_name (input) float target_val_def  float get_py	name of the target in <model>.edf  nominal / default (non-adjusted) target value.  If target_name is not defined in <model>.edf then target_val_adj is set to target&gt;_val_def  adjusted target value target_val_adj</model></model>
simenv_ get_run_py (	gets run number of current run as a character string	(function value) string get_run_py (function value)	current run number as string of the length 6 with leading zeros.  If an error occurred then run_char = ''
simenv_ ini_py (	initialize model coupling interface  Apply always before the first call of the other SimEnv functions in the model	string ini_py (function value)	return code of the spawn function for a SimEnv executable
simenv_ put_py ( var_name, field )	outputs model results to native SimEnv output file(s)	string var_name (input) declaration of field() according to <model>.mdf (input) put_py (function value)</model>	name of the variable in <model>.mdf to be output  data of variable var_name to be stored as simulation results. Maximum length of field is limited to 12.000 characters.  unused</model>
simenv_ slice_py ( var_name, idim, ifrom, ito )	Currently not available for Python models		

Tab. 5.4Model coupler functions for Python models

 Python coupling functions are declared in the file \$SE\_HOME/simenv.py. To use these functions in a Python model import it by

from simenv import \*

- and refer to it for example by simenv.get run py.
- For interfacing a Python model to SimEnv extend your operating system environment variable PYTHONPATH by \$SE\_HOME.
- Errors that occur during performance of one of the above functions are directly reported to <model>.mlog.

In Example 15.4 at page 126 the model world py.py is described in detail.

#### 5.4.1 Standard User-Defined Files for Python Models

#### <model>.ini

<model>.ini (see chapter 7.1 at page 47) is for Python models a mandatory script and has to have the same contents for all Python models:

```
$SE_HOME/py_model_ini
rc_py_model_ini = $?

# additional user-model specific commands can be implemented up from here
if test $rc_py_model_ini = 0
then
...
fi

exit $rc_py_model_ini
```

For an experiment restart (check chapter 7.2 at page 49) <model>.ini has to be performed again. To force this specify in <model>.cfg (check chapter 10.1at page 87) for the sub-keyword restart\_ini the value yes.

# 5.5 Model Interface at Shell Script Level

For models that do not allow to implement the model coupling interface at programming language level (e.g., because source code is not available) SimEnv supplies a coupling interface at shell script level: the shell script <model>.run (see chapter 7.1 at page 47) is used to wrap the model and optionally to have at disposal corresponding functionality of the SimEnv coupling functions of Tab. 5.3).

- For the model interface at the shell script level, i.e., within the shell script <model>.run the adjusted experiment targets for the current single run from the whole run ensemble can be made available within <model>.run to forward them by any means the modeller is responsible for to the model under investigation.
  - One common way to forward experiment targets to the model is to place current numerical target values as arguments to the model at model command line. Another way could be to read the targets from a special file in a special file format.
- Directly before performing simenv\_get\_sh make sure that the shell script variables target\_name and target\_def\_val have be specified. At the end of each simenv\_get\_sh these variables are set again to empty strings.
- After running . \$SE\_HOME/simenv\_get\_sh an experiment target <target\_name> from the experiment description file <model>.edf is available in <model>.run as a shell script variable <target\_name> and the adjusted value of the target is available as \$<target\_name>.
- After running the model model output has to be identified and potentially transformed within <model>.run
  for SimEnv output. To do this simply write your own simenv\_put\_sh as a transformation program that
  reads in all the native model output and outputs it to SimEnv by applying the coupling functions
  simenv \* \* from the SimEnv model coupler at language level.

Tab. 10.8 lists the built-in (pre-defined) shell script variables that are used in \$SE\_HOME/simenv\_\*\_sh and finally in <model>.run.

Command name	Command description	Arguments	Argument description
. \$SE_HOME/ simenv_end _sh	wrap up current single run  Apply always as the last command in <model>.run</model>		
target_name= '' target_def_val=	gets a numerical modification (adjustment)	script variable target_name (input)	name of the target in <model>.edf</model>
for simenv_get value	for the target (pa- rameter / initial value / boundary	script variable target_def_val (input)	nominal / default (non-adjusted) target value.  If target_name is not defined in <model>.edf then target_val_adj is set to target&gt;_val_def</model>
_sh	value) to be ex- perimented with in the current single run	script variable target_name (output)	shell script variable with the same name as the value of target_name. Script variable value is the adjusted target value target_val_adj.
. \$SE_HOME/ simenv get run	gets run number of current run as an	run_char (output)	shell script variable with the current run number with leading zeros
_sh	integer and a character script variable	run_int (output)	shell script variable (type integer) with the current run number
. \$SE_HOME/ simenv_ini _sh	initialize current single run  Apply always as the first command in <model>.run</model>	SE_RUN (output)	operating system environment variable SE_RUN is set to the current run number of the simulation experiment
. \$SE_HOME/ simenv_put _sh	Not available at shell script level		Write your own simenv_put_sh at the language level using the SimEnv coupling functions from Tab. 5.3 or Tab. 5.4
. \$SE_HOME/ simenv_slice _sh	Not available at shell script level		

Tab. 5.5 Model coupler functions at shell script level

In Example 15.5 at page 127 the model shell script world\_sh.run is described in detail.

```
. $SE HOME/simenv ini sh
# get adjusted value for the a target p def, defined in the edf-file
target name='p def'
target def val=2.
. $SE_HOME/simenv_get_sh
                                        is available
# now shell script variable p def
# value of shell script variable p def is according to edf-file
# get adjusted value for a target p_undef, not defined in edf-file
target name='p undef'
target def val=-999.
```

```
. $SE_HOME/simenv_get_sh
# now shell script variable p_undef is available
# value of shell script variable p_undef is -999.

# ...
. $SE_HOME/simenv_end_sh

Example file: world_sh.run
```

**Example 5.2** Addressing target names and values for the model interface at shell script level

#### 5.6 Model Interface for GAMS Models

SimEnv allows to couple GAMS models to the experiment shell. A GAMS model for SimEnv can consist of a GAMS main model and GAMS submodels.

Therefore, two additional include-statements have to be inserted into these GAMS model source code files where experiment targets are to be adjusted or model variables are to be output. GAMS model source code files to be interfaced to SimEnv are one GAMS main model and a number of GAMS sub-model that are called directly from the main model. All these GAMS model source code files have to be located in the current working directory. Additional GAMS sub-programs (included files) are not affected bei SimEnv, but you should keep in mind that the GAMS code within SimEnv will be executed in a subdirectory of the current working directory (see below) and so the include statements have to be changed, if the files are adressed in a relative manner (see below).

#### The include files are

- \$include <GAMS\_model>\_simenv\_get.inc
   \$include <GAMS\_model>\_simenv\_put.inc
   where <GAMS\_model> is the name of a GAMS model file without extension .gms under consideration.
   The include statement \$include <GAMS\_model>\_simenv\_put.inc has to be placed in the GAMS model file at such a position where all the variables from the model output description file can be output by GAMS put-statements.
- During experiment preparation the file <GAMS\_model>\_simenv\_put.inc and during experiment performance files <GAMS\_model>\_simenv\_get.inc are generated automatically to forward GAMS model output to SimEnv data structures and to adjust investigated experiment targets, respectively. These include files correspond with the simenv\_put and simenv\_get functions at the language level (see chapter 5.3).
- For interfacing a GAMS model to SimEnv extend your operating system environment variable PYTHONPATH by \$SE\_HOME.
- In the course of experiment preparation the GAMS model and all sub-models that are specified in <model>.gdf (see below) are transformed automatically. Each GAMS model single run from the run ensemble is performed in a separate sub-directory of the current working directory. Transformed GAMS models and sub-models are copied to this sub-directory and are performed from there. Keep this in mind if you specify in any GAMS model include statements with relative paths.

In Example 15.6 at page 129 the model gams\_model.gms is described in detail.

Additionally, the following settings are valid:

- An ASCII GAMS description file <model>.gdf (see below) has to be supplied to specify the GAMS submodels and assigned targets and model variables in detail.
- Maximum dimensionality of any model output variable declared in <model>.mdf is 4 for GAMS models.

Note the following information:

To output the GAMS model status to SimEnv a

PARAMETER modstat

has to be declared and the statement

modstat = <model\_name>.modelstat

has to be incorporated in the GAMS model above the \$include <GAMS\_model>\_simenv\_put.inc line. The variable modstat has to be stated in the model output description file <model>.mdf and the GAMS description file <model>.gdf.

Relevant information is output to the model log file <model>.mlog.

#### 5.6.1 Standard User-Defined Files for GAMS Models

#### <model>.ini

<model>.ini (see chapter 7.1 at page 47) is for GAMS models a mandatory script and has to have the contents for all GAMS models:

```
$SE_HOME/gams_model_ini
rc_gams_model_ini = $?

# additional user-model specific commands can be implemented up from here
if test $rc_gams_model_ini = 0
then
...
fi

exit $rc_gams_model_ini
```

For an experiment restart (check chapter 7.2 at page 49) <model>.ini has to be performed again. To force this specify in <model>.cfg (check chapter 10.1 at page 87) for the sub-keyword restart ini the value yes.

#### <model>.run

<model>.run (see chapter 7.1 at page 47) has for each GAMS model the same contents:

```
. $SE_HOME/simenv_ini_sh
$SE_HOME/gams_model_run
. $SE_HOME/simenv_end_sh
```

#### <model>.end

<model>.end (see chapter 7.1 at page 47) is for GAMS models a mandatory script and has to have the contents for all GAMS models:

```
$SE_HOME/gams_model_end
# additional user-model specific commands can follow
```

Python script language is used to prepare, run and to end a GAMS model.

#### <model>.edf

Corresponding experiment targets in the experiment description file <model>.edf (see chapter 6.1 at page 35) and in the GAMS model source code must have same names. In the GAMS model code the targets specified in the experiment description file have to be of type PARAMETER and have be defined before the include statement \$include simenv\_get.inc.



#### <model>.mdf

Corresponding variables in the model output description file and in the GAMS model source code must have same names. The variable type has to be always float in the model output description file. In GAMS model code the model variables declared in the model output description file can be of the numeric types VARIABLES or PARAMETER. Currently, maximal dimensionality of GAMS model output is restricted to 4.

With respect to	Example 15.6	6 the model output	description file could look like
coordinate coordinate coordinate	T	descr unit values	<pre>canning plants plant number equidist_end 1(1)2</pre>
coordinate coordinate coordinate	market market market	descr unit values	<pre>canning markets market number equidist_end 1(1)3</pre>
variable variable variable variable variable	a a a a	descr unit type coords var extents	plant capacity cases float plant 1:2
variable variable variable variable variable	x x x x	descr unit type coords var_extents	shipment quantities cases float plant , market 1:2 , 1:3
variable variable variable	Z Z Z	descr unit type	total transportation costs 10^3 US\$ float
variable variable	modstat modstat	descr type	model status float  Example file: gams_model.mdf

**Example 5.3** Model output description file for a GAMS model

#### 5.6.2 GAMS Description File <model>.gdf

The ASCII GAMS description file <model>.gdf is intented to create a block of lines for each GAMS submodel with a simenv\_get.inc file and/or a simenv\_put.inc file. The block holds the specific characteristics of GAMS model input and output needed by SimEnv to generate GAMS put-statements. All model variables from the model output description file and all targets from the target description file have to be used in this file again.

<model>.gdf is an ASCII file that follows the coding rules in chapter 11 at page 99 with the keywords, names, sub-keywords, and info as in Tab. 5.1.

<keyword></keyword>	<name></name>	<sub- keyword&gt;</sub- 	Line type	Max. line nmb.	<info></info>	Explanation
gdf	<nil></nil>	descr keep_runs	0	any 1	<string> <value_list></value_list></string>	GAMS coupling description value list of run numbers where single GAMS model runs are to be stored by keeping their corresponding sub-directories For syntax see Tab. 11.4
		time_limit	0	1	<pre><positive_ integer=""></positive_></pre>	CPU limit in seconds for each GAMS model single run
		options	0	1	<string></string>	string of options, GAMS is started with from command line
model	model_name	descr	0	1	<string></string>	(sub-)model description
	(without	type	m	1	[ main   sub ]	identifies GAMS main or submodel
	extension .gms)	get	m	exactly number of targets	<target_name></target_name>	get resulting adjustment for <target_name> to this model</target_name>
		put	m	exactly number of model vari- ables	<pre>(<var_name> {.<suffix_set>} {(<index_set>)}) {<format>}</format></index_set></suffix_set></var_name></pre>	put values of SimEnv model output variable <var_name> from this model to SimEnv output.  GAMS variable <var_name> has the specified suffix and index sets and is output according to <format></format></var_name></var_name>

 Tab. 5.6
 Elements of a GAMS description file <model>.gdf

To Tab. 5.6 the following additional rules and explanations apply:

- For the description of line type check Tab. 11.3 at page 100.
- Each target and each model variable as declared in <model>.edf and <model>.mdf respectively has to be used in the <info>-field of <model>.gdf exactly one time.
- To each GAMS model <model\_name> an arbitrary number of targets and model variables can be assigned to by the corresponding get and/or put sub-keyword.
  - To each sub-model (type = sub) at least one get or one put sub-keyword must be assigned to. The main model (type = main) can be configured without any get and put keywords. This is useful when the main model simply calls sub-models.
- Each model <model\_name> in <model>.gdf with at least one get sub-keyword has to have an \$include <model\_name>\_simenv\_get.inc statement in the corresponding GAMS model file <model\_name>.gms
- Each model <model\_name> in <model>.gdf with at least one put sub-keyword has to have an \$include <model name> simenv put.inc statement in the corresponding GAMS model file <model name>.gms
- There has to be exactly one main GAMS model, identified by the <subkeyword> type. All other models have to be of type = sub.
- The <info>-field for the <sub-keyword> put is adapted to GAMS syntax to output GAMS model variables.
   Afterwards this output is used to generate the appropriate SimEnv output.
   <index\_set> is mandatory for variables with a dimensionality > 0. Otherwise, specification of <index\_set> is forbidden. Indices as used in the GAMS model are separated from each other by comma.
- The sub-keyword time\_limit enables limitation of each GAMS model single run in the run ensemble to a
  maximum CPU-time consumption. If this threshold is reached the single run is aborted and the following
  single run started. In general, SimEnv nodata values will be assigned to the results of the aborted single
  runs. The time limit keyword can be necessary since each GAMS model single run itself is an optimiza-



tion procedure which could result in an unfeasible CPU time consumption. If the sub-keyword is not used in the gdf-file CPU-time cosumption per single run is unlimited.

With res	pect to Example 15	.6 the GAMS descri	ption file could look like
gdf gdf gdf gdf		descr descr descr keep_runs	GAMS model output description for the examples in the SimEnv User's Guide list 0,1
model model model model model model model	gams_model gams_model gams_model gams_model gams_model gams_model gams_model	descr type get get put put put put	<pre>this is the only GAMS model to use main dem_ny dem_ch x.l(i,j):10:5 a(i):10:5 z.l modstat</pre>
			Example file: gams_model.gdf

**Example 5.4** GAMS description file <model>.gdf

gams model

gams\_model

model

model

If the model gams\_model from the above Example 5.5 would be coupled with two additional GAMS sub-models sub\_m1 and sub\_m2 where both sub-models interact with SimEnv the GAMS description file could look like

main

modstat

(without taking into consideration plausibility with respect to model contents)

type

put

model model model	sub_m1 sub_m1 sub_m1 sub_m1	type get put put	sub dem_ny x.l(i,j):10:5 a(i):10:5
model model model	sub_m2 sub_m2 sub_m2	type get put	sub dem_ch z.1
or			
model	gams_model	type	main
model model model	sub_m1 sub_m1 sub_m1 sub_m1	type get put put	<pre>sub dem_ny x.l(i,j):10:5 a(i):10:5</pre>
model model model model	sub_m2 sub_m2 sub_m2 sub_m2	type get put put	sub dem_ch z.1 modstat

**Example 5.5** GAMS description file for coupled GAMS models



#### 5.6.3 Files Created during GAMS Model Performance

During experiment performance of a GAMS model each single run from the experiment is performed individually in a directory run<run\_char> of the current working directory. Each directory is generated automatically before performing the corresponding single run and removed after perfomance of this single run. With the sub-keyword keep\_runs the user can force to keep sub-directories for later check of the transformed model code.

Additionally to the files listed in Tab. 10.5, during the performance of a GAMS model the files <gams\_model>\_[ pre | main | post ].inc are created temporarily in the current working directory by <model>.ini and are deleted after the whole experiment where <gams\_model> is a placeholder for the model of type main and all models of type sub in the gdf-file.

#### 5.7 Distributed Models

SimEnv supports performance of distributed models. Distributed models may consist from a web or a chain of stand-alone sub-models, i.e., the model dynamics are computed by performing a set of stand-alone sub-models that normally exchange information.

Each of these sub-models can use SimEnv functionality, i.e., simenv\_get\_\*, simenv\_get\_run\_\*, simenv\_put\_\*, or simenv\_slice\_\*. In each sub-model simenv\_ini\_\* and simenv\_end\_\* have to be incorporated in. Sub-models can be implemented in different programming languages. Additionally, the corresponding SimEnv model coupling functionality at shell script level (simenv\_\*\_sh modules) can be applied. As usual, the overall model is wrapped into a shell script <model>.run (see chapter 7).

The model description file <model>.mdf collects all the state variables from all sub-models and the experiment description file <model>.edf collects all the targets for all sub-models.

Sub-models can reside on different machines. The only prerequisite is that the current working directory and the model output directory can be mapped to each of these machines.

# 5.8 Running Interfaced Models Outside SimEnv

To run a model coupled to SimEnv outside the simulation environment in its native mode as before code transformation the following simple rules have to be applied to the model:

For Fortran and C/C++ models:

Link the model with the object library \$SE\_HOME/libsimenvdummy.a instead of \$SE\_HOME/libsimenv.a. SimEnv function values (return codes) from this library are zero, function simenv\_get\_\* forwards target\_val\_def to target\_val\_adj, simenv\_get\_run\_\* returns integer run 0 and character run string ''(six blanks).

• For Python models:

Replace in the model source code

from simenv import \*

рy

from simenvdummy import \*

Function values (return codes) from simenvdummy.py are zero, dummy function simenv\_get\_py forwards target\_val\_def to target\_val\_adj, simenv\_get\_run\_py returns run 000000.

For GAMS models:

Handle in the model source code

\$include <model>\_simenv\_get.inc

and

\$include <model>\_simenv\_put.inc

as comment lines.



## 6 Experiment Preparation

Experiment preparation is the first step in experiment performance of a model coupled to the environment. In an experiment description file <model>.edf all information to the selected experiment type and its numerical equipment is gathered in a structured way.

## 6.1 Experiment Description File <model>.edf

<model>.edf is an ASCII file that follows the coding rules in chapter 11 at page 99 with the keywords, names, sub-keywords, and info as in Tab. 6.1.

<keyword></keyword>	<name></name>	<sub- keyword&gt;</sub- 	Line type	Max. line nmb.	<info></info>	Explanation
edf	<nil></nil>	descr	0	any	<string></string>	experiment description
		type	m	1	[behaviour   monte carlo   sensitivity   optimization]	experiment type
target	target_	descr	0	1	<string></string>	target description
	name	unit	0	1	<string></string>	target unit
		type	m	1	see Tab. 6.2	adjustment type
		default	m	1	<value></value>	target default value <target_def_val></target_def_val>
		adjusts	c3	1	<pre><experiment- specific=""></experiment-></pre>	experiment-specific information
specific	<nil></nil>	<experiment- specific&gt;</experiment- 	m	<ex- peri- ment- spe- cific&gt;</ex- 	<experiment- specific&gt;</experiment- 	experiment-specific information

Tab. 6.1 Elements of an experiment description file <model>.edf

To Tab. 6.1 the following additional rules and explanations apply:

- For the description of **line type** check Tab. 11.3 at page 100.
- Target names must differ from model variables and coordinate names in the model output description file (see chapter 5.1) and from built-in and user-defined operator names for model output post-processing (see chapter 8.6.2).
- A target name is the symbolic parameter / driver / initial value / boundary value name, corresponding to targets of the investigated model. Correspondence is achieved by applying the SimEnv model coupling function simenv\_get\_\* in the model.
- The default value as specified in <model>.edf and not the default value from the model code is used to derive the adjusted value.
- For adjustment type multiply default values = 0. are forbidden.
- All experiment-specific information is explained in the appropriate chapters.
- · Specify at least one experiment target.
- When preparing an experiment an experiment input file <model>.edf\_adj is generated with the values to
  be finally used for the resulting adjustments. These values are applied to the default values of the targets according to the specified adjustment type (see Tab. 6.2 below) before finally influencing the dynamics of the model. The sequence of elements (columns) of each record of <model>.edf adj corre-

sponds with the sequence of targets in the target name space (see chapter 11 at page 99), the sequence of records corresponds with the sequence of single model runs of the experiment. For each experiment a single model run with run number 0 is generated automatically as the nominal run of the model without adjustments. This run does not have an assigned record in <model>.edf\_adj.

Adjustment type	Meaning
set	value setting: Use the adjustment to the target default value within the SimEnv function simenv_get_* as the final adjusted value. Not available for local sensitivity analysis
add	addition: Add the declared adjustment to the target default value within the SimEnv function simenv get * to get the final adjusted value to use.
multiply	multiplication:  Multiply the declared adjustment with the target default value within the SimEnv function simenv_get_* to get the final adjustment to use.  Differing implementation for local sensitivity analysis (check chapter 6.4.1).

 Tab. 6.2
 Adjustment types in experiment preparation

## 6.2 Behavioural Analysis

The experiment-specific information for experiment description files in Tab. 6.1 at page 35 is defined for behavioural analysis as follows:

<keyword></keyword>	<name></name>	<sub- keyword&gt;</sub- 	Line type	Max. line nmb.	<info></info>	Explanation
target	target_ name	adjusts	а	1	<value_list></value_list>	value list of target value adjustments <adj_val> to apply. For syntax see Tab. 11.4</adj_val>
specific	<nil></nil>	comb	m	1 or any	[ default   <combination>   file {<path>/} <file_name> ]</file_name></path></combination>	information how to scan the spanned target space

Tab. 6.3 Experiment-specific elements of an edf-file for behavioural analysis

To Tab. 6.3 the following additional rules and explanations apply:

- For the description of **line type** check Tab. 11.3 at page 100.
- For <sub-keyword> = comb the following rule yields:
  - <info> = [ default | <combination> ] for available <sub-keyword> = adjusts <info> = [ file {<path>/}<file name> ] for unavailable <sub-keyword> = adjusts
- Values of a value list have to be unique for available <sub-keyword> = adjusts and each target
   Assigned values from file {<path>/}<file\_name> can be multiple defined for each target.
- <path> must not contain environment variables from operating system level.



#### 6.2.1 Adjustments

Adjustment type	Set	Add	Multiply
adjusted target value =	<adj_val></adj_val>	<target_def_val> + <adj_val></adj_val></target_def_val>	<target_def_val> * <adj_val></adj_val></target_def_val>

#### 6.2.2 The Combination

- The combination **<combination>** defines the way in which the space spanned by the experiment targets will be inspected by SimEnv: This is done by applying operators "\*" and "," to all stated experiment targets.
  - The operator "\*" combines adjustments of different targets and so their resulting values combinatorially ("for all mesh points in a grid").
    - Compare with experiment description file (a) from Example 6.1 below.
  - The operator "," combines adjustments of different targets and so their resulting values parallel ("on the diagonal").
    - For the operator "," the targets must have the same number of adjustments.
    - Compare with experiment description file (b) from Example 6.1 below.
  - The operator "," has a higher priority than the operator "\*". Parentheses are not allowed:
     For example, p1 \* p2 , p3 \* p4 always combines p2 and p3 in parallel and this combinatorially with p1
     and p4. A parallel combination of p1 \* p2 with p3 \* p4 by (p1 \* p2) , (p3 \* p4) is not possible.
     Compare with experiment description file (c) from Example 6.1 below.
  - In <combination> each target has to be used exactly once.
- By the default combination default all experiment targets are combined combinatorially.
  - comb default of the experiment description file (a) from Example 6.1 below is equivalent to comb p1 \* p2.
- Specification of file is only allowed for unused adjusts-sub-keywords all over the edf-file.
  - The adjustments are read from the adjustment data file {<path>/}<file name>.
  - All targets are assumed to be combined in parallel. Each record of the data file represents one simulation run. The sequence of the adjustments (sequence of columns) in each record corresponds with the sequence of the targets in the target name space (see chapter 11 at page 99).
  - Syntax rules for value lists at page 99 yield.
  - Identical adjustments for a target are allowed.
  - During model output post-processing restricted capabilities for the operator behav apply for this experiment layout.
  - Compare with experiment description file (d) from Example 6.1 below. Combination is implicitly as comb p1, p2. Experiment description files (b) and (d) in Example 6.1 below describe the same experiment.

### 6.2.3 Example

The first three experiment description files (a) to (c) represent an experiment description according to Fig. 4.3 (a) to (c) at page 12.

Results in values ...

					ixesuits iii values
(a)	edf		descr	Experiment description for	the examples
	edf		descr	in the SimEnv User's Guide	(Fig. 4.3 (a))
	edf		type	behaviour	
	target	p1	descr	parameter p1	
	target	p1	unit	without	
	target	p1	type	add	
	target	p1	default	1.	
	target	p1	adjusts	list 1, 3, 7, 8	2,4,8,9 for p1

```
p2
   target
                  descr
                             parameter p2
             p2
   target
                             without
                  unit.
                  type
            p2
   target
                             multiply
   target
             p2
                  default
   target
           p2
                  adjusts
                              list 1, 2, 3, 4
                                                         ... 2,4,6,8 for p2
   specific
                  comb
                              default
(b) edf
                  descr
                             Fig. 4.3 (b)
                 type
   edf
                            behaviour
   target p1 type
target p1 default
                            multiply
                             1.
   target p1 adjusts
                            list 1, 3, 7, 8
                                                         ... 1,3,7,8 for p1
           p2 type
                            multiply
   target
           p2 default
   target
   target specific
            p2 adjusts equidist_end 1(0.5)2.5 ... 2,3,4,5 for p2
                  comb
                              p1,p2
                           Fig. 4.3 (c)
(c) edf
                   descr
                            behaviour
   edf
                  type
   target p1 type
target p1 default
                              set
                              1.
   target p1 adjusts
                            list 1, 3, 7, 8
                                                         ... 1,3,7,8 for p1
   target p2 type
                              set
   target
           p2 default
                             2..
   target
           p2 adjusts
                            equidist end 1(1)4
                                                         ... 1,2,3,4 for p2
           p3 type
                            multiply
   target
   target p3 default target p3 adjusts
            p3 adjusts
                             list 1.1, 1.5, 2.4
                                                         ... 3.3,4.5,7.2 for p3
   specific
                  comb
                             p2,p1*p3
(d) edf
                  type
                             behaviour
                                               file world.dat d:
   target pl
                             multiply
                  type
                                                 1 0
   target p1
target p2
target p2
                  default
                             1.
                                                  3
                                                       1
                                                  7
                                                        2
                  type
                              add
           p2 default
                              2.
                                                  8
                                                       3
                           file world.dat_d
   specific
                                                         ... (1,2),(3,3),(7,4),(8,5)
                  comb
                                                         ... for (p1,p2)
                                             Example files: world.edf_a to world.edf_d
```

**Example 6.1** Experiment description file <model>.edf for behavioural analysis

#### **6.2.4 Experiment Performance**

- Firstly, a model run 000000 with the default values of the experiment targets is performed.
- According to the keyword comb the appropriate runs are generated.
- The sequence of the runs corresponds with the sequence of the adjustments in the ASCII file <model>.edf adj (check chapter 6.1 at page 35 for more information).



## 6.3 Monte Carlo Analysis

The experiment-specific information for experiment description files in Tab. 6.1 at page 35 is defined for Monte Carlo analysis as follows:

<keyword></keyword>	<name></name>	<sub- keyword&gt;</sub- 	Line type	Max. line nmb.	<info></info>	Explanation
target	target_ name	adjusts	m	1	[ <distribution>   file {<path>/} <file_name> ]</file_name></path></distribution>	distribution and distribution parameters to be applied for the target or import of an external sample <distr_val> from <file_name></file_name></distr_val>
		sample	c4	1	[ random   latin hypercube ]	sampling strategy: random or latin hypercube sampling LHS
specific	<nil></nil>	runs	m	1	<nr_of_runs></nr_of_runs>	number of runs > 10 to be performed for the experiment

Tab. 6.4 Experiment-specific elements of an edf-file for Monte Carlo analysis

To Tab. 6.4 the following additional rules and explanations apply:

- For the description of line type check Tab. 11.3 at page 100.
- <distribution> = <distr\_shortcut> ( <distr\_param\_1> { , <distr\_param\_2> } ) (check Tab. 6.5)
- For implicitly specified distributions according to Tab. 6.5 adjustments are applied to the specified distribution parameters of the distributions. Afterwards, a sample <distr\_val> is generated from the distribution with the adjusted distribution parameters. Adjustment types add and multiply are not applied to the distribution parameter <distr\_param> = standard deviation. Instead, the specified standard deviation from the experiment description file is used (adjustment type set is applied).
- For explicitly specified samples of any distribution by the ASCII file <file\_name> adjustments are applied directly to the sample values <distr\_val> from the file. For syntax rules for files check chapter 11. Each record of the ASCII file can hold only one sample value. Sample size has to be identical to <nr\_of\_runs> from the specific-keyword.
- In random sampling, there is no assurance that sampling points will cover all regions of the selected distribution. With Latin hypercube sampling LHS this shortcoming is reduced: The sampling range of the target is divided into <nr\_of\_runs> intervals of equal probability according to the selected distribution and from each interval exactly one sampling point is drawn. For more information on LHS check Fig. 6.1 below and see Imam & Helton (1998) and Helton & Davis (2000).
- The number of runs <nr\_of\_runs> must be greater than 10.

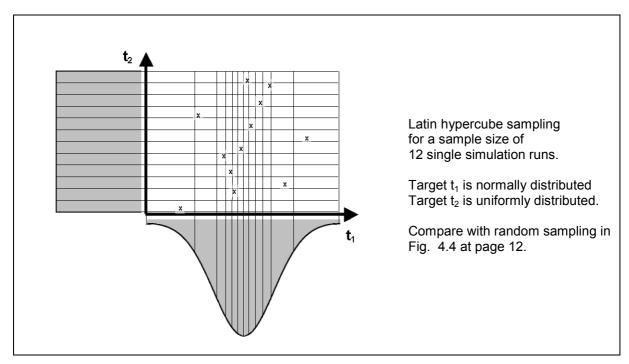


Fig. 6.1 Monte Carlo analysis: Latin hypercube sampling

## 6.3.1 Adjustments

Adjustment type	Set	Add	Multiply	
for distribution:		<target_def_val> +</target_def_val>	<target_def_val> *</target_def_val>	
adjusted distr_param =	<distr param=""></distr>	<distr_param></distr_param>	<distr_param></distr_param>	
	alou_param	not for standard deviation		
		instead, adjustment type "set" is applied		
for file:	<distr val=""></distr>	<target_def_val> +</target_def_val>	<target_def_val> *</target_def_val>	
adjusted target_value =	<distr_val></distr_val>	<distr val=""></distr>	<distr val=""></distr>	

### 6.3.2 Distribution Functions and their Parameters

Distribution function	<distr_ shortcut&gt;</distr_ 	<distr_param_1></distr_param_1>	<distr_param_2></distr_param_2>	Restriction
uniform	U	lower boundary	upper boundary	lower boundary < upper boundary
normal	N	mean value	variance	variance > 0
lognormal	L	mean value of a normally distributed target	variance of a nor- mally distributed target	variance > 0
exponential	Е	mean value		mean value > 0

Tab. 6.5Probability density functions and their parameters

For more information on the distribution functions see chapter 4.3 and Tab. 4.2.

### 6.3.3 Example

(e)	edf edf edf		descr descr type	Experiment descrip in the SimEnv User Monte Carlo	otion for the examples
	target target target target target target	p2 p2 p2 p2 p2 p2	descr unit type default sample adjusts	parameter p1 without multiply 2. latin hypercube distr U(0.5,1.5)	p2 is a realization of a uniform distrib. between 0.5*2 and 1.5*2
	target target target target	-	type default sample adjusts	add 1. random distr N(0,0.4)	p1 is a realization of a normal distribution with mean = 1+0 and variance = 0.4
	target target target	p3 p3 p3	type default adjusts	file world.dat_e	realization of p3 is read from file world.dat_e and afterwards 3 is added
	specific		runs	250	Example file: world.edf_e

**Example 6.2** Experiment description file <model>.edf for Monte Carlo analysis

#### 6.3.4 Experiment Performance

- Firstly, a model run 000000 with the default values of the experiment targets is performed which represents the deterministic case.
- The sequence of the runs corresponds with the sequence of the adjustments in the ASCII file <model>.edf\_adj. <model>.edf\_adj is generated from random numbers of the appropriate distributions U(0,1), N(0,1), L(0,1), and/or E(1). For more information on <model>.edf\_adj check chapter 6.1 at page 35.
- If the resulting distribution parameters do not fulfil the restrictions in Tab. 6.5 the following adaptations are applied and corresponding warnings are output to <model>.mlog.

Distribution	Condition	Adaptation
U	lower boundary > upper boundary	boundaries are interchanged
U	lower boundary = upper boundary	lower boundary := lower boundary - 0.5 upper boundary := upper boundary + 0.5
E	mean < 0	mean := -mean
E	mean = 0	mean := abs(model default value) for model default value ≠ 0 1 else

 Tab. 6.6
 Probability density functions: Distribution parameters - conditions and adaptation

## 6.4 Local Sensitivity Analysis

The experiment-specific information for experiment description files in Tab. 6.1 at page 35 is defined for local sensitivity analysis as follows:

<keyword></keyword>	<name></name>	<sub- keyword&gt;</sub- 	Line type	Max. line nmb.	<info></info>	Explanation
target	target_ name	adjusts	f	0		sub-keyword adjusts is forbid- den for this experiment type
specific	<nil></nil>	incrs	m	1	<value_list></value_list>	increments <incr_val> &gt; 0. For all targets defined by a value list. Value list has to be strictly monotonic increasing. For syntax see Tab. 11.4</incr_val>

Tab. 6.7 Experiment-specific elements of an edf-file for local sensitivity analysis

To Tab. 6.7 the following additional rules and explanations apply:

- For the description of **line type** check Tab. 11.3 at page 100.
- Values from the value list must be positive and unique.
- **Note** that computation of adjusted values for adjustment type multiply in local sensitivity analysis differs from all other experiment types (see chapter 6.4.1 below.

### 6.4.1 Adjustments

Adjustment type	Set	Add	Multiply
adjusted target value =	undefined for this	<target_def_val> ±</target_def_val>	<target_def_val> *</target_def_val>
	experiment type	<incr val=""></incr>	(1 ± <incr val="">)</incr>

As an example, the absolute sensitivity function (see Tab. 4.3 at page 15) is then as follows:

for adjustment type Add sens\_abs(<target\_def\_val>, ±<incr\_val>) =

z(<target\_def\_val> ± <incr\_val>) - z(<target\_def\_val)

± <incr\_val>



#### 6.4.2 Example

```
(f) edf
                            Experiment description for the examples
                  descr
   edf
                  descr
                            in the SimEnv User's Guide
   edf
                            sensitivity
                  type
          p1
   target
                  descr
                            parameter p1
          p1
                  unit
                            without
   target
          p1
   target
                            add
                  type
   target
           p1
                  default
   target p2
                type
                            multiply
   target
           p2
                 default
                            2.
           рЗ
   target
                 type
                            multiply
           р3
   target
                  default
                            3.
                            list 0.001, 0.01, 0.05, 0.1
   specific
                  incrs
                                                        Example file: world.edf
```

**Example 6.3** Experiment description file <model>.edf for local sensitivity analysis

### 6.4.3 Experiment Performance

- Each experiment target will be adjusted by the same increments as those stated in the incrs info-field
- For finite sensitivity functions several runs have to be performed:
  - A nominal run with the default values of the experiment targets (run number 000000)
  - Per target and per increment two runs with the default values of all targets except that one under consideration, where the adjustment is applied according to the above adjustment rules
  - Accordingly, the number of resulting runs is 2 \* number\_of\_targets \* number\_of\_increments + 1
- Results of each model run are stored and sensitivity functions are applied during model output postprocessing.

The following sensitivity functions can be performed:

Linear, squared, absolute, relative as well as a symmetry test.

• The sequence of the simulation runs is determined in the following manner:

nominal run

loop over increment sequence

loop over experiment targets

end loop

end loop

loop over negative increment sequence

loop over experiment targets

end loop

end loop



# 6.5 Optimization

The experiment-specific information for experiment description files in Tab. 6.1 at page 35 is defined for local sensitivity analysis as follows:

<keyword></keyword>	<name></name>	<sub- keyword&gt;</sub- 	Line type	Max. line nmb.	<info></info>	Explanation
target	target_ name	adjusts	m	1	<pre><lower_bound>:   <upper_bound></upper_bound></lower_bound></pre>	real valued lower bound and upper bound to define the target range where the cost function is to be minimized on. <lower_bound> &lt; &lt; target_def_val&gt; &lt; &lt; upper_bound&gt;</lower_bound>
specific	<nil></nil>	cost_fct	m	1	<result_ expression&gt;</result_ 	cost function to minimize.  0-dimensional result expression formed according to the rules of the SimEnv postprocessor. Do not apply multirun operators.

 Tab.
 6.8
 Experiment-specific elements of an edf-file for an optimization experiment

To Tab. 6.8 the following additional rules and explanations apply:

• For the description of line type check Tab. 11.3 at page 100.

### 6.5.1 Adjustments

Adjustment type	Set	Add	Multiply
adjusted target value =	<adj_val></adj_val>	<target_def_val> + <adj_val></adj_val></target_def_val>	<target_def_val> * <adj_val></adj_val></target_def_val>

### 6.5.2 Example

(g)	edf edf edf		descr descr type	Experiment description for the examples in the SimEnv User's Guide optimization	3
	target target target target target	p1 p1 p1 p1 p1	descr unit type default adjusts	<pre>parameter p1 without set 112:12</pre>	, 12>
	target target target target target target	p2 p2 p2 p3 p3	type default adjusts type default adjusts	set 2. 1:10 set 3. -12:12	



```
target
           р4
                              set
                  type
target
           р4
                  default
                              4.
                  adjusts
                              1:10
target
           р4
                              -sum(bios) maximize bios over land masses
specific
                  cost fct
                                                               Example file: world.edf_g
```

**Example 6.4** Experiment description file <model>.edf for an optimization experiment

#### 6.5.3 Experiment Performance

- This is the only experiment type where the adjustments for the targets of the single runs are not determined before the experiment but in the course of the experiment by the optimization algorithm. Consequently, the file <model>.edf\_adj is not created during experiment prepration but is written during experiment performance.
- In parallel to the file model>.edf\_adj an ASCII file <model>.edf\_cf is written during experiment performance with the value of the cost function for each of the single runs.
- The status of an optimization experiment can be acquired by the SimEnv service simenv.sts. For more information check Tab. 10.3.
- The optimization algorithm itself is controlled by additional technical parameters and options that are normally fixed by SimEnv. To modify them copy from \$SE\_HOME the ASCII file **simenv.oopt** to the directory you want to start an optimization experiment and edit this file. During the experiment the edited file is used instead of the file with the default constellation in \$SE\_HOME. The description of the options and parameters can be found in Ingber (2004).
- Optimization experiments can not restarted by the SimEnv service simenv.rst.
- The values for the sub-keywords begin\_run and end\_run in the configuration file <model>.cfg are ignored for an optimization experiment. The experiment always starts with run number 0 and ends if one of the criterions in the file simenv.oopt (see above) is fullfilled.
- The optimal targets, the corresponding value of the cost function and the number of the corresponding single run as the results of the optimization experiment are documented at the end of the model log-file <model>.mlog.
- A protocol from the optimization procedure is delivered by SimEnv in the ASCII file <model>.olog.
- The initial seed for the optimization technique is fixed. That's why the algorithm results for the same optimization problem always in the same sampling sequence in the target space.

# 7 Experiment Performance

After experiment preparation experiment performance is the second step in running a model coupled to SimEnv. Each multi-run experiment can be performed sequentially or in parallel. Besides a new-start of an experiment a restart after an experiment interrupt or only for an experiment slice can be handled by SimEnv.

## 7.1 Experiment Start

- Currently an experiment can be performed sequentially on the login-machine and in parallel and/or sequential mode in a job class controlled by the LoadLeveler. In parallel mode the single runs of the run ensemble are distributed to all allocated nodes with their assigned processors. One communication processor is responsible for experiment management.
- The user can define an experiment preparation shell script <model>.ini that is performed additionally after standard experiment preparation when starting a new experiment. For experiment restart <model>.ini is performed only on request (see chapter 7.2 below).
   In <model>.ini additional settings / checks can be performed. For return codes unless zero from
  - <model>.ini the experiment will not be started. ). Make sure that <model>.ini has execute permission by chmod u+x. For Python and GAMS models <model>.ini is a mandatory script with pre-defined contents. Check chapters 5.4.1 and 5.6.1 for more information.
- The model to be applied within the SimEnv experiment has to be wrapped in the shell script <model>.run. <model>.run is performed for each single run within the run ensemble.
  - Make sure that
    - . \$SE HOME/simenv ini sh is the first command and
    - . \$SE\_HOME/simenv\_end\_sh is the last command

in <model>.run (see Tab. 5.5 at page 28 and Example 7.2 below).

- Ensure by chmod u+x that <model>.run has execute permission.
- To cancel the whole experiment after the performance of the current single run due to any condition
  of this run make sure a file <model>.\$run\_char.err exists as an indicator to stop. You can create
  this file in the model or in <model>.run. For the latter
  - Perform . \$SE\_HOME/simenv\_get\_run\_sh to get the current run number <run\_int> and <run\_char> (see Tab. 5.5 at page 28 and Example 7.2 below).
  - Touch the file <model>.\$run\_char.err.
- For GAMS models <model>.run has a pre-defined structure. Check chapter 5.6.1 for more information.
- The model variables to be output during experiment performance are declared in the model output description file <model>.mdf
- The type and the targets of the experiment to be performed are declared in the experiment description file <model>.edf
- Mapping between experiment targets and targets in the model source code is achieved by application of the generic SimEnv function simenv\_get\_\* in the model code or at shell script level.
- Output of model variables declared in <model>.mdf into SimEnv structures is achieved by the application of the generic SimEnv function simenv\_put (and simenv\_slice) in the model source code.
- Model output from run number <run> is stored in the file <model>.out<run\_char>.[ nc | ieee ] if the sum over all model output variables of a single run is less than the appropriate value specified in <model>cfg. Otherwise, model output from the complete experiment is stored in <model>.outall.[ nc | ieee ].
- For each experiment type a run number 0 with the default values of all experiment targets will be performed additionally to the runs declared in the experiment description file <model>.edf.
- During experiment performance a model log-file <model>.mlog is written where adjustments of experiment target values and possibly workarounds for wrong re-adjustments (only for experiment type Monte Carlo analysis, see Tab. 6.6) are stored. All model output to the terminal is re-directed within SimEnv to the model protocol file <model>.mlog.
- During experiment performance an experiment log-file <model>.elog is written with the minutes of the experiment.

- Do not start and/or submit another experiment from a working directory where an experiment is still running.
- After the experiment has been finished the model-specific output from the experiment can be wrapped up with the optional shell script <model>.end.
- After the experiment has been finished an e-mail is send on demand (check chapter 10.1) to the address as specified in <model>.cfg.
- For more information check Fig. 7.1.

```
For the shell script world_*.ini the following contents could be defined:

# coarse 0.5° x 0.5° land-sea mask from file land_sea_mask.05x05

# in the current directory

# to a 4° x 4° resoluted land-sea-mask in file land_sea_mask.coarsed

# in the current directory to use for all single runs

land_sea_mask 4 4

rc_land_sea_mask=$?

# exit from world_*.ini with return code != 0

# as an indicator not to start the experiment

exit $rc_land_sea_mask

Example files: world_[f|c|cpp|py|sh].ini
```

**Example 7.1** Shell script <model>.ini for user-model specific experiment preparation

```
For the shell script world f.run the following contents could be defined:
# always perform at begin:
. $SE HOME/simenv ini sh
# run the model:
world f
# assuming a model return code != 0 as an indicator to stop
# the whole experiment for any reason.
# Touch the file below in the current working directory $SE WD
# as an indicator to SimEnv for this.
if test $? -ne 0
      . $SE HOME/simenv get run sh
      touch $SE WD/world f.$run char.err
fi
# always perform at end:
. $SE HOME/simenv end sh
                                                           Example file: world_f.run
```

**Example 7.2** Shell script <model>.run to wrap the user model

For the shell script world f.end the following contents could be defined:

```
# remove the file of the coarsed land-sea mask
rm -f land sea mask.coarsed
```

Example file: world\_[f | c | cpp | py | sh ].end

Example 7.3 Shell script <model>.end for user-model specific experiment wrap-up

### 7.2 Experiment Restart

When an experiment was interrupted / has failed due to any reason or in the case of partial experiment performance (see chapter 7.3 below) it can be restarted several times:

- Simply restart the experiment by simenv.rst without changing any of the SimEnv files describing the experiment and/or the model. The only exception may be the information for the experiment-keyword in the general model configuration file <model>.cfg.
- simenv.rst has the same usage as simenv.run
- Restart can be launched on an other machine / in an other job class than that of the interrupted experiment.
- Dependent on the experiment log-file <model>.elog, written in the interrupted experiment a single model run from the complete run ensemble in the restart experiment will be
  - Performed if this run has neither a start nor a finish information in the elog-file
  - Not performed if this run has a start as well as a finish information in the elog file
  - Performed anew if the run has a start information but no finish information in the elog-file. For this case a model restart shell script <model>.rst can be provided by the user optionally to prepare restart of this single model run (e.g., by deleting non-SimEnv temporary or output files). Make sure that <model>.rst has execute permission by chmod u+x.

Make sure that . \$SE\_HOME/simenv\_ini\_sh is the first command in <model>.rst.

After running \$SE\_HOME/simenv\_get\_run\_sh the shell script variables run\_int and run\_char are available in <model>.rst (see above).

- Experiment restart works without standard SimEnv experiment preparation. Instead, experiment preparation files and other information from the interrupted experiment will be used.
- The optional experiment preparation shell script **<model>.ini** will be performed only on demand. This request is specified in the configuration file **<model>.cfg** with the sub-keyword restart\_ini.
- <model>.cfg will be checked anew for experiment restart. Avoid to change information in <model>.cfg for a restart. The only exception is the information related to the experiment-keyword.
- Minutes of the restarted experiment will be appended to the <model>.mlog and <model>. elog files, respectively from the interrupted experiment.
- Restart can be applied to an experiment several times successively.
- Experiment restart can be performed also as partial experiments, independently on the partial status of the original model
- Experiment re-start is not possible for the experiment type optimization.

For the model world\_sh (check Example 15.5 at page 127) the following contents could be defined for the restart script world\_sh.rst:

```
# always perform at begin
. $SE HOME/simenv ini sh
```

# get run number

. \$SE HOME/simenv\_get\_run\_sh

```
# remove all files from the temporary directory and the directory itself
if test -d run$run char
then
    cd run$run_char
    rm -f *
    cd ..
    rmdir run$run_char
fi
Example file: world_sh.rst
```

**Example 7.4** Shell script <model>.rst to prepare model performance during experiment restart

### 7.3 Experiment Partial Performance

- SimEnv enables to perform an experiment partially by performing only a run slice out of the whole run
  ensemble.
- Therefor assign appropriate run numbers to the corresponding experiment keywords in <model>.cfg.
- Make sure that begin run number and end run number represent run number from the experiment (including run number 0) and that begin run number ≤ end run number.
- A partial experiment performance is also possible for an experiment restart.
- For more information check Fig. 7.1.

## 7.4 Job Control by POE and LoadL

- For experiment performance controlled by the parallel operating environment POE and the LoadLeveler make sure that the environment variable SE\_HOME is set in your .profile-file correctly.
- On a login node to a parallel machine there is an additional SimEnv dialogue whether the experiment is to be submitted by POE and the LoadLeveler to a parallel or sequential job class of this parallel machine or is to be performed locally at the login node.
- Default job control files are supplied by SimEnv to ensure communication with POE and LoadLeveler. These job control files may be copied to the current working directory, can be modified and will then be used instead of the default job control files to start an experiment at a parallel or sequential job class. If necessary, copy \$SE\_HOME/simenv.jcf\_par and/or \$SE\_HOME/simenv.jcf\_seq to the current working directory SimEnv is started from, modify the file(s) according to the needs of the experiment you want to perform and / or the machine you want to use and start afterwards simenv.run (or simenv.rst). If available in the current working directory, these modified job control files are used instead of the original files in \$SE\_HOME.
  simenv.jcf\_seq submits a job to a sequential batch class, simenv.jcf\_par to a parallel batch class.
- Default job control files enable automatic restart of the experiment by the LoadLeveler after an interrupt of the job in a parallel or sequential job class caused by POE, the LoadLeveler or the operating system. The user does not need to restart the experiment manually after such an event.



# 7.5 Experiment-Related User Scripts and Files

Script / file	Explanation	Used for (*)	Exist status				
	Scripts (**)						
<model>.run</model>	model shell script to wrap the model executable . \$SE_HOME/simenv_ini_sh has to be the first command in <model>.run . \$SE_HOME/simenv_end_sh has to be the last command in <model>.run Model coupler functions at shell script level can be applied in <model>.run Pre-defined contents for GAMS models (check chapter 5.5)</model></model></model>	S R	mandatory				
<model>.rst</model>	model shell script to prepare single model run restart for such single runs that were started by not finished during the previous experiment start / restart . \$SE_HOME/simenv_ini_sh has to be the first command in <model>.rst . \$SE_HOME/simenv_get_run_sh can be applied in <model>.rst (check chapter 5.5)</model></model>	R	optional				
<model>.ini</model>	model shell script to prepare simulation experiment additionally to standard SimEnv preparation Experiment will be not performed if return code from this script is unequal zero. For experiment re-start <model>.ini will be performed only on request.</model>	S (R)	optional, for Python and GAMS models mandatory				
<model>.end</model>	model shell script to clean up simulation experiment from non-SimEnv files	S R	optional				
	Files						
<model>. <run_char>.err</run_char></model>	touch this file in the model, in <model>.run and/or <model>.rst as an indicator to stop the complete experiment after single run <run_char> has been finished</run_char></model></model>	Α	optional				
simenv.jcf_par	user-specific job control file to submit an experiment to a par- allel class by the LoadLeveler Copy from \$SE_HOME on demand	L	optional				
simenv.jcf_seq	user-specific job control file to submit an experiment to a sequential class by the LoadLeveler.  Copy from \$SE_HOME on demand	L	optional				
simenv.oopt	user-specific control and option file for experiment type Optimization Copy from \$SE_HOME on demand	0	optional				

Tab. 7.1Experiment-related user scripts and files

(\*): script applied for

R: Restart of an experiment by \$SE\_HOME/simenv.rst <model>

S: Start of an experiment by \$SE\_HOME/simenv.run <model> file applied for

A: All experiment perform. at the login machine or by LoadLeveler submission

L: LoadLeveler experiment submission

O: Optimization experiment performance

(\*\*): make sure the shell script has execute permission by chmod u+x

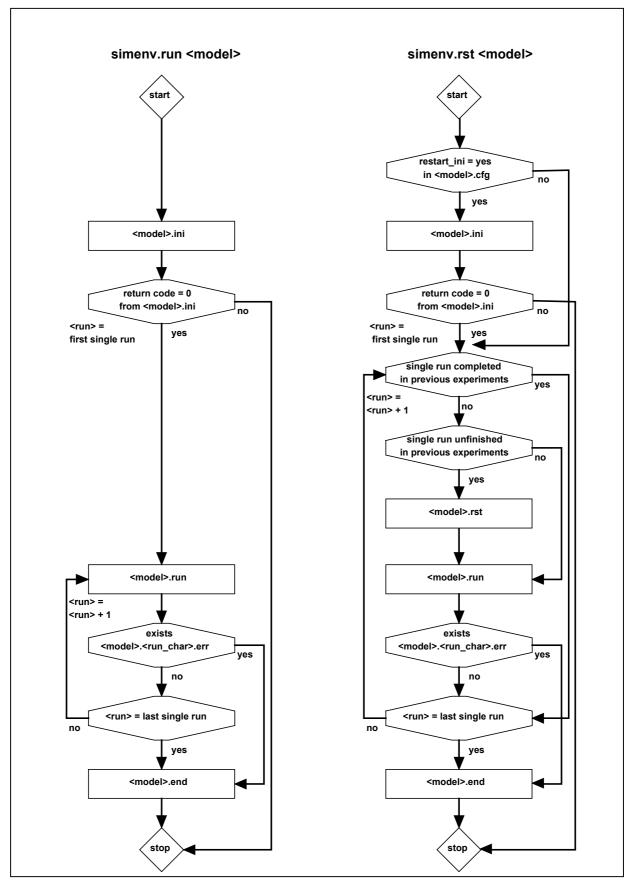


Fig. 7.1 Flowcharts for performing simenv.run and simenv.rst
First and last single run always refer to the corresponding settings in <model>.cfg

# 7.6 Saving Experiments

To save experiments for later use, e.g., by SimEnv post-processing, make sure to store from the experiment the following files:

•	<mdel>.out[ all   <r< th=""><th>un_char&gt; ].[ nc   ieee ]</th><th>from the model output directory</th></r<></mdel>	un_char> ].[ nc   ieee ]	from the model output directory
•	<model>.cfg</model>		from the current working directory
•	<model>.mdf</model>		from the current working directory
•	<model>.edf</model>		from the current working directory
•	<model>.elog</model>	(optional)	from the current working directory
•	<model>.mlog</model>	(optional)	from the current working directory
•	<model>.edf_adj</model>	(for optimization)	from the current working directory



# 8 Experiment Post-Processing

Goal of post-processing is to navigate within the model / experiment output space by deriving interactively output functions / data that are to be visualized in experiment evaluation afterwards. Therefor SimEnv supplies operators that can be applied to model output and reference data. There are built-in basic and advanced operators and built-in experiment-specific operators. The user can define its own private operators and easily couple them to the post-processor. Operator chains and recursions are possible. Macros can be defined as abbreviations for operator chains.

### 8.1 Operands

Operands in expressions can be

- Model output variables (see below)
- Experiment targets
- Constants in integer or float notation
- Character strings
- Operators
- Macros (see chapter 8.8)

To each operand (with the exception of character string operands) a

Dimensionality dim(operand) and

extents ext(operand,i) with i=1,...,dim(operand) and coordinates coord(operand,i) with i=1,...,dim(operand) are assigned to. The dimensionality is the number of dimensions, an extent is related to each dimension and represents the number of elements in that dimension. Extents are always greater than 1. To each dimension a coordinate is assigned to. Coordinates have a name and from all coordinate values the co-

specification for operands follows that for model output variables. For more information see chapter 5.1.
 Operators transform dimensionality, dimensions, and coordinates of the their non-character operator arguments into unique dimensionality, dimensions and coordinates of the operator result (see chapter 8.3)

ordinate is defined for a subset is assigned to the extent of the dimension of the operand. Coordinate

- Operands of dimensionality 0 do not have a coordinate assignment.
- Experiment targets and constants always have a dimensionality of 0.
- Consequently, a post-processor result as a sequence of operators applied to operands also has unique dimensionality, extents and coordinates.

# 8.2 Model Output Variables

- A variable of dimensionality n corresponds with a n-dimensional array and is defined at a n-dimensional
  grid, spanned up from the coordinate values of the assigned coordinates The complete data field of a
  model output variable or parts of it can be addressed in model output post-processing (see below). Dimensionality, dimensions and coordinate description of this data field is derived from the model variable
  description
- Model output variables are specified in the ASCII model output description file <model>.mdf by their
  - Name
  - Dimensionality
  - Extents
  - Coordinate assignment to each dimension (for all of above descriptions see Tab. 5.1 at page 20)
  - Data type (see Tab. 5.2 at page 21).
  - Use simenv.chk to check variables description in model output description file <model>.mdf
- Addressing of model output data fields or parts of it is done in model output post-processing by corresponding model output variables names.



- For variables with a dimensionality greater than 0 it is possible to address only a part of the whole variable field by
  - Specifying for a dimension an index range by
     i = index\_value\_1 { : index\_value\_2 }
     index\_value\_1 ≤ index\_value\_2
     index\_value\_2 = index\_value\_1 if index\_value\_2 is missing.
     i = stands for index addressing
  - Specifying for a dimension an coordinate range by
     c = coordinate\_value\_1 { : coordinate\_value\_2 }
     coordinate\_value\_1 ≤ coordinate\_value\_2 for strictly increasing coordinate values
     coordinate\_value\_1 ≥ coordinate\_value\_2 for strictly decreasing coordinate values
     coordinate\_value\_1 = coordinate\_value\_2 if coordinate\_value\_2 is missing
     c= stands for coordinate addressing
  - Index and coordinate ranges are separated from each other by a comma, the sequence of ranges for all dimensions is enclosed in brackets and is appended after the variable name.
    - For one variable c= and i= can be used in mixed mode for different dimensions.
    - \* denotes the complete range of a dimension.
    - c= \* is identical to i= \* is identical to \*
  - In the general SimEnv configuration file <model>.cfg (see chapter 10.1 at page 87) a global default for index and/or coordinate addressing is established for the whole post-processing session. This global default can be overwritten locally by using c= and/or i=.

```
Having a model variable definition as in Example 5.1 at page 22 then in model output
post-processing
                                                 and
at.mo
atmo(*,*,*,*)
                                                 and
atmo(c=*,*,i=*,*)
atmo(c=88:-88, c=-178:178, c=1:16, c=1:20) and
atmo(i=1:45, i=1:90, i=1:4, i=1:20)
                                                and
atmo(i=1:45, c=-178:178, *, *)
                                                and
atmo(1:45,1:90,1:4,1:20)
                                                and (with address default = index in model.cfg)
atmo(1:45, c=-178:178, 1:4, 1:20)
                                                and (with address default = index in model.cfg)
                                     all address all 45*90*4*20 values and
                                     the following holds true for this addressed variable:
                                     Dimensionality = 4
                                     Coordinates = lat , lon , level , time
                                     Extents = 45, 90, 4, 20
                                     addresses all values of last 10 decades
atmo(*,*,*,c=11:20)
                                     Dimensionality = 4
                                     Coordinates = lat , lon , level , time
                                     Extents = 45, 90, 4, 10
atmo(*,*,c=1,c=1)
                                     addresses all values of the first decade for level 1
                                     Dimensionality = 2
                                     Coordinates = lat , lon
                                     Extents = 45.90
                                     addresses all values of level 1for the last decade at
atmo(c=0, *, 1, i=20)
                                     equator
                                     Dimensionality = 1
                                     Coordinates = Ion
                                     Extents = 90
                                     addresses all values of level 1for the last decade at
atmo(i=23,*,1,i=20)
                                     equator
                                     Dimensionality = 1
                                     Coordinates = Ion
                                     Extents = 90
```

```
addresses the value for the last decade at (lat,lon,level,time) = (0^{\circ},2^{\circ},1,20) Dimensionality = 0 Coordinates = (without) Extents = (without) addresses the values for the last decade at (lat,lon,level,time) = (0^{\circ},2^{\circ},1,20) and (0^{\circ},6^{\circ},1,20) Dimensionality = 1 Coordinates = lon Extents = 2 atmo (c=0, c=1, c=1, c=20) error in addressing: c=1 for lon does not exist Example file: world.post_bas
```

**Example 8.1** Addressing model output variables in model output post-processing

## 8.3 Operators

- Operators transform dimensionality, dimensions, and coordinates of the their non-character operator arguments into unique dimensionality, dimensions and coordinates of the operator result.
  - Multi-argument operators that demand a certain relation between dimensionalities, dimensions and coordinates of their arguments
  - Single-argument operators that replicate dimensionality, dimensions and coordinates from the only argument to the operator result
  - Operators that increase dimensionality of the operator result and assign new coordinates to the additional dimensions (check Tab. 8.1 below) or form new coordinates from resulting target adjustments (see chapter 8.5.2 for behavioural analysis).
- SimEnv post-processing operators may have two special types of arguments:
  - Character arguments:
    - Only character strings enclosed in ' ' are valid as arguments. Some built-in operators (e.g., count) have a pre-defined set of valid character argument strings (e.g., for operator count strings all, def, and undef)
  - Integer or float constant arguments:
    - Only constants in appropriate format are valid as arguments. Model variables of dimensionality 0 or general operands with dimensionality 0 are invalid.
    - Examples: bios\_g, sin(bios\_g)
  - If defined, character and constant arguments are always the first arguments of an operator. If both argument types are defined for an operator then the sequence is character arguments followed by constant arguments.
- Operators are generic with respect to the data types of their operands: Each non-character argument can be used with operands of all defined data types (see chapter 5.1). Internally, arguments of any type are converted float representation. This may lead to undefined double arguments in float representation.
- Results of SimEnv post-processing operators are always of the type float.
- SimEnv post-processing follows the standard approach for description of operators for basic as well as advanced built-in or user-defined operators.
- Advanced built-in or user-defined operators
  - Have a unique name and a number of operands
  - The sequence of operands is enclosed in parentheses directly after the operator name
  - Operands are separated by a comma.
  - Recursions of the same operator (also for user-defined operators) are possible.

    Example: log10 (min n (3 , min n (log10 (atmo (\*, \*, 1, c=20)) , 400) , 10\*bios g))
- Elemental operators use the common form of notation:

Example: atmo g + 345



Coordinate name	Coordinate values (check Tab. 11.4 for syntax)	Operators, generating the coordinate
bin	equidist_end 1(1) number_of_bins	hgr, hgr_l, hgr_e
index	equidist_end 1(1) operator_dependent	minprop, maxprop, minprop_I, maxprop_I
run	equidist_end 1(1) number_of_runs	ens
stat_measure	equidist_end 1(1)10	stat

Tab. 8.1Additional coordinates

### 8.3.1 Operands and Coordinate Checking

The requirement for a lot of operators that their arguments must have same coordinates for same dimensions may restrict application of post-processing especially for hypothesis check heavily. To enable a broader flexibility with respect to this situation a general solution is provided by SimEnv post-processing: With the sub-keyword coord\_check in the general configuration file <model>.cfg three different modi can be assigned globally to SimEnv post-processing:

- coord check = strong
  - To ensure for two arguments with same dimensionalities and extents to have same coordinates it is necessary that
  - Assigned coordinate values for corresponding dimensions are unique
  - Assigned coordinate names for corresponding dimensions are unique coord check = strong is the default
- coord check = weak
  - To ensure for two arguments with same dimensionalities and extents to have same coordinates it is necessary that
  - Assigned coordinate values for corresponding dimensions are unique
  - Assigned coordinate names may differ.
  - Coordinate description of an appropriate operator result dimension is delivered from the first operand.
- coord\_check = without
  - To ensure for two arguments with same dimensionalities and extents to have same coordinates
- Neither coordinate names nor coordinate values for corresponding dimensions are checked Coordinate description of an appropriate operator result dimension is delivered from the first operand. Check Example 8.2 for some instances.

Having a model variable definition as in Example 5.1 at page 22 then the checking rules for coordinates are applied in the following manner to operands with dimensionality 1:

Expression	Same coordinates for <coord_check> =</coord_check>		
	strong	weak	without
bios(*,*,*) + atmo(c=84:-56,*,c=1,*) (same coordinate names, same coordinate values)	yes	yes	yes
<pre>atmo_g(*) + hgr(20,atmo) (differing coordinate names, same coordinate values)</pre>	no	yes	yes
<pre>atmo_g(c=10:20) + atmo_g(c=6:16) (same coordinate names, differing coordinate values)</pre>	no	no	yes
atmo_g(c=20) + atmo(c=0,c=2,c=1,c=1) (two operands with dimensionality 0)	yes	yes	yes

While determination of coordinate information is unique for <coord\_check> = strong, coordinate information is delivered from the first summand for <coord\_check> = [ weak | without ].

**Example 8.2** Checking rules for coordinates



# 8.4 Built-in Elemental, Basic, and Advanced Operators

### 8.4.1 Built-in Elemental Operators

Name	Meaning	Argument restriction(s) / result description (see Tab. 8.3)	Argument value restriction	Precedence
(	left parenthesis	-		first
)	right parenthesis	-		first
arg1 ** arg2	exponentiation	(2)	arg1 > 0	second
arg1 * arg2	multiplication	(2)		third
arg1 / arg2	division	(2)	arg2 ≠ 0	third
arg1 + arg2	addition (dyadic +)	(2)		fourth
arg1 – arg2	subtraction (dyadic -)	(2)		fourth
+ arg	identity (monadic +)	(1)		fourth
– arg	negation (monadic -)	(1)		fourth

Tab. 8.2Built-in elemental operators

- n-dimensional matrix algebra of built-in elemental operators is performed element by element Example:
  - atmo(\*,\*,1,\*) \* bios(\*,\*,\*) = "atmo(i,j,1,k) \* bios(i,j,k)" for all addressed(i,j,k)
- If an argument value restriction is not fulfilled for an operand element the corresponding element of the operator result is undefined.

Argument restriction(s) / result description		Argument restriction(s)	Result description (check chapter 8.1 for syntax)	
(1)		dimensionality, extents and coordinates of the only non-character argument <u>arg</u> can be arbitrary	same dimensionality, extents and coordinates as the only non-character argument: dim(res) = dim( <u>arg</u> ) ext(res,j) = ext( <u>arg,j</u> ) for all j coord(res,j) = coord( <u>arg,j</u> ) for all j	
(2) = (2.1) or (2.2)	(2.1)	all non-character arguments with same dimensionality, extents and coordinates (*) arg	same dimensionality, extents and coordinates as all the non-character arguments: dim(res) = dim(arg) ext(res,j) = ext(arg,j) for all j coord(res,j) = coord(arg,j) for all j	
	(2.2)	some non-character arguments with same non-zero dimensionality, extents and coordinates (*) arg, all the other non-character arguments with dimensionality zero	same dimensionality, extents and coordinates as all the non-character arguments with non-zero dimensionality:  dim(res) = dim(arg) ext(res,j) = ext(arg,j) for all j coord(res,j) = coord(arg,j) for all j the zero-dimensional argument is applied to each element of the non-zero dimensional argument	

Argument restriction(s) / result description		Argument restriction(s)	Result description (check chapter 8.1 for syntax)	
(3)		dimensionality, extents and coordinates of the only non-character argument can be arbitrary	dimensionality 0: dim(res) = 0	
(4) = (4.1) or (4.2)	(4.1)	all non-character arguments with same dimensionality, extents and coordinates (*)	dimensionality 0: dim(res) = 0	
	(4.2)	some non-character arguments with same non-zero dimensionality, extents and coordinates (*), all the other non-character arguments with dimensionality zero	dimensionality 0: dim(res) = 0 the zero-dimensional argument is applied to each element of the non- zero dimensional argument	
(5)		dimensionality, extents and coordinates of the first non-character argument <u>arg</u> can be arbitrary, all the other character arguments have to have dimensionalities, extents and coordinates (*) of this argument or have to have dimensionality 0	same dimensionality, extents and coordinates as the first non-character argument: dim(res) = dim(arg) ext(res,j) = ext(arg,j) for all j coord(res,j) = coord(arg,j) for all j	
(6)		without arguments	dimensionality 0: dim(res) = 0	

Tab. 8.3Classified argument restriction(s) / result description(\*): for the different levels of checking a coordinate description see chapter 8.3.1

# 8.4.2 Built-in Basic and Advanced Operators

Name	Meaning	Argument restriction(s) / result description (see Tab. 8.3)	Argument value restriction	Example		
Basic operators						
abs(arg)	absolute value	(1)		abs(-3) = 3.		
dim(arg1,arg2)	positive difference	(2)		dim(10,5) = 5. dim(5,10) = 0.		
exp(arg)	exponentiation	(1)		exp(1.) = 2.7183		
int(arg)	truncation value	(1)		int(7.6) = 7. int(-7.6) = -7		
log(arg)	natural logarithm	(1)	arg > 0	log(2.7183) = 1.		
log10(arg)	decade logarithm	(1)	arg > 0	log10(10) = 1.		
mod(arg1,arg2)	remainder	(2)	arg2 ≠ 0	mod(10,4) = 2.		
nint(arg)	round value	(1)		nint(7.6) = 8.		
sign(arg)	sign of value	(1)		sign(-3) = -1. sign(0) = 0.		
sqrt(arg)	square root	(1)	arg ≥ 0	sqrt(4) = 2.		
Trigonometric operators						
sin(arg)	sine	(1)		sin(0) = 0.		
cos(arg)	cosine	(1)		cos(0) = 1.		
tan(arg)	tangent	(1)	arg ≠ π/2±n*π	tan(0) = 0.		
cot(arg)	cotangent	(1)	arg ≠ ±n*π	cot(1.5708) = 0.		
asin(arg)	arc sine	(1)	abs(arg) ≤ 1	asin(0) = 0.		

Name	Meaning	Argument restriction(s) / result description (see Tab. 8.3)	Argument value restriction	Example
acos(arg)	arc cosine	(1)	abs(arg) ≤ 1	acos(1) = 0.
atan(arg)	arc tangent	(1)	\ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \	atan(0) = 0.
acot(arg)	arc cotangent	(1)		acot(0) = 1.5708
sinh(arg)	hyperbolic sine	(1)		sinh(0) = 0.
cosh(arg)	hyperbolic cosine	(1)		cosh(0) = 1.
tanh(arg)	hyperbolic tangent	(1)		tanh(0) = 0.
coth(arg)	hyperbolic cotangent	(1)	arg ≠ 0	coth(3.1416) = 1.
· (- <b>3</b> )	, ,, <u> </u>	ellaneous operators		, ,
	1	<u> </u>		T
classify(arg1, arg2)	classify arg2 into arg1 classes	(1) dim(arg2) > 0 arg1 = number of classes 2 ≤ arg1 ≤ number_of_values of arg2 = 0: automatic determination: number of classes = max(2,number_of_values/10) integer constant argument		classify( (10,atmo)
clip(arg1,arg2)	clip arg2 according to arg1	dim(arg2) > 0 dim(res), ext(res,i) depend on arg1 and arg2 arg1 = clip range character argument		clip( '0,*,1,10', atmo)
cumul(arg1,arg2)	cumulates arg2 according to arg1	(1) dim(arg2) > 0 arg1 = cumulation indicator per dimension character argument		cumul('0001', atmo)
flip(arg1,arg2)	flips arg2 according to arg1	(1), but coordinates is also flipped		flip('0001', atmo)
get_experiment (arg1,arg2, arg3,arg4)	et_experiment (arg1,arg2, arg3,arg4)  include an other experiment (arg1,arg2, arg3,arg4)  (1)  arg1 = experiment directory character argument arg2 = model experimented with character argument arg3 = file to transform result coordinates character argument arg4 = result from this experiment		ment nented with ment n result ment	<pre>get_experiment( 'mod_res', 'mod', 'mod.trf', avg(atmo)-400.)</pre>
get_table_fct( arg1,arg2)	table function with linear interpolation of table arg1 applied to arg2	(1)		<pre>get_table_fct ('table.usr', atmo)</pre>
if(arg1, arg2,arg3,arg4)	Teonomonal il-constituer i enaracier aroument		if('<',atmo,400, atmo)	
mask( arg1,arg2,arg3)	masks values of arg2 (set them undefined) by comparing arg2 and arg3 using operator arg1	(5) arg1 = comparison operator		mask('<',atmo,

Name	Meaning	Argument restriction(s) / result description (see Tab. 8.3)	Argument value restriction	Example
matmul(arg1, arg2)	matrix multiplication	dim(arg1) = dim(arg2) = dim(res) = 2 ext(res,i) according to matrix multi- plication rules		
nr_of_runs	number of single runs in the experiment	(6)		nr_of_runs()
rank(arg1,arg2)	assigns rank numbers to arg2 according to ranking type argument arg1	(1) dim(arg2) > 0 arg1 = ranking type [ tie_plain   tie_min   tie_avg ]		<pre>rank('tie_avg', atmo)</pre>
run(arg1,arg2)	values of arg 2 for the selected single run number explicitly or implicitly coded in arg1	(1)  arg1 = run number selection  = 0 for default run (all experiment types)  = <run_number> (for Monte Carlo analysis, 0 ≤ arg1 ≤ number_of_runs)  = filter argument (for behavioural analysis, same as filter argument of operator behav, check chapter 8.5.2) character argument</run_number>		<pre>run('0',atmo) run('sel_t(p1(4)) ',atmo)</pre>
transpose(arg1, arg2)	transpose arg2 according to sequence in arg1	dim(arg2) > 1 $dim(res) = dim(arg2)$		transpose ('3142',atmo)
undef()	undefined value	(6)		undef()

**Tab. 8.4** Built-in advanced operators (without standard aggregation / moments operators)

The following explanations yield for the operators in Tab. 8.4:

- All operators but experiment and matmul are applied to each element of the argument(s). These
  operators deal with an unfulfilled argument value restriction for an operand element in a way that the corresponding element of the operator result will be undefined.
- The **operator classify** transforms an operand arg2 that has dimensionality > 0 into arg1 classes 1 ,..., arg1. Classes are assumed to be equidistant. Lower boundary of the dynamics range from arg2 assigned to class number 1 is the minimum of all values of arg2, upper boundary assigned to class number arg1 is the maximum of all values of arg2.
- The **operator clip** clips an operand that has dimensionality > 0. The portion to clip from the operand arg2 is described by character argument arg1. Argument arg1 uses syntax for model output variable addressing (see chapter 8.1 at page 55). Note, that for all dimensions of argument arg2 lower bound index is 1. This applies also to model variables where the lower bound index is unequal 1 in the model output description file. In general, extents differ between the result of the operator clip and the argument arg2. Clip reduces the dimensionality of the result with respect to the argument arg2 to clip if the portion to be clipped is limited to one value for at least one dimension.
  - A character argument arg1 = '\*,\*,...' results for operator clip in the identity of argument arg2.



- The **operator cumul** cumulates an operand that has dimensionality > 0. Cumulation is performed for all values of the argument arg2 from the first addressed index position up to the current index position. With the character argument arg1 these dimensions are identified that are to be cumulated. Character 1 at position i means cumulation across dimension i while a 0 stands for no accumulation. cumul('0...0',arg) results in the identity to arg.
- The **operator flip** enables flipping of variable fields. For a one-dimensional field (a vector) flip changes the value of the first index position with the value of the last position, the value of the second position with that of the last but one position, etc. With the character argument arg1 these dimensions are identified that are due to flip. Character 1 at position i means flipping also for dimension i while a 0 stands for no flipping at this dimension. Flipping includes adaptation of coordinates and the assigned grid. cumul('0...0',arg) results in the identity to arg.
- The **operator get\_experiment** is to access to external SimEnv model output from the same or an other model performed with the same or another experiment type and stored in the same or in an other model output format. Model variables can differ from that used for the current model. Use for the experiment directory arg1 always that working directory the external experiment was started from. The external experiment is always post-processed completely over all single runs. Environment variables from operating system level in the specification of the directory are not allowed. Argument arg3 is the coordinate transformation file. It can be used to transform coordinates from the external result for usage in the current expression of the current experiment. If no coordinate transformation file is to be used argument arg3 is empty (''). If after potential application of an coordinate transformation file the imported expression has same coordinate names as defined in the original experiment coordinate descriptions are checked against each other, otherwise coordinate descriptions are imported from the external into the original experiment. For syntax of coordinate transformation files check chapter 11.3.

Attention: Make sure no SimEnv service is running from the directory arg1 of the external experiment.

- With the **operator get\_table\_fct** a table function arg1 is applied to each element of the operand arg2. If necessary, table values are interpolated linearly. Outside the definition range of the table function the first and/or the last table value is used. File arg1 to hold the table function must be an ASCII file with two columns: The first column of each line is the argument value x, the second column the function value f(x). Arguments have to be ordered in a strictly increasing manner. Syntax rules for comments and separators in the table function file are the same as for user defined files (check chapter 11.2). Environment variables from operating system level in the specification of the file name arg1 are not allowed. Check the table function world.dat\_tab in the examples directory of \$SE\_HOME for more information.
- The **operator** if supplies a general conditional if-construct. It operates for each element of the operand arg2 in the following way:

```
if (condition(arg1,arg2)) then
                res=arg3
        else
                res=arg4
        endif
                                arg2 < 0
with condition(arg1,arg2):
                                                         (arg1 = '<')
                                arg2 ≤ 0
                                                          (arg1 = '<=')
                                arg2 > 0
                                                         (arg1 = '>')
                                arg2 ≥ 0
                                                         (arg1 = '>=')
                                                         (arg1 = '=')
                                arg2 = 0
                                arg2 != 0
                                                         (arg1 = '!=')
                                arg2 def
                                                         (arg1 = 'def')
                                arg2 undef
                                                         (arg1 = 'undef')
```

• The **operator mask** supplies a method to mask values. It operates for each element of the operand arg2 in the following way:

```
if (condition(arg1,arg2,arg3)) then
                res=undef()
                res=arg2
        endif
with condition(arg1,arg2,arg3): arg2 < arg3
                                                        (arg1 = '<')
                                arg2 ≤ arg3
                                                        (arg1 = '<=')
                                arg2 > arg3
                                                        (arg1 = '>')
                                arg2 ≥ arg3
                                                        (arg1 = '>=')
                                arg2 = arg3
                                                        (arg1 = '=')
                                arg2 != arg3
                                                        (arg1 = '!=')
```

- The **operator matmul** performs a simple matrix multiplication for 2-dimensional arguments arg1 and arg2.
- The **operator nr\_of\_runs** returns the number of performed single runs of the current post-processed experiment without the run number 0 of the nominal constellation. It does not have an argument.
- The **operator rank** transforms all values of an operand arg2 that has dimensionality > 0 into their ranks. Small values get low ranks, large values get high ranks. Character argument arg1 determines how to rank ties, i.e., values of arg2 that are identical or have a maximum absolute difference of 1.e-6:

```
Assume an argument arg2 with 6 values

arg1 = 'tie_plain' returns ranks

(2, 1, 2, 2, 2, 3)

same minimal rank 2; next rank is 3,
does not take into account the number of identical values

(2, 1, 2, 2, 2, 6)

same minimal rank 2; next rank is 6,
taking into account the number of identical values

(3.5, 1, 3.5, 3.5, 3.5, 6)

same mean rank 3.5; next rank is 6,
taking into account number of identical values
```

- The operator run selects a single run from the run ensemble. The operator run must not contain experiment-specific (multi-run) operators as operands, while these operators may refer to the operator run. Additionally, run must not contain itself as an argument.

  Character argument arg1 can hold explicitly the run number string Monte Carlo analysis. Run number 0 corresponds with the default single run 0 and is permitted as arg1 for all experiment types. For behavioural analysis a filter of the operator behav (see chapter 8.5.2) is applied as arg1 to select a unique run number unequal zero. For this purpose, a single run can be selected by the select-operator (check Tab. 8.11) of the operator behav. For Monte Carlo analysis, single runs with a run number unequal zero are selected explicitly. Therefore, the file <model>.edf\_adj holds the targets to be adjusted to the default values for the current experiment. Run number n corresponds with record number n of this file. For more information on <model>.edf\_adj check chapter 6.1 at page 35. For examples see Example 8.4 and Example 8.5.
- The **operator transpose** enables to transpose an operand that has a dimensionality > 1. Sequence of extents of the transposed result is described by character argument 1: It consists of figures 1, ..., dim(arg2) where the figure sequence corresponds with the re-ordered sequence of the operator result extents.

A character argument arg1 = '123...' results for operator transpose in the identity of argument arg2.

• The **operator undef** supplies a 0-dimensional result as undefined. This operator can be used as an argument for the if-operator.



Generic aggregation and moment operator	Meaning
min	minimum of values
max	maximum of values
sum	sum of values
avg	arithmetic mean of values
var	variance of values
avgg	geometric mean of values
avgh	harmonic mean of values
avgw	weighted mean of values
hgr	histogram of values
count	number of values
minprop	minimal, suffix related property of values
maxprop	maximal, suffix related property of values

 Tab.
 8.5
 Built-in generic standard aggregation / moment operators

The generic operators in Tab. 8.5 can be applied during model output post-processing to derive aggregations and moments from operands in different ways by appending suffixes to the generic operator name:

### • Appending **no suffix**:

Aggregate the only non-character argument(s)

Result is a scalar (an operator result of dimensionality zero) for all but operators hgr, minprop and max-prop.

For operator hgr dimensionality of the result is 1, the extent is the specified number of bins for the histogram and the coordinate assigned has the name bin. Coordinate values are equidistant with 1 as the first value and an increment of 1.

For operators minprop and maxprop dimensionality of the result is 1. For argument dimensionality greater / equal 1 extent of the result is equal to the argument dimensionality. Assigned coordinate name is index. Coordinate values are equidistant with 1 as the first value and an increment of 1. For argument dimensionality 0 result dimensionality is 0.

#### • Appending **suffix \_n** (for n arguments)

Aggregate an arbitrary number of arguments with argument restriction(s) / result description according to (2) in Tab. 8.3 at page 60 element by element

Currently, only operators min n and max n are implemented.

Result has same dimensionality, extents and coordinates as the arguments

### • Appending **suffix \_I** (for loop)

Aggregate the only non-character argument(s) separately for selected dimensions. Dimensions to select are described by an additional loop character argument (corresponds with the group by-clause of the standard query language SQL of relational database management systems).

Result has a lower dimensionality as the only non-character argument according to the loop character argument.

For operator hgr\_l, dimensionality is increased additionally by one, the additional extent is the specified number of bins for the histogram and the additional coordinate assigned to has the name bin. Coordinate values are equidistant with 1 as the first value and an increment of 1.

For operators minprop\_I and maxprop\_I dimensionality is modified in the same manner like for operators minprop and maxprop, respectively.

Aggregation and moment operator	Argument restriction(s) / result description (see Tab. 8.3)		
min(arg)			
max(arg)			
sum(arg)			
avg(arg)	(3)		
var(arg)			
avgg(arg)			
avgh(arg)			
avgw(arg1,arg2)	(4.1)		
	arg2 = weight		
hgr(arg1,arg2)	dim(res) = dim(arg2)+1		
	ext(res,dim(res)) = number of bins		
	coord(res,dim(res))= name = bin		
	values = equidist_end 1(1) number of bins		
	arg1 = number of bins: 4 ≤ arg1 ≤ number_of_values or		
	= 0: automatic determination:		
	number of bins = max(4,number_of_values/10)		
201101/2021 2022)	integer constant		
count(arg1,arg2)	(3)		
	arg1 = character argument = [ all   def   undef ]		
	dim(res)		
minprop(arg)	ext(res,1) = dim(arg)		
	$dim(res) = 0 \qquad else$		
maynron(arg)	returns the index of that element of arg where the extreme is reached the first		
maxprop(arg)	time according to the processing sequence of the argument field arg by the		
	Fortran column-wise storage model (see Glossary).		

 Tab.
 8.6
 Built-in standard aggregation / moment operators without suffix

Aggregation and moment operator	Argument restriction(s) / result description (see Tab. 8.3)	
min_n(arg1,,argn)	(4)	
max_n(arg1,,argn)	(4)	
minprop_n(arg1,,argn)	(4)	
maxprop_n(arg1,,argn)	returns per result element the argument position (1 n) where the extreme is reached the first time. Processing sequence starts with arg1.	

Tab.8.7Built-in standard aggregation / moment operators with suffix \_n

Aggregation and moment operator	Argument restriction(s) / result description		
min_l(arg1,arg2) max_l(arg1,arg2) sum_l(arg1,arg2) avg_l(arg1,arg2) avgg_l(arg1,arg2) avgh_l(arg1,arg2) avgw_l(arg1,arg2,arg3) hgr_l(arg1,arg2,arg3)	dim(non-character argument(s))  > 1  ext(non-character argument(s))  = arbitrary  dim(res), ext(res,i) according to  arg1 and the non-character argument(s)  arg1 = loop character argument	dim(arg2) = dim(arg3) ext(arg2,i) = ext(arg3,i) arg3 = weight  dim(res) = 1 + dim(res) of all other operators ext(res,dim(res)) = number of bins coord(res,dim(res)) = name = bin values = equidist_end 1(1) number of bins arg2 number of bins 4 ≤ arg1 ≤ number_of_values or 0: automatic determination = max(4,number_of_values/10) integer constant arg2 = [ all   def   undef ]	
	as shove but	character argument	
minprop_l(arg1,arg2)	as above, but:	returns the indices of those elements of	
maxprop_l(arg1,arg2)	dim(res) is increased by 1 w.r.t. above.	arg2 where the extreme is reached the first time according to arg1 and to a Fortran-like	
	ext(res,dim(res)) = dim(arg1)	(column-wise) processing sequence (see	
	coord(res,dim(res)): name = index	Glossary) of the argument field arg2.	
	values =	Sisseary, or the argument held argz.	
	equidist_end 1(1)"n"		

 Tab.
 8.8
 Built-in standard aggregation / moment operators with suffix \_I

The loop character argument is characterised as follows:

- The length of the string is equal to the dimensionality of the non-character argument
- The string consists of 0 and 1
- 0 at position n means: aggregate over the corresponding dimension n of the argument
- 1 at position n means: do not aggregate over the corresponding dimension n of the argument
- Loop character arguments completely formed of 0 or 1 are forbidden

For the operator hgr\_l bins are determined on the base of the minimum and maximum value of the total argument arg2.

assuming address default=coordinate in <model>.cfg then in model output post-processing value of variable atmo g atmo g Dimensionality = 1 Coordinates = time Extents = 20maximum from atmo and 10 for each element of atmo if ('<', atmo-10, 10, atmo) equivalent with max\_n(atmo,10) Dimensionality = 4 Coordinates = lat , lon , level , time Extents = 45, 90, 4, 20 last two decades for level 1 at equator clip('i=23,\*,1,19:20',atmo)equivalent with atmo(i=23,\*,1,19:20) Dimensionality = 2 Coordinates = lon , time Extents = 90.2global all-level mean over the last two decades avg(atmo(\*,\*,\*,19:20)) Dimensionality = 0 Coordinates = (without) Extents = (without) indices of this element of atmo where the maximum of atmo maxprop(atmo) is reached the first time Dimensionality = 1 Coordinates = index Extents=4 min n(atmo(84:-56,\*,1,19:20),10.)minimum per grid cell for level 1 without polar regions for the last two decades from atmo and 10 Dimensionality = 3 Coordinates = lat , lon , time Extents = 36, 90, 2min 1('10', atmo(20:-20, \*, 1, 20))zonal tropical minima of atmo for the last decade and level 1 Dimensionality = 1 Coordinates = lat Extents = 11 minprop 1('10', atmo(20:-20, \*, 1, 20))zonal tropical indices of those elements of atmo for the last decade and level 1 where the minimum is reached the first time Dimensionality = 2 Coordinates = lat , index Extents = 11, 2hgr 1('10',8,atmo(20:-20,\*,1,20)) zonal tropical histograms with 8 bins of atmo for the last decade and level 1 Dimensionality = 2 Coordinates = lat, bin Extents = 11, 8 avg 1('100', min 1('1011', atmo(20:-20, \*, \*, \*))) temporally averaged all-level zonal tropical minima Dimensionality = 1 Coordinates = lat Extents = 11

Having a model variable definition as in Example 5.1 at page 22 and



**Example 8.3** Post-processing with advanced operators

### 8.5 Experiment-Specific Operators

- Experiment-specific operators are to navigate and process in the experiment space.
- Experiment specific operators must not be applied recursively.
- Addressing a variable within an experiment specific operator normally results in application of the operator on the whole run ensemble or parts of it and in aggregating across the run ensemble according to the operator.
- Addressing a variable outside an experiment specific operator results in application of the basic, advanced and/or user-defined operator on the variable for the default run number 0 of the experiment.
- If the dimensionality of an operator result is higher than that of one of its operands the additional dimensions of the result are appended to the dimensions of the operand. Examples for such operators are ens (for Monte Carlo analysis post-processing) and behav (for certain constellations of behavioural analysis post-processing).

### 8.5.1 Multi-Run Standard Aggregation / Moment Operators

Tab. 8.9 summarises multi-run standard aggregation / moment operators for behavioural analysis, Monte Carlo analysis and optimization. They work on the whole run ensemble (for Monte Carlo analysis and optimization) or parts of it (for certain constellations of behavioural analysis post-processing). They are used with suffix \_e for Monte Carlo analysis and optimization and without any suffix for behavioural analysis.

Aggregation and moment operator	Argument restriction(s) / result description (see Tab. 8.3)		
min(arg)			
max(arg)			
sum(arg)			
avg(arg)	(1)		
var(arg)			
avgg(arg)			
avgh(arg)			
avgw(arg1,arg2)	(2.1)		
	arg2 = weight		
hgr(arg1,arg2)	dim(res) =dim(arg2)+1		
	ext(res,dim(res)) = number of bins		
(heuristic probability	coord(res,dim(res))= name = bin		
density function)	values = equidist_end 1(1) number of bins		
	arg1 = number of bins		
	4 ≤ arg1 ≤ number_of_runs or		
	0: automatic determination = max(4,number_of_runs/10)		
	integer constant		
count(arg1,arg2)	(1)		
	arg1 = [all   def   undef ]		
	character argument		
minprop(arg)	(1)		
maxprop(arg)	returns the run number where the extreme is reached the first time.		
	Processing sequence starts with run number 1.		

 Tab. 8.9
 Multi-run standard aggregation / moment operators

### 8.5.2 Behavioural Analysis

There is only one experiment specific operator for behavioural analysis. With this operator behav

- A single run can be selected from the run ensemble
- The complete run ensemble can be addressed
- Sub-spaces from the experiment space can be addressed and
- Sub-spaces can be projected by aggregation and moment operators

dependent on the way the experiment target space was to be scanned according to the comb-sub-keyword in the experiment description file.

To show the power of the operator behav the simple experiment layouts as described in Fig. 4.3 at page 12 are used as examples.

- With behav it is possible to address for any operand a single run out of the run ensemble by fixing values of experiment targets p1 and p2 (for Fig. 4.3 (a)), a value of the parallel targets p1 or p2 (for Fig. 4.3 (b)), and values of targets p3 and p1 or p2 (for Fig. 4.3 (c)). Dimensionality and extents of the operator result is the same as that of the operand.
- Without any selection in the target experiment space (p1,p2) and/or (p1,p2,p3) the dimensionality of the operator result is formed from the dimensionality of the operand enlarged by the dimensionality of the experiment space. Two additional dimensions are appended to the operand for Fig. 4.3 (a), one additional dimension for Fig. 4.3 (b), and two additional dimensions for Fig. 4.3 (c). For the latter two cases it is important which of the axis p1 and p2 is used for further processing and/or output of the operator result. The extents of the appended dimensions are determined by the number of target adjustments.
- As a third option it is possible to select only a sub-space out of the experiment space to append to the operand. For Fig. 4.3 (a) this could be the sub-space formed from the first until the third adjustment value of p1 and all adjustment values of p2 between 3 and 7. Dimensionality of the operator result increases by 2 and extents of these additional dimensions are 3 and 2 with respect to the corresponding Example 6.1 (a) in chapter 6.2.3 at page 37.



• The operator behav also enables to aggregate operands in the experiment space. In correspondence with the example in the last bullet point for Fig. 4.3 (a) the operand could be aggregated (e.g., averaged) over the first until the third adjustment value of p1 autonomously for all runs with different values of p2 and afterwards this intermediate result (that now depends only on p2) could be summed up for all adjustment values of p2 between 3 and 7. Consequently the result has the same dimensionality as the operand of behav. Sequence of performing aggregations is important.

Name	Meaning	Argument restriction(s) / result description	Argument value restriction
behav(arg1,arg2)	navigation and aggregation in the experiment space for arg2 according to arg1	arg1= selection / aggregation filter character argument according to Tab. 8.14 dim(res) = dim(arg2) + appended dimensions according to arg1	

 Tab.
 8.10
 Experiment-specific operators for behavioural analysis

Placeholder	Explanation		
<filter></filter>	' { <operator_1> {, <operator_2> {, <operator_n> } } } '</operator_n></operator_2></operator_1>		
<operator></operator>	[ <select_operator>   <aggreg_operator>   <show_operator> ]</show_operator></aggreg_operator></select_operator>		
<select_operator></select_operator>	sel { _ <target_value_type>} ( <target_name> { <target_value_range> } )</target_value_range></target_name></target_value_type>		
<aggreg_operator></aggreg_operator>	<pre><aggreg_type> {_<target_value_type>} (<target_name> { <target_value_range> } )</target_value_range></target_name></target_value_type></aggreg_type></pre>		
<show_operator></show_operator>	show( <target_name> )</target_name>		
<target_name></target_name>	name of the experiment target according to the experiment description file		
<target_value_type></target_value_type>	specification how to interpret <value_range></value_range>		
	i as adjustment indices (indices always count from 1)		
	v as adjustment values		
	t as resulting target values		
<target_value_range></target_value_range>	[ ( <value_1> { : <value_2> } )   (*) ]</value_2></value_1>		
	for <value_2> = <nul> : value_2 = value_1</nul></value_2>		
	(*): use all values from <target_name></target_name>		
<aggreg_type></aggreg_type>	an aggregation / moment operator from Tab. 8.5 at page 65.		
	The following restrictions apply:		
	aggregations avgw and hgr can not be used		
	aggregation count has a differing syntax:		
	count_ <target_value_type> ( [ all   def   undef ] ,</target_value_type>		
	<target_name> { <target_value_range> } )</target_value_range></target_name>		

Tab. 8.11Syntax of the filter argument 1 for operator behav

The following rules yield for the operator **behav**:

• Generally, by the filter argument arg1 those runs from the run ensemble are selected and/or aggregated (here interpreted as filtered) that are used for the formation of the result.

Consequently, if no filter is specified all runs are used:

behav(' ',atmo g)

The select operator has to be specified only if values are to be restricted by a corresponding target value range.

For the aggregation and the select operator the target value type is redundant if the value range represents the full range of values by <target\_name> or <target\_name>(\*):

sel(p1) = sel(p1(\*)) = seli(p1) = selv(p1) = selv(p1) and all are redundant.

- The show-operator can be used to force a certain experiment target to be used in the result of the operator behav if this target is used in parallel with other targets. By default, the first target of a parallel target sub-space as declared in the comb-line of the experiment description file is used in the behav-result.
- Aggregation operators reduce dimensionality of the covered experiment target space in the behav-result.
   The sequence of aggregation operators the first argument of the operator behav influences the result:
   Computation starts with the first aggregation operator and ends with the last:

```
avg(p1), min(p2) normally differs from min(p2), avg(p1)
```

- An unused experiment target in the selection and aggregation filter contributes with an additional dimension to arg2 to the result of the operator behav. The extent of this additional dimension corresponds with the number of adjustments to this target in the experiment description file.
  - A target that is restricted by any of the select operators also contributes with an additional dimension to the result of the operator behav if the number of selected values is greater than 1. The extent of the additional dimension corresponds with the number of selected values of this target by the select operator. Consequently, an empty character string arg1 forces to output the operand arg2 over the whole target space of the experiment.
- The name of the coordinate that is assigned to an additional dimension is the name of the corresponding target. Coordinate description and coordinate unit (see 5.1 at page 19) are associated with the corresponding information for the target from the experiment description file.
  - Coordinate values are formed from resulting target values. For strictly ordered target adjustments in the experiment description file and finally for strictly ordered resulting target values the coordinate values are ordered accordingly in an increasing or decreasing manner. Unordered target adjustments and finally unordered target values are ordered in an increasing manner for coordinate usage.
  - The result of the operator behav is always arranged according to ascending coordinate values for all additional dimensions.
- Independently from the sequence of the applied aggregation-, select- and show-operators the targets
  that contribute to additional dimensions of the result of the operator behav are appended to the dimensions of the operand arg2 of behav according to the sequence they are declared in the experiment description file (and not to the sequence they are used in the comb-line of the experiment description file).
   From parallel changing targets that target is used in this sequence that is addressed explicitly or implicitly by the show-operator.
- For experiment targets that are changed in the experiment in parallel, that increase dimensionality of the result and where a show-operator is missing the first target from this parallel sub-space in the comb-line is used in the result.
- For experiments that use an adjustment file (keyword file) instead of adjustment definitions (keyword comb) all experiment targets are assumed to be adjusted in parallel.

```
Having a model variable definition as in Example 5.1 at page 22 and
assuming address default=coordinate in <model>.cfg
Assume the experiment layout in Example 6.1 (c) at page 38 and
the corresponding experiment description file (c) from Example 6.1 at page 37
then in result-processing
behav(' ',bios(*,*,20))
                                     last time step of bios dependent on (p2,p1) and p3
                                     Dimensionality = 4
                                     Coordinates = lat , lon , p2, p3
                                     Extents = 36, 90, 4, 3
behav('show(p1)',bios(*,*,20))
                                     last time step of bios dependent on (p1,p2) and p3
                                     Dimensionality = 4
                                     Coordinates = lat , lon , p1, p3
                                     Extents = 36, 90, 4, 3
behav('sel t(p2(4)), sel i(p3(1))', atmo(*, *, 1, *))
                                     select the single run out of the run ensemble for level 1
                                     p2 = 4 and p3 = 3.3
                                     Dimensionality = 3
                                     Coordinates = lat , lon , time
                                     Extents = 45, 90, 20
```



```
behav('sel i(p2(1:3)), sel v(p3(1:2))', atmo(*, *, 1, 20))
                                     last time step of atmo for level 1 depend. on (p2,p1) and p3
                                     use only runs for p2 = 1, 2, 3 and for p3 = 3.3, 4.5
                                     Dimensionality = 4
                                     Coordinates = lat , lon , p2, p3
                                     Extents = 45, 90, 3, 2
behav('avg i(p2(1:3)), sel i(p3(2:3))', atmo(*, *, 1, *))
                                     mean of atmo for level 1 and for runs with p2 =1, 2, 3
                                     for each value of p3 = 4.5, 7.2
                                     Dimensionality = 4
                                     Coordinates = lat , lon , time , p3
                                     Extents = 45, 90, 20, 2
behav('min(p2), max(p3)', avg(atmo(*, *, 1, 19:20)))
                                     determine single minima of avg(atmo) for level 1 and the
                                     last two decades for each value of p2
                                     afterwards determine from that the maximum over all p3.
                                     Dimensionality = 0
                                     Coordinates = (without)
                                     Extents = (without)
behav('max(p3), min(p2)', avg(atmo(*, *, 1, 19:20)))
                                     Result differs normally from min(p2),max(p3)
                                     (previous expression)
behav('count(def,p3),sel_i(p2=1)',bios(*,*,20))/3
                                     determine single numbers of defined values of
                                     bios for last decade for runs with p2=1.
                                     Result consists of values 0 (for water) and 1 (for land)
                                     Dimensionality = 2
                                     Coordinates = lat , lon
                                     Extents = 36, 90
behav('', atmo(*, *, 1, 20) - run('sel i(p1(1)), sel i(p3(3))',
       atmo(*,*,1,20))
                                     deviation of the last time step of atmo for level 1
                                     from the run with p1=1, p2=1, p3=3.3
                                     dependent on (p2,p1) and p3
                                     Dimensionality = 4
                                     Coordinates = lat , lon , p2, p3
                                     Extents = 45, 90, 4, 3
                                                                     Example file: world.post c
```

**Example 8.4** Post-processing operator behav for behavioural analysis

### 8.5.3 Monte Carlo Analysis

Tab. 8.12 shows experiment specific operators for Monte Carlo analysis that can be used in post-processing besides the general multi-run aggregation operators listed in Tab. 8.9 at page 70 and supplemented with a suffix \_e. For a definition of these operators check Tab. 8.5 at page 65.

Name	Meaning	Argument restriction(s) / result description (see Tab. 8.3)	Argument value restriction
cnf(arg1,arg2)	positive distance of confidence measure from mean avg_e(arg2)	(1) arg1 error probability	arg1 = [ 0.001   0.01   0.05   0.1 ] real*4 constant argu- ment
cor(arg1,arg2)	correlation coefficient between arg1 and arg2	(2.1)	
cov(arg1,arg2)	covariance between arg1 and arg2	(2.1)	
ens(arg)	whole Monte Carlo run ensemble	<pre>dim(res) = dim(arg)+1 ext(res,dim(res)) =</pre>	
krt(arg)	kurtosis (4 <sup>th</sup> moment)	(1)	
med(arg)	median	(1)	
qnt(arg1,arg2)	quantile of arg2	(1) arg1 quantile value	0. ≤ arg1 ≤ 100. real*4 constant argu- ment
reg(arg1,arg2)	linear regression coefficient to forecast arg2 from arg1: arg2 = reg(arg1,arg2)*arg1 + n	(2.1)	
rng(arg)	range = max_e(arg) - min_e(arg)	(1)	
skw(arg)	skewness (3 <sup>rd</sup> moment)	(1)	
stat(arg1,arg2, arg3,arg4, arg5)	basic statistical measures of arg5	dim(res) = dim(arg)+1 ext(res,dim(res)) = 10 coord(res,dim(res)) =     name = stat_measure     values =         equidist_end 1(1)10	0. ≤ arg1 < arg2 ≤ 100. quantile values real*4 constant arguments arg3, arg4 = [ 0.001

Tab. 8.12Experiment-specific operators for Monte Carlo analysis<br/>(without standard aggregation / moment operators)

The following explanations yield for the operators in Tab. 8.12:

• The operator **stat** supplies basic statistical measures for argument arg5. The operator stat is a standalone operator: It must not be operand of any other operator. Contrary, argument arg5 can be composed from other non-multi-run operators. To store the statistical measures, dimensionality of stat is that of argument arg5, appended by an additional dimension with an extent of 10. Appended coordinate description is pre-defined by SimEnv (check Tab. 8.1).

These ten data fields correspond with the following statistical measures:

- 1. Deterministic run ( run number 0)
- 2. Run ensemble minimum
- 3. Run ensemble maximum
- 4. Run ensemble mean



- Run ensemble variance
- 6. Run ensemble median
- 7. Run ensemble quantile of quantile value arg1
- 8. Run ensemble quantile of quantile value arg2
- 9. Run ensemble positive distance of confidence measure from run ensemble mean for value arg3
- 10. Run ensemble positive distance of confidence measure from run ensemble mean for value arg4

For the definition of the statistical measures check the corresponding single operators in Tab. 8.9 and Tab. 8.12. Operator stat has been designed for application of an appropriate visualization technique in result evaluation in future.

```
Having a model variable definition as in Example 5.1 at page 22 and
assuming address default=coordinate in <model>.cfg
Assume the Monte Carlo experiment from Example 6.2 (e) at page 41
then in model output post-processing
                                     global run ensemble mean of p1*atmo for level 1
avg e(p1*atmo(*,*,1,19:20))
                                     and the last two decades
                                     Dimensionality = 3
                                     Coordinates = lat , lon , time
                                     Extents = 45, 90, 2
                                     global mean of atmo for level 1 and the last two decades
avg(atmo(*,*,1,19:20))
                                     for run number 0
                                     Dimensionality = 0
                                     Coordinates = (without)
                                     Extents = (without)
                                     run ensemble values of atmo for level 1 and the last decade
ens(atmo(*,*,1,20))
                                     Dimensionality = 3
                                     Coordinates = lat , lon , run
                                     Extents = 45, 90, 250
minprop_e(atmo(*,*,1,19:20)) run ensemble run number for level 1 and the last two
                                     decades
                                     where the minimum of atmo is reached the first time
                                     Dimensionality = 3
                                     Coordinates = lat , lon , time
                                     Extents = 45, 90, 2
var e(atmo(*,*,1,19:20))-atmo(*,*,1,19:20)
                                     anomaly for run ensemble variance from the nominal
                                     run for level 1 the last two decades
                                     Dimensionality = 3
                                     Coordinates = lat , lon , time
                                     Extents = 45, 90, 2
var e(atmo(*,*,1,19:20)-run('0',atmo(*,*,1,19:20)))
                                     global run ensemble variance of the anomaly of atmo for
                                     level 1 and the last two decades.
                                     Differs normally from the previous expression
                                     Dimensionality 4
                                     Coordinates = lat , lon , time
                                     Extents = 45, 90, 4, 20
hgr e(0, \min 1('10', atmo(20:-20, *, 1, 20)))
                                     histogram with 25 bins for the zonal tropical minima
                                     for level 1 and the last decade
                                     Dimensionality = 2
                                     Coordinates = lat, bin
                                     Extents = 11, 25
```

```
stat (25,75,0.01,0.05, min_l ('10',atmo(20:-20,*,1,20)))
basic statistical measures for the zonal tropical minima
of atmo for level 1 and the last decade
Dimensionality = 2
Coordinates = lat, stat_measure
Extents = 11, 10

Example file: world.post_e
```

**Example 8.5** Post-processing operators for Monte Carlo analysis

### 8.5.4 Local Sensitivity Analysis

Tab. 8.13 shows the experiment specific operators for local sensitivity analysis that can be used in post-processing. For a definition of these operators check Tab. 4.3 at page 15.

Name	Meaning	Argument restriction(s) / result description	Argument value restriction
sens_abs(arg1, arg2)	absolute sensitivity measure for arg2 according to arg1		
sens_rel(arg1, arg2)	relative sensitivity measure for arg2 according to arg1	arg1= selection /	
lin_abs(arg1, arg2)	absolute linearity measure for arg2 according to arg1	aggregation filter character argument	
lin_rel(arg1, arg2)	relative linearity measure for arg2 according to arg1	dim(res) = dim(arg2) + appended dimensions	
sym_abs(arg1, arg2)	absolute symmetry measure for arg2 according to arg1	according to arg1	
sym_rel(arg1, arg2)	relative symmetry measure for arg2 according to arg1		

 Tab.
 8.13
 Experiment-specific operators for local sensitivity analysis

Placeholder	Explanation			
<filter></filter>	'{ <select 1="" operator=""> {, <select 2="" operator=""> {, <select 3="" operator=""> } } }'</select></select></select>			
<select_operator></select_operator>	[ selt   seli   sels ] { _ <value_type>} ( <value_range> )</value_range></value_type>			
	with selt = select target range			
	seli = select increment range			
	sels = select sign range (only for sens_abs an	d sens_rel)		
<value_type></value_type>	specification how to interpret <value_range></value_range>			
	i as position indices (always count from 1) for selt and seli			
	v as increment values	for seli		
	n as target <b>n</b> ames for selt			
	as signs (+ or -) for sels			
<value_range></value_range>	[ ( <value_1> { : <value_2> } )   (*) ]</value_2></value_1>			
	for <value_2> = <nul> : value_2 = value_1</nul></value_2>			
	(*): use all values from <target_name></target_name>			

Tab.8.14Syntax of the filter argument 1 for the local sensitivity operators



The following rules yield for the filter argument in local sensitivity operators:

• Generally, by the filter argument arg1 those runs from the run ensemble are selected (here interpreted as filtered) that are used for the formation of the result.

Consequently, if no filter is specified all runs are used:

```
sens_abs(' ',atmo_g)
```

The filter operator has to be specified only if values are to be restricted by corresponding target values, increment values and/or sign ranges.

• For the above three select operators selt, seli and sels the value type is redundant if the value range represents the full range of values by [ selt | seli | sels ] (\*):

```
selt(*) = selt n(*) = selt i(*) and all are redundant.
```

- Each select operator can be applied only once within the filter argument.
- For <value\_type> = i, i.e. if a value range is specified by position indices those targets are selected for selt and/or those increments are selected for sell that correspond with the specified position indices. Position indices are assigned from index 1 to the targets and or increments according to their specification sequence in the corresponding experiment description file <model>.edf.
- If more than one target, increment value and/or sign was selected by the filter argument arg1 it contributes with an additional dimension to the result of the local sensitivity operator:

```
    For targets an additional dimension target_sequ
    For increments an additional dimension incr
    For signs an additional dimension sign
```

is appended to the dimensions of the argument arg2 to form the result of the local sensitivity operator. The extent of this additional dimension corresponds with the defined and/or selected number of targets, increment values and/or signs. For a definition of the additional dimensions check Tab. 10.7.

Firstly, dimension target\_sequ is appended on demand, secondly dimension incr and thirdly dimension sign.

```
Having a model variable definition as in Example 5.1 at page 22 and
assuming address default=coordinate in <model>.cfg
Assume the experiment description file (f) from Example 6.3 at page 43
then in result-processing
sens_abs(' ',atmo_g)
                                     absolute sensitivity measure for atmo q
                                     for all targets, increments and signs
                                     Dimensionality = 4
                                     Coordinates = time , target_sequ , incr , sign
                                     Extents = 20, 3, 4, 2
sens rel('sels n(+), selt i(1)', atmo g)
                                     relative sensitivity measure for atmo g
                                     for target p1 and all positive increments
                                     Dimensionality = 2
                                     Coordinates = time, incr
                                     Extents = 20, 4
sens abs('seli v(0.001:0.05)',atmo g)
                                     absolute sensitivity measure for atmo q
                                     for all targets, increment values 1 to 3 and all signs
                                     Dimensionality = 4
                                     Coordinates = time, target sequ, incr, sign
                                     Extents = 20, 3, 3, 2
lin abs('seli v(0.001:0.05)',atmo g)
                                     absolute linearity measure for atmo g
                                     for all targets and increment values 1 to 3
                                     Dimensionality = 3
                                     Coordinates = time , target sequ , incr , sign
                                     Extents = 20, 3, 3
                                                                      Example file: world.post f
```

**Example 8.6** Post-processing operators for local sensitivity analysis

### 8.5.5 Optimization

The goal of an optimization experiment is to minimize a cost function by determining the corresponding optimal point in the target space. Nevertheless, the specified model output from all single runs is stored during the experiment.

While the single run that corresponds with the optimal cost function can be post-processed in the single-run modus, the whole experiment can be post-processed as a Monte Carlo analysis. Keep in mind that the targets do not follow an pre-defined distribution.

### 8.6 User-Defined Operators

### 8.6.1 Declaration of User-Defined Operator Dynamics

- User-defined operators consist of a declarative and a computational part.
  - In the declarative part consistency of the non-character operands are checked and dimensionality, extents and coordinates of the result are defined.
  - In the computational part the result of the operator in dependency of the operands is computed.
- User-defined operators are specified in the ASCII operator description file <model>.odf. This file is used
  to check user-defined operators syntactically during result-post-processing.
- Check usr\_opr\_<opr>.f and apply the assigned operator <opr> for examples of user-defined operators.
- In SimEnv the declarative and computational part of an user-defined operator <opr> is hosted in a file usr\_opr\_<opr>>.f. The assigned executable has the name <opr>>.opr and has to be located in this directory that is stated in <model>.cfg as the hosting directory opr\_directory for user-defined operators.
- Use the shell script operator\_f.lnk <opr> to compile and link from usr\_opr\_<opr>.f an executable <opr>.opr that represents the user-defined operator <opr>.
- Use the simenv.chk to check user-defined operators
- Any user-defined operator can be transformed directly without changes to a built-in operator
- The functions to declare and compute user-defined operators listed below use a named common block simenv.

Function name	Function description	Inputs / outputs / function value	Inputs / outputs / function value description
	Functions to host	declarative and	computational part in <model>.f</model>
icheck_user_ def_operator ( )	checks consis- tency of operator arguments and defines dimen- sionality and di- mensions of result	integer*4 icheck_user_ def_operator (function value)	return code = 0 ok ≠ 0 inconsistency between operands
icompute_user_ def_operator (	computes result of the operator in dependency on	real*4 result(1) (output)	result vector of the operator
result )	operands	integer*4 icompute_user_ def_operator (function value)	return code = 0 ok ≠ 0 user-defined interrupt of calculation

 Tab.
 8.15
 Operator functions: Declarative and computational part



Function name	Function description	Inputs / outputs / function value	Inputs / outputs / function value description		
Functio	Functions to get and put structure information in declarative and computational part				
iget_char_arg ( iarg, char )	gets string and length of the string of a character argument	integer*4 iarg (input) character*100 char (output) integer*4 iget_char_arg	argument number  string of the character argument  length of character argument		
iget_dim_arg ( iarg, iext	iarg4 > 0: gets dimensionality and extents of an argument iarg4 = 0: gets dimensionality and extents of the	integer*4 iarg (input) integer*4 iext(9) (output) integer*4	argument number, 0 for result  extents iext(1) iext(iget_dim_arg) of argument / result  dimensionality of argument / result		
iget_len_arg ( iarg	iarg4 > 0: gets length of an argument iarg4 = 0: gets length of re- sult	iget_dim_arg (function value) integer*4 iarg (input) integer*4 iget_len_arg (function value)	argument number, 0 for result  length of argument / result		
iget_nr_arg ( ) iget_type_arg ( iarg	gets number of arguments of the current operator iarg4 > 0: gets data type of an argument	integer*4 iget_nr_arg (function value) integer*4 iarg (input)	number of arguments argument number, 0 for result		
)	iarg4 = 0: gets data type of result	integer*4 iget_type_arg (function value)	type of argument / result = -1 byte = 4 float = -2 short = 8 double = -4 int		
iget_co_chk_ modus ( )	gets level of coor- dinate check for arguments ac- cording to <model>.cfg</model>	integer*4 iget_co_chk_ modus (function value)	level of coordinate check for arguments = 0 without = 1 weak = 2 strong		
iget_co_arg ( iarg, ico_blk, ico_beg )	gets coordinate block numbers and coordinate begin numbers of an argument	integer*4 iarg (input) integer*4 ico_blk(9) (output) integer*4 ico_beg(9) (output) integer*4 iget_co_arg (function value)	block number of the coordinate ico_blk(1) ico_blk(idimens) begin numbers of the coordinate ico_beg(1) ico_beg(idimens)  return code = 0 ok		

Function name	Function description	Inputs / outputs / function value	Inputs / outputs / function value description
iget_co_val ( ico_blk,	gets coordinate value at a position from a coordinate	integer*4 ico_blk (input)	block number of the coordinate
ico_pos, co_val )		integer*4 ico_pos (input)	position of the value to get within all coordinate values
		real*4 co_val (output)	coordinate value
		integer*4 iget_co_arg (function value)	return code = 0 ok = 1 ico_pos out of range = 2 storage exceeded
ichk_2args ( iarg1,	checks two arguments on same dimensionality,	integer*4 iarg1 (input)	argument number
	extents and coor- dinates	integer*4 iarg2 (input)	argument number
		integer*4 ichk_2args (function value)	return code = 0 ok = 1 differing dimensionalities = 2 differing extents = 3 differing coordinates according to
<pre>iput_struct_res ( inplace, idimens {, iext,</pre>	puts potential in- place-storage, dimensionality, extents, coordinate bock and begin numbers of the	integer*4 inplace (input)	potential inplace-indicator for result. result can be computed in-place with the following non-character arguments = -1 all = 0 none > 0 e.g. = 135 with args 1, 3 or 5
ico_blk, ico_beg }	result Currently, only	integer*4 idimens (input)	dimensionality of the result
)	coordinates from the arguments can be assigned to the result.  Apply only in the declarative part.	integer*4 iext(9) (input)	only for idimens > 0: extents iext(1) iext(idimens) of the result
		integer*4 ico_blk(9) (input)	only for idimens > 0: coordinate block numbers ico_blk(1) ico_blk(idimens) of the result
		integer*4 ico_beg(9) (input) integer*4	only for idimens > 0: coordinate begin numbers in block ico_blk ico_beg(1) ico_beg(idimens) of the result return code
		iput_dim_res (function value)	= 0 ok ≠ 0 inconsistency between operands

 Tab.
 8.16
 Operator functions to get and put structural information

All of these functions return -999 as an error indicator if the argument iarg is undefined.



Function name	Function description	Inputs / outputs / function value	Inputs / outputs / function value description					
Function	Functions to get and check argument values and put results in computational part							
arg1 ( index	gets value of a non-character argument with index index	integer*4 index (input)	vector index of an argument					
arg9 ( index		real*4 arg1 arg9 (function value)	value of an argument arguments of any type are transferred to float representation					
clip_undef ( value	a real*8 value on an undefined	real*8 value (input)	value to be checked					
real*4 result underflow: sets a real*8 value to zero if appropriate	real*4 clip_undef (function value)	clipped value normally identified with a result res e.g., res(i)=clip_undef(value8)						
is_undef ( value	checks whether value is undefined before processing	real*4 value (input)	argument to be checked					
)	it	integer*4 is_undef (function value)	= 0 value is defined = 1 value is undefined					
set_undef ( )	sets a result to be undefined	real*4 set_undef (function value)	normally identified with a result res e.g., res(i)=set_undef()					

Tab.8.17Operator function to get / check / put arguments and results

n-dimensional matrices are forwarded to user-defined operators as one-dimensional vectors, using the Fortran column-wise storage model (see Glossary):

Matrices are stored column-wise to the vector, starting with the highest dimension.

In Example 15.7 at page 131 implementation of the user-defined operator mat\_mul is described in detail.

### 8.6.2 Operator Definition File <model>.odf

<model>.odf is an ASCII file that follows the coding rules in chapter 11 at page 99 with the keywords, names, sub-keywords, and info as in Tab. 8.18. <model>.odf describes the user-defined operators.

<keyword></keyword>	<name></name>	<sub- keyword&gt;</sub- 	Line type	Max. line nmb.	<info></info>	Explanation
odf	<nil></nil>	descr	0	any	<string></string>	general operator descriptions
operator	operator_	descr	0	1	<string></string>	operator description
	name	nr_args	m	1	<integer_value></integer_value>	number of arguments defined for the operator operator_name 0< <integer_value> &lt; 10</integer_value>
		nr_charargs	m	1	<integer_value></integer_value>	from <integer_val>: number of character arguments defined for the operator operator_name 0 ≤ <integer_value> ≤ nr_args</integer_value></integer_val>

 Tab.
 8.18
 Elements of an operator description file <model>.odf

To Tab. 8.18 the following additional rules and explanations apply:

• For the description of **line type** check Tab. 11.3 at page 100.

odf odf		descr descr	Operator description for the examples in the SimEnv User's Guide
_	char test char_test char_test		test character arguments 3 2
_	corr coeff corr_coeff corr_coeff	nr_args	correlation coefficient 2 0
-	div div div	descr nr_args nr_char_args	division 2 0
	<pre>simple_div simple_div simple_div</pre>		division without special cases 2 0
operator operator operator	<pre>mat_mul mat_mul mat_mul</pre>	descr nr_args nr_char_args	<pre>matrix multiplication 2 0</pre>
			Example files: world_[f c cpp py sh].odf

**Example 8.7** User-defined operator description file <model>.odf

### 8.6.3 Handling Undefined Results

In user-defined operators

- Check always whether an argument value val is undefined by is\_undef(val) before it is processed.
- Set a result to be undefined by the function set\_undef()
   Check usr\_opr\_mat\_mul.f in Example 15.7 or usr\_opr\_div.f in the examples directory for a more detailed example.
- If things go so wrong that processing of the whole expression has to be stopped alternatively it is possible to
  - Set all elements of the results to be undefined
  - Set icompute user def operator ≠ 0 (otherwise set it always = 0)
  - In both cases processing of the next operators will be suppressed and consequently processing of the expression will be stopped
  - Check usr opr char test.f for a detailed example

### 8.7 Undefined Results

- By performing operator chains and due to possibly unwritten model output during simulation parts of the intermediate and/or final result values can be undefined within the float data representation.
- If an operand is completely undefined the computation of the result is stopped without evaluating the following operands and operators.
- For nodata value representation check Tab. 10.11.

### 8.8 Macros and Macro Definition File <model>.mac

- A macro in model output post-processing is an abbreviation for an expression, consisting of operator chains applied on operands.
- Generally, they are model related and they are defined by the user.
- Macros are identified in an post-processing expression by the suffix m.
- A macro is plugged into an expression by putting it into parentheses during parsing:

```
equ_100yrs_m*test_mac_m from Example 8.8 below is identical to (avg(atmo(c=20:-20,*,c=1,c=11:20))-400)*(1+(2+3)*4)
```

- Macros must not contain macros.
- Use simenv.chk to check macros. During the macro check validity of the following information is not checked:
  - Un-pre-defined character arguments of built-in operators (check Tab. 15.8)
  - Constant arguments of built-in operators (check Tab. 15.9)
  - Character arguments of user-defined operators
  - Operators with respect to dimensionality and dimensions of its operands

In SimEnv macros are hold in the file <model>.mac. <model>.mac is an ASCII file that follows the coding rules in chapter 11 at page 99 with the keywords, names, sub-keywords, and info as in Tab. 8.19. <model>.mac describes the user-defined macros.



<keyword></keyword>	<name></name>	<sub- keyword&gt;</sub- 	Line type	Max. line nmb.	<info></info>	Explanation
mac	<nil></nil>	descr	0	any	<string></string>	general macro descriptions
macro	macro_	descr	0	1	<string></string>	macro description
	name	unit	m	1	<string></string>	unit of the value of the macro
		define	m	any	<string></string>	macro definition string macro definition can be ar-
						ranged at a series of define-
						lines according to the rules in
						chapter 8.9

 Tab. 8.19
 Elements of an macro description file <model>.mac

To Tab. 8.19 the following additional rules and explanations apply:

• For the description of **line type** check Tab. 11.3 at page 100.

mac mac		descr descr	Macro definitions for the examples in the SimEnv User's Guide
macro	equ_100yrs	descr	<pre>2<sup>nd</sup> century tropical level 1 average without avg(atmo(c=20:-20,*,c=1,c=11:20))</pre>
macro	equ_100yrs	unit	
macro	equ_100yrs	define	
macro	tst	descr	test macro
macro	tst	define	1+(2+3)*
macro	tst	define	4
			Example files: world_[f c cpp py sh].mac

**Example 8.8** User-defined macro definition file <model>.mac

# 8.9 Handling the Post-Processor

- By the string "Enter an expression" the user is requested to enter an expression.
- Expressions can be continued on a new input line (continue expression) by finishing the previous input line by the operators "+", "-", "\*", "/", and/or "\*\*" or the separator "," of operands in operators.
- White spaces are filtered out from the expression input string, also from character arguments.
- The post-processing is left by entering <ret> instead of an expression.



# 9 Visual Experiment Evaluation

Experiment evaluation is based on application of visualization techniques to the output data, computed during experiment post-processing and stored in NetCDF format. Currently, a preliminary version is implemented.

Analysis and evaluation of post-processed data selected and derived from large amount of relevant model output benefits from visualization techniques. Based on metadata information of the post-processed experiment type, the applied operator chain, and the dimensionalities of the post-processor output pre-formed visualization modules are evaluated by a suitability coefficient how they can map the data in an appropriate manner.

The visualization modules offer a high degree of user support and interactivity to cope with multi-dimensional data structures. They cover among others standard techniques such as isolines, isosurfaces, direct volume rendering and a 3D difference visualization techniques (for spatial and temporal data visualization). Furthermore, approaches to navigate intuitively through large multi-dimensional data sets have been applied, including details on demand, interactive filtering and animation. Using the OpenDX visualization platform techniques have been designed and implemented, suited in the context of analysis and evaluation of simulated multi-run output functions.

Currently, visual experiment evaluation is the only SimEnv service that comes with a graphical user interface. In this user interface a help-services is implemented that should be used to gather additional information on how to select post-processed results for visualization and on visualization techniques provided by SimEnv.

# 10 General Control, Services, User Files, and Settings

In a general configuration file <model>.cfg the user controls general settings for the simulation environment. Besides simulation performance and model output post-processing SimEnv supplies a set of auxiliary services to check status of the model, to dump model and post-processor output and files and to clean a model from output files. General settings reflect case sensitivity, nodata values and other information related to SimEnv.

### 10.1 General Configuration File <model>.cfg

In the ASCII file <model>.cfg general SimEnv control variables can be declared. <model>.mdf is an ASCII file that follows the coding rules in chapter 11 at page 99 with the keywords, names, sub-keywords, and info as in Tab. 10.1.

<keyword></keyword>	<name></name>	<sub-keyword></sub-keyword>	Line type	Max. line nmb.	<info></info>	Explanation
cfg	<nil></nil>	descr	0	any	<string></string>	general configuration description
general	<nil></nil>	message_level	0	1	[ info   warning   error ]	specifies which message type to show during simenv.chk and in <model>.mlog</model>
model	<nil></nil>	out_directory	0	1	<directory></directory>	model output directory
		out_format	0	1	[ netcdf   ieee ]	model output format
		out_size_ threshold	0	1	<non_negative_ integer_value&gt;</non_negative_ 	file size threshold in kBytes for lumped model output
		out_ieee_ blocksize	0	1	<pre><positive_ integer_value=""></positive_></pre>	block size in kBytes for IEEE model output
experiment	<nil></nil>	restart_ini	0	1	[ no   yes ]	perform <model>.ini for ex- periment re-start</model>
		begin_run	0	1	<non_negative_ integer_value&gt;</non_negative_ 	begin single run number
		end_run	0	1	[ last   <non_negative_ integer_value&gt; ]</non_negative_ 	end single run number
		email	0	1	<string></string>	email notification address
postproc	<nil></nil>	out_directory	0	1	<directory></directory>	post-processing output directory
		out_format	0	1	[ netcdf   ieee   ascii ]	post-processing output format
		address_default	0	1	[ coordinate   index ]	post-processing address de- fault for model variables
		coord_check	0	1	[ strong   weak   without ]	post-processing coordinate check by operators
		opr_directory	0	1	<directory></directory>	directory the post-processors looks for user-defined operator executables
		visualization	0	1	[ yes   no ]	determine whether to continue with visualization

**Tab. 10.1** Elements of a general configuration file <model>.cfg

To Tab. 10.1 the following additional rules and explanations apply:

• For the description of line type check Tab. 11.3 at page 100.

<keyword></keyword>	<sub-keyword></sub-keyword>	<info>-default value (*)</info>	For more information see
cfg	descr	<nil></nil>	above
general	message_level	info	above
model	out_directory	./	above
	out_format	NetCDF	chapter 12
	out_size_threshold	10	below
	out_ieee_blocksize	50	below
experiment	restart_ini	no	chapter 7.2
	begin_run	0	chapter 7.1 - 7.3
	end_run	last	chapter 7.1 - 7.3
	email	<nil></nil>	chapter 7.1
postproc	out_directory	./	above
	out_format	NetCDF	chapter 12
	address_default	coordinate	chapter 8.2 and below
	coord_check	strong	chapter 8.3.1 and below
	opr_directory	./	chapter 8.6
	visualization	yes	below

**Tab. 10.2** <info> value defaults for the general configuration file (\*): in the case of absence of the appropriate sub-keyword

The following explanations yield:

 <string>, <directory>, <non\_negative\_integer\_value> and <positive\_integer\_value> are placeholder for corresponding strings.

#### • For <keyword> = general, <sub-keyword>= message\_level:

Message output during simenv.chk and to <model>.mlog is controlled by this information.

Specify info for errors and warnings and additional information

warning for errors and warnings

error for errors

output.

### For <keyword> = experiment, <sub-keyword>= [ begin\_run | end\_run ]:

With the exception of an optimization experiment SimEnv enables to perform an experiment partially by performing only an experiment slice out of the whole run ensemble (see chapter 7.3 at page 50). Therefor assign appropriate run numbers to this two descriptors. Make sure that begin and end run represent run number from the experiment (including run number 0) and that begin run ≤ end run. The string 'last' always represents the last simulation run of the whole run ensemble.

For an optimization experiment these two sub-keywords are ignored.

#### • For <keyword> = experiment, <sub-keyword>= email:

After performing an experiment an email is sent to the email address specified in <string>.

#### • For <keyword> = model, <sub-keyword> = out size threshold:

Specify here the threshold in kBytes for the sum of the size of all model output variables (according to their extents and data types) that is used to decide whether the SimEnv model output data for the whole run ensemble is stored into one file <model>.outall.[ nc | ieee ] or in single output files <model>.out<run\_char>.[ nc | ieee ].

### • For <keyword> = model, <sub-keyword> = out\_ieee\_blocksize:

IEEE compliant model output for single files is written in single records with a length of <out\_ieee\_blocksize> kBytes. If <out\_size\_threshold> is less than this value, this value is adapted to <out\_size\_threshold>.

#### For <keyword> = postproc, <sub-keyword> = address\_default:

During post-processing portions of multi-dimensional model output variables can be addressed by coordinate (c= ...) or index (i= ...) reference. A default is established here.



- For <keyword> = postproc, <sub-keyword> = coord\_check:

  During post-processing feasibility of application of an operator on its operands is checked with respect to the coordinate description of the operands. Different levels of this check are possible. A default is established here.
- For <keyword> = postproc, <sub-keyword> = visualization:
  Specifies whether to continue with model evaluation by visualization for NetCDF post-processor output.

Please keep in mind to ensure consistency of control settings in <model>.cfg across different SimEnv services. As an example you have to run experimentation, post-processing and dump with the same model output file size threshold out\_size\_threshold for binary output in <model>.cfg.

cfg cfg	descr descr	General configuration file for the examples in the SimEnv User's Guide
general	message level	info
model model model	<pre>out_directory out_format out_size_threshold</pre>	mod_out netcdf 100
experiment experiment	begin_run end_run	0 last
postproc postproc postproc postproc postproc	out_directory out_format address_default coord_check opr_directory visualization	<pre>res_out netcdf index strong ./ no</pre>

**Example 10.1** User-defined general configuration file <model>.cfg

# 10.2 Main and Auxiliary Services

The following SimEnv service commands are available from the SimEnv home directory \$SE\_HOME. Besides experiment performance and model output post-processing there are additional auxiliary SimEnv services to check input information consistency, to monitor the status of simulation experiments, to dump files of model and post-processor output and to wrap up the SimEnv workspace.

SimEnv command	Use to
	Main Services
simenv.run <model></model>	prepare and <b>run</b> an experiment
simenv.rst <model></model>	restart an experiment
simenv.res <model> { [ new   append   replace ] } {<run>}</run></model>	perform experiment <b>result</b> post-processing for run number <run> or for the whole run ensemble (<run> = -1, default).  Before entering post-processing those output files <model>.res<res_char>.[ nc   ieee   ascii ] and <model>.inf<res_char>.[ ieee   ascii ] with the highest two-digit number <res_char> are identified and new result files for <res+1> are written / the results are appended / or the result files are replaced by a new ones.</res+1></res_char></res_char></model></res_char></model></run></run>

SimEnv command	Use to
simenv.vis	perform visual post-processor output visualization for that NetCDF post-processor
<model></model>	output file with the highest two digit number <res_char> (latest, default) or with the file</res_char>
{ [ latest   <res> ] }</res>	number <res>. Visualization runs on a remote host.</res>
simenv.cpl	complete sequence of SimEnv commands
<model></model>	simenv.chk, simenv.run, simenv.res, simenv.vis
{ <run> }</run>	simenv.res is performed with input file <file> (if available) and interactively, for both</file>
{ <file> }</file>	optionally only for single run <run>.</run>
	Auxiliary Services
simenv.chk	check on model script files ( <model>.run, <model>.rst, <model>.ini, <model>.end)</model></model></model></model>
<model></model>	and check <model>.cfg</model>
	<model>.odf</model>
	<model>.mdf</model>
	<model>.edf</model>
	<model>.gdf</model>
	<model>.mac</model>
	existing model and post-processor output files
oim on v oto	generate pre-experiment output statistics get the current <b>status</b> of an
simenv.sts <model></model>	experiment that was started from a login node of a parallel machine and that is
{ <sleep> }</sleep>	running in a parallel or sequential job class of this machine
( sicepr )	To use this command login to the parallel machine
	Does only work properly for model start, but not for model restart.
	optimization experiment
	To use this command change to the working directory the experiment was started
	from.
simenv.dmp	dump SimEnv model output and post-processor output files
<model></model>	Files to dump have to match the SimEnv file name convention for model and/or post-processor output and are expected to be in the directories as stated in <model>.cfg. Model output variables and post-processor results in IEEE and/or ASCII format with a dimensionality greater than 1 are listed according to Fortran column-wise storage model for multi-dimensional fields (see Glossary).</model>
	To use this command change to the working directory the experiment and post- processing were started from.
simenv.cln	clean up model and post-processor output files
<model></model>	Deletes all model output files, post-processor output files, log-files, and auxiliary files
	of a model.
	To use this command change to the working directory the experiment and post- processing were started from.
simenv.cpy	copy all SimEnv example files <model>* from the examples directory of \$SE_HOME</model>
<model></model>	to the current directory.
	Additionally, example files of user-defined operators and for models world_[ f   c   cpp
	py   sh ]* common user defined files are copied. All files are only copied of they do not
	already exist in the current directory, this SimEnv service is started from.
simenv.hlp	acquire basic SimEnv help information for the specified topics
<topics></topics>	
simenv.key <user_name></user_name>	generate a ssh2- <b>key</b> to get password-free access to the visualization server.  Start this service at machine aix02 only one time before the first use of simenv.vis.  To get this access finally contact the SimEnv developers after running the service.

Tab. 10.3Service commands

Do not start a SimEnv service from a working directory, excepted simenv.sts, if there is a running SimEnv service that was started from this working directory.



# 10.3 User Scripts and Files

Script / file (in the current working direc- tory \$SE_WD)	Explanation	Exist status	For more information see chapter
<model>.cfg</model>	ASCII user-defined general configuration file	optional	10.1
<model>.mdf</model>	ASCII user-defined model (variables) description file	mandatory	5.1
<model>.edf</model>	ASCII user-defined experiment description file	mandatory	6.1
<model>.mac</model>	ASCII user-defined macro description file	optional	8.8
<model>.odf</model>	ASCII user-defined operator description file	optional	8.6.2
<model>.gdf</model>	ASCII user-defined GAMS model output description file	mandatory for GAMS models	5.6.2
<model>.run (*)</model>	model shell script to wrap the model executable	mandatory	7.5
<model>.rst (*)</model>	model shell script to prepare single model run restart	optional	7.5
<model>.ini (*)</model>	model shell script to prepare simulation experiment additionally to standard SimEnv preparation	optional, mandatory for Python and GAMS models	7.5
<model>.end (*)</model>	model shell script to clean up simulation experiment	optional, mandatory for GAMS models	7.5
<model>. <run_char>.err</run_char></model>	touch this file in the model, in <model>.run and/or <model>.rst as an indicator to stop the complete experiment after <model>.run has been finished for single model run <run_char></run_char></model></model></model>	optional	7.5
simenv.jcf_par	user-specific job control file to submit a job by the LoadLeveler to a parallel class	optional	7.5
simenv.jcf_seq	user-specific job control file to submit a job by the LoadLeveler to a sequential class	optional	7.5
simenv.oopt	user-specific control and option file for experiment type optimization	optional	6.5.3
<pre><opr>.opr (in the opr- directory according to <model>.cfg)</model></opr></pre>	executable for user-defined operator <opr></opr>	optional	8.6

Tab. 10.4 User scripts and files
(\*): make sure the shell script has execute permission by chmod u+x

Permanent files			
experiment preparation (all but optimization)	ASCII adjustment input file for the run ensemble derived from <model>.edf. Record no. n+1 corresponds to single run no. n.</model>		
experiment performance (optimization)	Column no. m of each record is the adjustment for experiment target no. m in the edf-file		
experiment performance (if model output of a single run ≥ out_size_threshold in <model>.cfg)</model>	model output of run number <run> to be processed by the post-processor (for experiment performance in a parallel job class at a parallel machine files <model>.out<run_char>.[ nc   ieee ] are created</run_char></model></run>		
evperiment performance	temporarily) model output of all runs		
(if model output of a single run < out_size_threshold	to be processed by the post-processor		
	ASCII minutes file of experiment performance		
ехрепшент реполнансе	(simenv.run and all simenv.rst)		
experiment performance	ASCII minutes file of model performance (simenv.run and all simenv.rst)		
experiment post-processing	output file of a post-processor session		
evperiment post processing	output structure description file of a post processor		
experiment post-processing	output structure description file of a post-processor session		
experiment performance (only for GAMS models)	sub-directory for GAMS model performance that are kept according to the keep_runs sub-keyword in <model>.gdf</model>		
experiment performance	additional ASCII minutes file of experiment per-		
(only for experiment type optimization)	formance		
experiment performance	ASCII file of cost function values.		
(only for experiment type optimization)	Record no. n+1 corresponds to single run no. n.		
Temporai (do not delete during exp			
all services	structured ASCII representation of <model>.cfg</model>		
experiment preparation, experiment post-processing	structured binary representation of <model>.[ mdf   edf   odf   mac ]</model>		
experiment performance (only for performance in a	ASCII file with the job-id. Used for performance of the command simenv.sts		
	experiment performance (optimization) experiment performance (if model output of a single run > out_size_threshold in <model>.cfg)  experiment performance (if model output of a single run &lt; out_size_threshold in <model>.cfg) experiment performance  experiment performance  experiment performance  experiment post-processing  experiment performance (only for GAMS models)  experiment performance (only for experiment type optimization) experiment performance (only for experiment type optimization)  Temporar (do not delete during experiment post-processing  experiment preparation, experiment post-processing  experiment post-processing</model></model>		



File / location	Generated in	Explanation
asa_opt, asa_out, asa_usr_out \$SE_WD	experiment performance (only for experiment type optimization)	auxiliary files for experiment type optimization
		and dispersions for CANAC model newform and that
run <run_char></run_char>	experiment performance (only for GAMS models)	sub-directory for GAMS model performance that are not kept according to the keep runs sub-
\$SE WD	(only for GAMO models)	keyword in <model>.gdf</model>
<model></model>	experiment performance	auxiliary files
[ pre   main   post ].inc	(only for GAMS models)	<model =="" all="" and="" gams="" main="" names<="" sub-model="" td=""></model>
\$SE_WD		
simenv_*.tmp	different services	auxiliary files
\$SE_WD		

Tab. 10.5User files generated during SimEnv performanceFor \$SE\_WD see Tab. 10.12.

Fig. 10.1 sketches usage of main SimEnv user scripts and files in the course of model interfacing, experiment preparation and performance, post-processing, and evaluation.

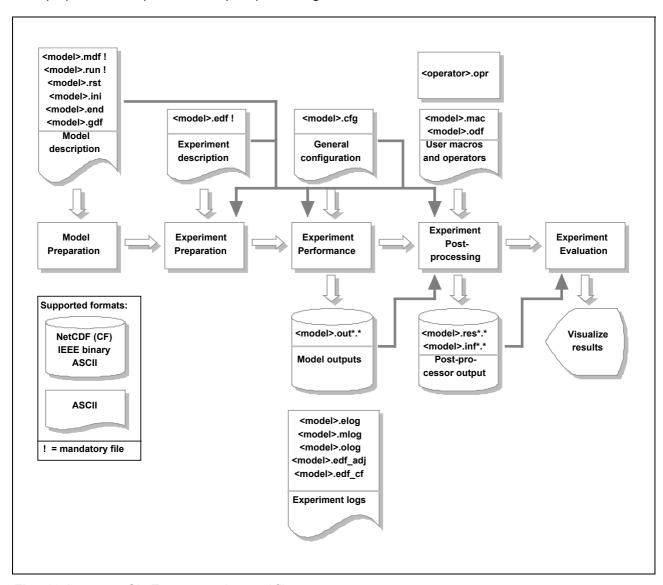


Fig. 10.1 SimEnv user scripts and files

### 10.4 Built-in Items, Reserved Names

Tab. 10.6 lists the built-in (pre-defined) model variables that are generally output during experiment performance to SimEnv model output structures and are available in model output post-processing without defining them in the model output description file <model>.mdf.

Model variable name	Dimen- sionality	Extents	Data type	Meaning
sim_time	0		float	elapsed simulation time in seconds (rounded
				to 2 decimal places) per single run for <model>.run</model>

Tab. 10.6Built-in model variables

Tab. 10.7 lists the built-in (pre-defined) coordinates that are used in model output post-processing when additional dimensions are generated by an operator.

Coordinate name	Generated by operator	Meaning	Definition (check Tab. 11.4)
bin	hgr, hgr_l, hgr_e	bin number	
index	minprop, maxprop, minprop_I, maxprop_I	index number	
run	ens	run number	equdist_end 1(1)999999
stat_measure	stat	basic statistical measures	equalst_end 1(1)999999
target_sequ	lin_abs, lin_rel, sens_abs, sens_rel, sym_abs, sym_rel	target sequence:  1 <sup>st</sup> target in edf-file = 1,  2 <sup>nd</sup> target in edf-file = 2,	
sign	sens_abs, sens_rel	sign of incremental change: $+\varepsilon = +1$ , $-\varepsilon = -1$	equidist_end -1(2)1
<target_name></target_name>	behav	target values	
incr	lin_abs, lin_rel, sens_abs, sens_rel, sym_abs, sym_rel	increment values	dependent on operator arguments

Tab. 10.7 Built-in coordinates

Tab. 10.8 lists the built-in (pre-defined) shell script variables that are used in \$SE\_HOME/simenv\_\*\_sh and finally in <model>.run.

Shell script variable name	Meaning
run_int	current run number as integer
run_char	current run number as character string
target_name	target name for simenv_get_sh
target_def_val	default target value for simenv_get_sh

Tab. 10.8Built-in shell script variables in \$SE\_HOME/simenv\_\*\_sh

Tab. 10.9 lists the reserved (forbidden) names and file names that can not be declared in user-defined files.

Element	Reserved (forbidden) names
model name <model></model>	simenv
<name> in user-defined files model.[ mdf   edf   odf   mac ] excepted for GAMS model source code file names (check chapter 11)</name>	first character must be a letter or a digit built-in model variables according to Tab. 10.6 built-in coordinates according to Tab. 10.7 built-in shell script variables according to Tab. 10.8
	built-in shell environment variables according to Tab. 10.12 special keywords in <model>.edf for behavioural analysis: [ default   file ]</model>
<file_name> in <info> in user-defined files model.[ mdf   edf   odf   mac ] (check chapter 11)</info></file_name>	SimEnv file names according to Tab. 10.4 and Tab. 10.5

**Tab. 10.9** Reserved names and file names in user-defined files and for models

# 10.5 Case Sensitivity

As stated in Tab. 10.10 all names used in SimEnv are case insensitive. Internally, they are mapped on a lowercase representation and this lowercase representation is used also for model and/or post-processor output files in NetCDF, IEEE and/or ASCII format.

Where?	Entity	Case sensitivity	Example
user-defined files (see Tab. 11.1)	<ul> <li>keyword</li> <li>name         exception:         GAMS model file name in         <model>.gdf</model></li> <li>sub-keyword</li> </ul>	insensitive	experiment END_RUN last
	<ul> <li>information <info> exceptions:</info></li> <li>any file name: for <sub-keyword> = <string>_directory and for <info> starting with string "file"</info></string></sub-keyword></li> <li><info> for <sub-keyword> = [ descr   unit ])</sub-keyword></info></li> </ul>	insensitive	experiment end_run LAST cfg descr This is specific comb file AbC.d
model interface	variable and target name	insensitive	<pre>call simenv_put_f('ATMO',atmo)  target_name='P1' target_value=1 \$SE_HOME/simnev_get_sh</pre>

Where?	Entity	Case sensitivity	Example
post-processing	<ul> <li>variable and target name</li> <li>operator name</li> <li>number</li> <li>macro name</li> <li>macro identifier _m</li> </ul>	insensitive	exp(atmo) + 3*EXP(ATMO)
	<ul> <li>character arguments         of built-in operators         with pre-defined values         (check Tab. 15.8)</li> </ul>	insensitive	count('ALL' , atmo)
	character arguments     of built-in operators     without pre-defined values	check Tab. 15.8	<pre>get_table_fct('MyFile.dat' ,</pre>
	character arguments     of user-defined operators	sensitive	char_test('arg11' , 'Arg21' , atmo)

Tab. 10.10Case sensitivity of SimEnv entities

# 10.6 Nodata Representation

For model output with the SimEnv model coupling interface functions and for post-processor output the following data type specific nodata values are used to represent undefined (unwritten) model output or undefined post-processor output:

Data type	Nodata value
byte	127
short	32767
int	2147483648
float	3.4E+38
double	1.79D+308

Tab. 10.11Data type related nodata values

# 10.7 Environment Variables

The following operating system environment variables are used by SimEnv:

Environment variable	Meaning	Explanation
SE_HOME	SimEnv home directory	has to be defined by the user  Value = [/usr/local/simenv/bin
SE_RUN	run number of a single run	defined automatically in <model>.run Value = <run_int></run_int></model>
SE_1STRUN	first single run of an experiment	defined automatically in <model>.run Value = [ yes   no ]</model>
SE_WD	current SimEnv working directory	defined automatically within any SimEnv service Value = <path></path>
PYTHONPATH	path to search Python and Python files	has to be defined by the user.  Value = machine dependent and
(only for interfacing Python and GAMS models)		has to be expanded by \$SE_HOME Has to be included in the file \$HOME/.profile for experiment submission to the LoadLeveler.

Tab. 10.12Environment variables



### 11 Structure of User-Defined Files

Basic information to describe general control settings of SimEnv, model output variables, the experiment itself, macros and user-defined operators as well as GAMS model specific information is stored in user-defined files. They are ASCII files and have a common structure that is described in this chapter.

### 11.1 General Structure

All user-defined files listed in Tab. 11.1 have the same structure. They are ASCII-files with the following record structure:

{ <sep> } <keyword> <sep> { <name> <sep> } <sub-keyword> <sep> <info> { <sep> }

with

<name> is the name of a

model variable

GAMS model source file

experiment target

coordinate

user-defined operator or

macro

Declaration of <name> depends on the related keyword <keyword> <name> is case insensitive, excepted for the GAMS model source file

<keyword> is a string

Normally, more than one line with differing sub-keywords belong

to one "keyword-block".

<keyword> is case insensitive

<sub-keyword> is a string

Sub-keywords are defined only in relation to the user file and the keyword

under consideration.

<sub-keyword> is case insensitive

<info> = <substring> { <sep> <substring> ... }

is a string with user file, keyword and sub-keyword related information.

<info> is case insensitive with the exception of any file name and/or directory and

information for sub-keywords = [ descr | unit ]

<sep> is a sequence of white spaces

Lines consisting only from separator characters as well as lines starting with a # as the first non-separator character are handled as comment lines. For case sensitivity of all information <info> in user files check Tab. 10.10 at page 97.

File	Contents	See description	
10	Contonto	in chapter	at page
<model>.cfg</model>	general configuration file	10.1	87
<model>.mdf</model>	model output description file	5.1	19
<model>.gdf</model>	GAMS description file	5.6.2	31
<model>.edf</model>	experiment description file	6.1	35
<model>.odf</model>	operator description file	8.6.2	82
<model>.mac</model>	macro description file	8.8	83
arbitrary file name	coordinate transformation file	11.3	102

Tab. 11.1 User-defined files

Element	Constraints
line length	max. 160 characters
<name></name>	max. 20 characters
	(*) first character has to be a letter
	(*) must not end on _m
	(*) must not contain elemental operators and characters . and :
	(check Tab. 8.2 at page 59)
	for further constraints: check Tab. 10.9 at page 96
<info></info>	for sub-keyword = descr without <name>: max. 512 characters</name>
	(total sum over all lines)
	for sub-keyword = descr with <name>: max. 128 characters</name>
	for sub-keyword = <string>_directory: max. 70 characters</string>
	must not contain environmental variables
	for sub-keyword = unit: max. 32 characters
	for further constraints: check Tab. 10.9 at page 96

Tab. 11.2 Constraints in user-defined files
(\*): excepted for GAMS model source code file names

The **line type** in a description table for a user-defined file specifies whether a keyword / sub-keyword combination can be omitted.

Abbre- viation	User file		Explanation
m	all files	<b>m</b> andatory	
0	all files	<b>o</b> ptional	
c1	<model>.mdf</model>	conditional 1:	
	keyword = variable	forbidden	for variables with dimensionality = 0
	sub-keyword = [ coords	mandatory	for variables with dimensionality > 0
	var_extents ]		
c2	<model>.mdf</model>	<b>c</b> onditional 2:	
	keyword = variable	forbidden	for variables with dimensionality = 0
	sub-keyword = coord_extents	optional	for variables with dimensionality > 0
c3	<model>.edf</model>	<b>c</b> onditional 3:	
	keyword = target	mandatory	for experiment type = Monte Carlo analysis
	sub-keyword = adjusts	forbidden	for experiment type = local sensitivity analysis
		conditional	for experiment type = behavioural analysis
c4	<model>.edf</model>	<b>c</b> onditional 4:	
	for Monte Carlo analysis	mandatory	for adjusts = distr
	keyword = target	forbidden	for adjusts = file
	sub-keyword = sampling		
а	<model>.edf</model>	alternatively:	
	for behavioural analysis	either	mandatory for all experiment targets
	keyword = target	or	forbidden for all experiment targets
	sub-keyword = adjusts		
f	<model>.edf</model>	<b>f</b> orbidden	
	for local sensitivity analysis		
	keyword = target		
	sub-keyword = adjusts		

Tab. 11.3Line types in user-defined files

mac mac		descr descr	This is a macro description file for the SimEnv User's Guide
macro macro macro	<pre>pol_atmo pol_atmo pol atmo</pre>	descr unit define	<pre>atmo outside polar reg., final time, level 1 without atmo(c=84:-56,*,c=1,c=20)</pre>
macro	m1	define	avg(atmo_g(c=11:20))

**Example 11.1** Structure of a user-defined file

Sequence of keyword and sub-keyword lines can be arbitrary. For reasons of readability it is recommended to use a block structure like in the above example. Sequence of names in the separated name spaces (name spaces of coordinates, model variables, experiment targets, user-defined operators, macros) during processing is determined by the sequence the name occur the first time in the appropriate user file.

#### 11.2 Value Lists

For variables, coordinates and experiment targets value lists are supplied by the <info>-item. Value lists describe a sequence of values together with an order. The number of described values is greater than 1. Value lists may be restricted to strictly monotonic sequences. They follow the syntax rules in Tab. 11.4.

Value-list type	Syntax		Explana	ation
explicit	list	<value<sub>1&gt; , , <value<sub>n&gt;</value<sub></value<sub>	explicit list of value same syntax rules record of a file with (see below)	as for one
by reference	file	{ <path>/}<file_name></file_name></path>	file { <path>/}<file_contains explic<="" td="" the=""><td></td></file_contains></path>	
implicit equidist_end <beg_val> with end-element</beg_val>		 <beg_val> (<incr_val>) <end_val></end_val></incr_val></beg_val>	description of an e	equidistant list
			begin value	
			increment end value	<incr_val> <end val=""></end></incr_val>
			<pre>cha valdo <beg_val> ≠ <end< pre=""></end<></beg_val></pre>	_
			<incr_val> ≠ 0.</incr_val>	
implicit	equidist_nmb	<beg_val> (<incr_val>) <nmb_vals></nmb_vals></incr_val></beg_val>	description of an e	quidistant list
with number			of values with	
of values			begin value	<b>~</b>
			increment	_
			number of values	_
			   <beg_val></beg_val>	nd_val>
			<incr_val> ≠ 0.</incr_val>	
			<nmb_vals> &gt; 0, ii</nmb_vals>	nteger

**Tab. 11.4** Syntax rules for value lists

#### Syntax rules for a file {<path>/}<file\_name> with a list of values

- <path> must not contain environment variables from operating system level
- If <path> is specified in a relative manner it relates to the current working directory, the / a SimEnv service where <path> is used was started from.
- Has to be an ASCII file
- May be a multi-record file
- Max. record length: 1000 characters



- Values are separated from each other by white spaces or comma
- A series of connected (running) separators is treated as a single separator
- Record end is handled as a separator
- Real values can be stated in integer, real or exponential (scientific) format
- Records formed only from white spaces or records starting with first non-white space character # are handled as comments

```
1. list 3, 5, 7, 9, 11 describes the five values 3, 5, 7, 9, 11
2. equisist_end 3 (2) 11 is equivalent to 1.
3. equidist_nmb 3 (2) 5 is equivalent to 1.
4. file my_values.dat is equivalent to 1. with my_values.dat = 3, , 5, 7 9,11
5. equidist_end 3 (2) 11.9 is equivalent to 1.
6. equidist_end 11 (-2) 3 differs from 1. - 4.: values are identical, ordering sequence differs
```

**Example 11.2** Examples of value lists

### 11.3 Coordinate Transformation File

Some operators (currently, get\_experiment) enable access to external data. Most of these operators supply in general a multi-dimensional result that has to be equipped - as usual in SimEnv post-processing - with a coordinate assignment. By applying these operators it can be necessary to transform the coordinate description from the external data to fit the data to the current model and/or experiment under consideration. The following cases can be distinguished:

- A dimension of the external data does not have a coordinate assignment. A coordinate has to be assigned to this dimension.
- A coordinate description of the external data has to be modified in a way that it matches with a defined coordinate of the model / experiment under consideration.
- A coordinate description of the external data has to be incorporated with and/or without modifications into the coordinate set of the model / experiment under consideration.

Coordinate transformations for external data are supported by a coordinate transformation file that is assigned to the operator result as an argument of the operator. Coordinate transformation files follow the same syntax rules as all other user-defined files (see chapter 11.1).



<keyword></keyword>	<name></name>	<sub- keyword&gt;</sub- 	Line type	Max. line nmb.	<info></info>	Explanation
trf	<nil></nil>	descr	0	any	<string></string>	general transformation description
modify	external_	rename	0	1	new_name	renames external coordinate
	coordinate_ name	position_shift	0	1	position_shift_val	shifts all values of the external coordinate by the specified real value position_shift_val
		values_shift	0	1	values_shift_val	shifts the result values on the coordinate by integer values_shift_val positions
		values_add	0	1	<value_list></value_list>	defines values_shift_val val- ues to added to the coordi- nate values
assign	external_ coordinate_ name	coord	0	1	coord_name	assign to the dimension with external_coordinate_name an internal coordinate or a coordinate defined by the keyword coordinate
		coord_extent	0	1	co_val <sub>1</sub> :co_val <sub>2</sub>	assigns start and end coordinate value to the dimension of the result under consideration.
coordinate	new_	descr	0	1	<string></string>	coordinate axis description
	coordinate_	unit	0	1	<string></string>	coordinate axis unit
	name	values	0	1	<value_list></value_list>	strictly monotonic sequence of coordinate values (for syntax see Tab. 11.4)

**Tab. 11.5** Elements of an coordinate transformation file

To Tab. 11.5 the following additional rules and explanations apply:

- For the description of line type check Tab. 11.3 at page 100.
- With the sub-keyword values\_shift result values can be shifted on the corresponding coordinate by values\_shift\_val coordinate values. Consequently values\_shift\_val coordinate values have to be appended at the end of the coordinate for a positive value of values\_shift\_val and/or have to be inserted at the begin of the coordinate for a negative value of values\_shift\_val. Coordinate values that are obsolete because of this shift are removed from the coordinate definition.
  - For a coordinate that is defined with equidistant coordinate values the extent of the coordinate is performed automatically, simply by applying the equidistant rule for this coordinate.
  - For a coordinate with non-equidistant coordinate values the coordinate values necessary for the coordinate extension are defined by the sub-keyword **values add**.
  - If both **position\_shift** and values\_shift are specified for one coordinate, firstly position shift is applied to the coordinate and then the additional coordinate values from values\_shift are added the the coordinate without applying the position\_shift value.
- For the sub-keyword coord\_extent the same rules apply as for the sub-keyword coord\_extents from the model output description file <model>.mdf.
- For the keyword coordinate the same rules apply as for the keyword coordinate from the model output description file <model>.mdf.

Unlike all other user-defined files coordinate transformation files can not be checked by the SimEnv service simenv.chk.

Having a model variable definition as in Example 5.1 at page 22 and assuming address\_default=coordinate in <model>.cfg
Assume the experiment layout in Example 6.1 (c) at page 38 and the corresponding experiment description file (c) from Example 6.1 at page 37.

Assume additionally result from another experiment with a model named model and there an result modvar1+modvar2 that is defined for the following coordinates:

dimension	coordinate name	coordinate definition
1	dim1	list 1,10,100,1000
2	dim2	equidist_end 2(2)20
3	dim3	equidist_end 3(3)30
4	dim4	equidist_end 4(1)43
5	dim5	equidist end 5(1)50

#### Further, assume the coordinate transformation file model.trf as

trf trf		descr descr	example for a coordinate transformation file
modify modify modify modify	dim1 dim1 dim1	rename position_shift values_shift values_add	new1 3. +2 list 1006, 1009
modify	dim3	values_shift	-3
assign assign	dim4 dim4	coord coord_extent	lat 88.:-68.
assign assign	dim5 dim5	coord extent	new2 50.:5.
coordinate coordinate	new2 new2	descr values	<pre>new coordinate equidist_end 50(-1)5</pre>

## In SimEnv post-processing the result of the expression

get experiment('mydir','model','model.trf',modvar1+modvar2)

#### is a 5-dimensional data structure with

dimension	coordinate	coordinate	coordinate
	name	definition	use
1	new1	list 103,1003,1006,1009	list 103,1003,1006,1009
2	dim2	equidist_end 2(2)20	equidist_end 2(2)20
3	dim3	equidist_end 0(3)27	equidist_end 0(3)27
4	lat	equidist_end 88(-4)-88	equidist_end 88(-4)-68
5	new2	equidist_end 5(1)50	equidist_end 5(1)50

Coordinates new1 and new2 are defined only for the current expression.

**Example 11.3** Coordinate transformations by a transformation file



# 12 Model and Post-Processor Output Data Structures

This chapter summarizes information on available data structures for model and post-processor output. SimEnv supports several output formats from the experiment and the post-processor. NetCDF is a self-describing data format and can be used for model and post-processor output. Another format specifications for both outputs is IEEE compliant binary format and ASCII for post-processor output. This chapter describes all the used data structures.

Dependent on the specification of the supported post-processor output formats in <model>.cfg model output can be stored in NetCDF format and post-processor output in NetCDF, IEEE or ASCII format.

During experiment performance model output is written either to single output files <model>.out<run\_char>. [nc | ascii ] per experiment single run or to a common output file <model>.outall.[nc | ieee ] for all single runs from the experiment run ensemble. Output to single or a common file(s) depends on specification of the value for the out\_size\_threshold sub-keyword in <model>.cfg. <run\_char> is a six-digit placeholder for the corresponding single run number.

During model output post-processing output and structure of results is written to <model>.res<res\_char>.[ nc | ieee | ascii ] and <model>.res<res\_char>.[ ieee | ascii ]. <res\_char> is a two-digit placeholder for the number of the result file. It ranges from 01 to 99.

For IEEE and ASCII model output and post-processor output formats, multi-dimensional data is organized in the Fortran column-wise storage model (see Glossary).

# 12.1 NetCDF Model and Post-Processor Output

The intention for supplying NetCDF format for model and post-processor output is to provide the possibility to generate self-describing, platform-independent data files with metadata that can be interpreted by subsequent visualization techniques. The conventions applied for SimEnv represent a compromise between existing standards and the metadata requirements for a flexible and expressive visualization that is adapted to the requirements of the specific data sets of concern. SimEnv follows the NetCDF Climate and Forecast (NetCDF CF) metadata convention 1.0-beta4. Currently, SimEnv supports only up to 4-dimensional NetCDF output during experiment and post-processor performance.

Model output data types as declared in the model output description file <model>.mdf are transferred into NetCDF data types automatically (check the Table below). By default, post-processor output data is of type float

SimEnv data type (see Tab. 5.2)	NetCDF data type
byte	NF_BYTE
short	NF_SHORT
int	NF_INT
float	NF_FLOAT
double	NF_DOUBLE

Tab. 12.1 NetCDF data types

#### 12.1.1 Global Attributes

The global attributes used in SimEnv from the CF standard are :institution and :convention. In addition, the following global attributes are defined for model and post-processor output:

Name	Value	Data type
:creation_time	<yyyy-mm-dd hh:mm:ss=""></yyyy-mm-dd>	char
:model_name	<model></model>	char
:model_description	model description according to <model>.mdf</model>	char
:model_description_file	{ <path>/}<model>.mdf</model></path>	char
:experiment_type	[ behaviour   monte carlo   sensitivity ]	char
:experiment_description	experiment description according to <model>.edf</model>	char
:experiment_description_file	{ <path>/}<model>.edf</model></path>	char
:number_of_runs	<number of="" runs=""></number>	int

Tab. 12.2 Additional global NetCDF attributes

## 12.1.2 Variable Labelling and Variable Attributes

For coordinate variables, two cases of labelling are distinguished:

- If for a given predefined variable, target, model variable or post-processor result one of its coordinates spans the entire range of its general dimension, the already existing coordinate definition is used.
- Otherwise, this concerned coordinate is re-defined using the notation <variable\_name>\_dim\_<coordinate\_name>.

The following variable attributes are used according to the CF 1.0-beta4 standard:

Name	Value	Data type
<variable_name>:standard_name</variable_name>	[ <coordinate_name>  </coordinate_name>	char
	<pre><pre>cpredef_coordinate_name</pre></pre>	
	<pre><predef_var_name>  </predef_var_name></pre>	
	<target_name>  </target_name>	
	<variable_name>  </variable_name>	
	<result_name> ]</result_name>	
<variable_name>:long_name</variable_name>	[ <coordinate_description>  </coordinate_description>	char
	<pre><pre><pre>cpredef_coordinate_description&gt;  </pre></pre></pre>	
	<pre><pre><pre><pre>oredef_variable_description&gt;  </pre></pre></pre></pre>	
	<target_description>  </target_description>	
	<variable_description>  </variable_description>	
	<result_applied_operator_sequence> ]</result_applied_operator_sequence>	
<variable_name>:missing_value</variable_name>	<variable missing="" type-depending="" value=""></variable>	type-dep.
<variable_name>:axis</variable_name>	[X Y Z T bin run ]	char
(single coordinate variables only)		
<variable_name>:unit</variable_name>	[ <coordinate_unit>  </coordinate_unit>	char
	<pre><pre><pre>cpredef_coordinate_unit&gt;  </pre></pre></pre>	
	<pre><predef_variable_unit>  </predef_variable_unit></pre>	
	<target_unit>  </target_unit>	
	<variable_unit>  </variable_unit>	
	<result_unit>]</result_unit>	
<variable_name>:coordinates</variable_name>	<par1_lon> <par1_lat></par1_lat></par1_lon>	char
(multi-dimensional coordinate		
variables only)		
<variable_name>:fill_value</variable_name>	<variable fill="" type-depending="" value=""></variable>	type-dep.

Tab. 12.3Variable NetCDF attributes

- For post-processor output, the :standard\_name attribute simply counts the number of applied operations because the result name of an arbitrary operation is not known in general. For that reason, the :long\_name attribute would re-sample the :standard\_name attribute and it is used instead to provide the complete description of the applied operator sequence without defining an additional attribute. If macros are included, these are resolved and elementary operations are included only.
- For the :axis attribute of a coordinate variable exist defaults.

  For each post-processor result, the first coordinate is assumed to be the "X-axis", the second and third coordinate are assumed to represent the "Y-" and "Z-axis", and the fourth dimension is time T.

  For model results, these attribute values are assigned to coordinate variables describing geographical longitude, geographical latitude, level or height and time. In case other coordinate names are used, these are simply also used for the axis attribute.
- The :unit attribute is actually estimated for model output only depending on the description of the corresponding variable keywords in the <model>.mdf file. For post-processing output, it is only used as a placeholder and not calculated from the applied operator sequence so far.
- The :coordinates attribute serves to define coordinates depending on other ones and so to allow coordinate transformations. Actually, this attribute is not used.
- Actually, the **:fill\_value attribute** is not applied to coordinate variables. It is identically to the :missing\_value attribute but open for other definitions.

For visualization requirements, the following additional variable attributes have been defined for SimEnv:

Name	Value	Data type
<pre><variable_name>:monotony (coordinate variables only)</variable_name></pre>	[ increasing   decreasing   none ]	char
<variable_name>:coo_type</variable_name>	[1 2]	integer
<variable_name>:data_range</variable_name>	<min> <max></max></min>	char
<pre><variable_name>:index_range_<coordinate> (coordinate variables only)</coordinate></variable_name></pre>	<min_index> <max_index></max_index></min_index>	int
<variable_name>:simenv_data_kind</variable_name>	[ predefined model variable   model target   model output variable   postproc_result ]	char
<variable_name>:var_representation</variable_name>	[ positions   connections ] or both	char
<variable_name>:grid_shift</variable_name>	<shift_x> <shift_y></shift_y></shift_x>	real, dimension(2)
<variable_name>:north_pole</variable_name>	<lon_pole> <lat_pole></lat_pole></lon_pole>	real, dimension(2)

 Tab. 12.4
 Variable NetCDF attributes for visualization

- The :monotony attribute is applied to coordinate variables only and estimated from the coordinate values as defined in the <model>.mdf file. During post-processing additional coordinates can be generated for which no monotony may be estimated. In such cases, the attribute is set to "none".
- The :coo\_type attribute describes the grid representation of a given coordinate. A value of 1 indicates that all coordinate values are provided explicitly (suitable, e.g., for irregular grids). A value of 2 indicates a regular grid and a coordinate representation by its start value, increment and end value.
- The :data\_range attribute provides the real range that is covered by the related variable in the recent NetCDF file.
- The :index\_range attribute is used only in case a predefined variable, target, model variable or post-processing result covers not the complete range of a dimension as defined for a coordinate variable. It describes that sub-space for which the concerned target, variable or result is defined.
- The :var\_representation attribute is introduced to specify what operations are allowed on the data.
- The :grid\_shift attribute is actually still a placeholder for variables that are not defined in the centre of a grid box when quasi-regular grids are used.
- The :north\_pole attribute can be used if rotated grids are applied.

16-Jun-2004

# 12.2 IEEE Compliant Binary Model Output

IEEE compliant binary model output is written in records of fixed length to <model>.out<run\_char>.ieee and/or <model>.outall.ieee. Record length is determined by the sub-keyword out\_ieee\_blocksize and in interrelation to the sub-keyword out\_size\_threshold in <model>.cfg. For these two sub-keywords and potential modification of the value for out\_ieee\_blocksize check Tab. 10.1. Sequence of data for each single run is as follows:

- Experiment targets as specified in <model>.edf
   Sequence as in <model>.edf
- Built-in (pre-defined) model output variables Sequence as in Tab. 10.6
- Model output variables
   Sequence as in <model>.mdf

Storage demand for each model variable / target is according to its dimensionality, extents and data type. Storage demand in bytes for each model variable / target is readjusted to the smallest number of bytes divisible by 8, where the data can be stored. Multi-dimensional data fields are organized in the Fortran columnwise storage model (see Glossary).

In <model>.outall.ieee each single run starts with a new record. Sequence of single runs corresponds with sequence of the single run numbers <run>: Data from default single run 0 is stored in the first and potentially the following records.

Having a model output description file as in Example 5.1 and an experiment description file as in Example 6.1(a) each single run is stored in the following way:

Target / model variable	Extents	Data type	Storage demand [Byte]	Storage demand adjusted [Byte]
p1	1	float	4	8
p2	1	float	4	8
sim_time	1	float	4	8
atmo	45 x 90 x 4 x 20	float	1.296.000	1.296.000
bios	36 x 90 x 20	float	259.200	259.200
atmo_g	20	int	80	80
bios_g	1	int	4	8
				1.555.312

With out\_ieee\_blocksize = 100, which transforms to 100\*1024 =102.400 Bytes, one single run needs 16 records with a fixed length of 102.400 Bytes. Remaining bytes in the last record are undefined.

**Example 12.1** IEEE compliant model output data structure

# 12.3 IEEE Compliant Binary and ASCII Post-Processor Output

For IEEE and ASCII post-processor output result information is stored in two files:

- <model>.res<res char>.[ ieee | ascii ] holds the result dynamics
- <model>.inf<res\_char>.[ ieee | ascii ] holds structure and coordinate information

The IEEE post-processor output files <model>.res<res\_char>.ieee and <model>.inf<res\_char>.ieee are unformatted binary files with IEEE float / int number representation, while for the ASCII post-processor version <model>.res<res\_char>.ascii and <model>.inf<res\_char>.ascii formatted ASCII files are used. Files for both output file formats have for each result subsequently the following structure:



## Record structure of <model>.inf<res\_char>.[ ieee | ascii ] for each result:

record no. 1	integer*4	length of the string of the operator sequence
	max. char*512	character string of the operator sequence
record no. 2	10 integer*4	idim ext(1) ext(9)
		(ext(i) = 0  for  i > idim)
record no. 3	max. char*20	coordinate name of dimension 1
record no. 4	max. 10 float	coordinate values of dimension 1 in records of 10 values
record no. xxx	max. char*20	coordinate name of dimension idim
record no. xxx+1	max. 10 float	coordinate values of dimension idim in records of 10 values

## Record structure of <model>.res<res\_char>.[ ieee | ascii ] for each result:

Record no. 1	max. 10 float	result_values(1) result_value(length_result)	
		with length_result = product of ext(i) > 0	for idim > 0
		= 1	else

The vector result\_value is stored in the Fortran column-wise storage model (see Glossary). The nodata element for undefined result values is set to 3.4E38.

The Fortran code in Example 15.8 reads post-processing binary output files <model>.res<res\_char>.ascii and model>.inf<res\_char>.ascii in their general structure:

# 13 Prospects

SimEnv development and improvement is user-driven. Here you can find a list of the main development pathways in future.

#### General

- Graphical user interface
- Linux and Windows portability
- Unique number representations for binary output (big endians vs. small endians)

#### **Model interface**

#### **Experiment preparation**

· Experiment type stochastic analysis

#### **Experiment performance**

- Experiment performance for distributed models across networks
- Multi-file model output storage

#### **Experiment post-processing**

- Additional advanced operators (netcdf\_data, regrid, coarse, sort, ...)
- Advanced uncertainty and global sensitivity analyses operators
- C-interface for user-defined operators
- Flexible assignment of data types to operator results (currently: only float)
- Shared memory access for user-defined operators to avoid data exchange by external files
- Wrapping of pure C/C++-operators in Fortran to use them as built-in operators

#### **Experiment evaluation**

Advanced techniques for graphical representation of post-processor output, especially for multi-run operators

# 14 References

- Bohr, J. (1998) A summary on Probabilities
  - http://ic.net/~jnbohr/java/CdfDemoMain.html
- European Commission, Joint Research Centre IPSC (2004): SimLab 2.2 Reference Manual http://www.jrc.ce.eu.int/uasa/primer-sa.asp
- Flechsig, M. (1998) SPRINT-S: A Parallelization Tool for Experiments with Simulation Models. PIK-Report No. 47, Potsdam Institute for Climate Impact Research, Potsdam <a href="http://www.pik-potsdam.de/reports/pr-47/pr47.pdf">http://www.pik-potsdam.de/reports/pr-47/pr47.pdf</a>
- Flechsig, M., Böhm, U., Nocke, T., Rachimow, C. (2004): Techniques for Quality Assurance of Models in a Multi-Run Simulation Environment. 4th International Conference on Sensitivity Analysis of Model Output SAMO 2004, March 8-11, 2004, Santa Fe, U.S.A.
  - http://www.pik-potsdam.de/~flechsig/simenv/simenv\_poster\_samo2004.pdf
- Gray, P., Hart, W., Painton, L., Phillips, C., Trahan, M., Wagner, J. (1997) A Survey of Global Optimization Methods. Sandia National Laboratories, Albuquerque, U.S.A. http://www.cs.sandia.gov/opt/survey
- Helton, J.C., Davis, F.J. (2000): Sampling-Based Methods. In: Saltelli et.al (2000)
- Imam, R.L., Helton, J.C. (1998): An Investigation of Uncertainty and Sensitivity Analysis Techniques for Computer Models. Risk Anal. 8(1), 71-90
- Ingber, L. (1989): Very fast simulated re-annealing. Math. Comput. Modelling, 12(8), 967 973 http://www.ingber.com/asa89 vfsr.pdf
- Ingber, L. (1996) Adaptive simulated annealing (ASA): Lessons learned. Control and Cybernetics, 25(1), 33-54
  - http://www.ingber.com/asa96\_lessons.pdf
- Ingber, L. (2004) ASA-Readme.
  - http://www.inger.com/ASA-README.pdf
- Saltelli, A., Chan, K., Scott, E.M. (eds.) (2000) Sensitivity Analysis. J. Wiley & Sons, Chichester
- Saltelli, A., Tarantola, S., Campolongo, F., Ratto, M. (2004) Sensitivity Analysis in Practice: A Guide to Assessing Scientific Models. J. Wiley & Sons, Chichester
- Waszkewitz, J., Lenzen, P., Gillet, N. (2001) The PINGO package: Procedural interface for Grib formatted objects. Max-Planck-Institute for Meteorology, Hamburg and http://www.mad.zmaw.de/Pingo/pingohome.html
- Wenzel, V., Kücken, M., Flechsig, M. (1995) MOSES Modellierung und Simulation ökologischer Systeme. PIK-Report No. 13, Potsdam Institute for Climate Impact Research, Potsdam http://www.pik-potsdam.de/pik\_web/publications/pik\_reports/sum\_pr13.htm
- Wenzel, V., Matthäus, E., Flechsig, M. (1990) One Decade of SONCHES. Syst. Anal. Mod. and Sim. 7, 411-428
- Wierzbicki, A.P. (1984) Models and Sensitivity of Control Systems. Studies in Automation and Control. Vol. 5. Elsevier, Amsterdam

16-Jun-2004

# 15 Appendices

The appendices summarize the current version implementation, referred examples for model interfaces, user-defined operators and result import interfaces, they compile all post-processor built-in operators, and they supply a glossary of the main terms as used in this User's Guide.

16-Jun-2004

# 15.1 Version Implementation

# 15.1.1 Linking User Models and User-Defined Operators

For

- User models implemented in C/C++ or Fortran and
- User-defined operators to be used in result post-processing the following libraries have to be linked to couple them to the simulation environment
- \$SE\_HOME/libsimenv.a
- /usr/local/lib/libnetcdf.a

For running models again outside SimEnv check chapter 5.8.

## 15.1.2 Example Models and User Files

For the following models corresponding files of Tab. 10.4 of can be copied from the corresponding examples-directory of \$SE\_HOME to the user's working directory by running the SimEnv command simenv.cpy <model> from the working directory:

<model></model>	Source code	Explanation
world_f	Fortran	global atmosphere - biosphere test model
_	world_f.f	at a (lat x lon x level x time) = (45 x 90 x 4 x 20) resolution
world_c	С	global atmosphere - biosphere test model
	world_c.c	at a (lat x lon x level x time) = (45 x 90 x 4 x 20) resolution
world_cpp	C++	global atmosphere - biosphere test model
	world_cpp.cpp	at a (lat x lon x level x time) = (45 x 90 x 4 x 20) resolution
world_py	Python	global atmosphere - biosphere test model
	world_py.py	at a (lat x lon x level x time) = (45 x 90 x 4 x 20) resolution
world_sh	Shell script level	global atmosphere - biosphere test model
	world_sh.f	at a (lat x lon x level x time) = (45 x 90 x 4 x 20) resolution
	world_shput.f	
world_f_1x1	Fortran	global atmosphere - biosphere test model
	world_f_1x1.f	at a ( lat x lon x level x time ) = ( 180 x 360 x 16 x 20 ) resolution
pixel_f	Fortran	global atmosphere - biosphere test model for one lat-lon constella-
	Pixel_f.f	tion at a (level x time) = (4 x 20) resolution
gams_model	GAMS	GAMS model from Example 15.6
	gams_model.gms	

**Tab. 15.1** Implemented models for current version for <model> = world \* check also Example 1.1

Additionally, the following files are available in the corresponding examples directory of \$SE\_HOME:

File	Explanation
<model>.[ f   c   cpp   py   gms]</model>	model source code
<model></model>	model executable compiled and linked from <model>.[f c cpp]</model>
world.edf_[ a   b   c   d   e   f ]	experiment description files corresponding to Example 6.1, Example 6.2, and Example 6.3 to be copied to $world_[f c cpp py sh]$ .edf and/or $world_f_1x_1$ .edf
world.post_[ c   e   f   bas   adv ]	post-processor input file (complete experiment) for world.edf_[ c   e   f ] (simenv.res world_[ f   c   cpp   py   sh ] [ new   append   replace ]  < world.edf_[ c   e ])
	and/or all experiments (selected single run <run>)</run>
	(simenv.res world_[f c cpp py sh] [new append replace] <run></run>
world.dat_[ d   e   tab ]	data files for world.edf_[ d   e ] and/or world.post_adv
usr_opr_ <opr>.f</opr>	source code for user-defined operator <opr></opr>
<opr><opr< td=""><td>executable for user-defined operator <opr></opr></td></opr<></opr>	executable for user-defined operator <opr></opr>
model_[f c cpp].lnk <model></model>	compile <model>.[f   c   cpp ] and link to executable <model></model></model>
usr_opr_ <opr>.f</opr>	source code file for user-defined post-processing operator <opr></opr>
operator_f.lnk <opr></opr>	compile usr_opr_ <opr>-f and link executable <opr>-opr for user-defined post-processing operator <opr>-</opr></opr></opr>

**Tab. 15.2** Implemented model-related user files for current version for <opr> see Tab. 15.3 below

## 15.1.3 Example User-Defined Operators

The following user-defined operators are available from the corresponding examples directory of \$SE\_HOME:

Name	Explanation / restriction	Example
char_test('char1','char2',arg)	character test	see source code
corr_coeff(arg1,arg2)	correlation coefficient R	<pre>corr_coeff(bios,-bios) = -1.</pre>
div(arg1,arg2)	division as an example how the cor- responding built in basic operator works	div(-2,-4) = 0.5
mat_mul	matrix multiplication of 2- dimensional operands	mat_mul(mat1,mat2)
simple_div(arg1,arg2)	division without consideration of overflow, underflow, division by zero	$simple_{div(-2,-4)} = 0.5$

Tab. 15.3Available user-defined operators

# **15.1.4 Technical Limitations**

Entity		Limitation
User-defined files entities (check also chap	ter 11)	
max. length of a record in a user-defined file	[characters]	160
max. length of all global descriptions descr	[characters]	512
max. length of a local description descr	[characters]	128
max. length of a unit	[characters]	32
max. length of a directory	[characters]	70
max. length of a record of a referred data file	[characters]	1000
Model interface and experiment preparation	entities	
max. length of a name	[characters]	20
max. dimensionality of a model output variable		9
max. dimensionality of a model output variable for Python models		4
max. dimensionality of a model output variable for GAMS models		4
max. dimensionality of a model output variable stored in NetCDF for	ormat	4
max. number of model output variables in <model>.mdf</model>		50
max. number of coordinates in <model>.mdf</model>		30
max. number of experiment targets in <model>.edf</model>		50
max. number of slice definitions during interfacing a model		30
max. number of single model runs in an experiment		999.999
max. number of coordinate values and target adjustment values		200.000
Post-processing entities (per expression	n)	
max. number of arguments of an operator		9
max. dimensionality of a result stored in NetCDF format		4
max. number of post-processor output files		99
max. number of characters of an expression		512
max. number of all operands and operators of an expression		200
max. length of a constant	[characters]	20
max. number of constants		30
max. number of user-defined operators in <model>.odf</model>		45
max. number of allocatable main memory segments		10
max. allocatable main memory	[MBytes]	240

Tab. 15.4 Current SimEnv technical limitations

# 15.2 Examples for Model Interfaces, User-Defined Operators, and Result Import Interfaces

### 15.2.1 Example Implementations for Model world

According to Example 1.1 at page 4 dynamics of the model world depend on four model paramters p1, p2, p3, and p4:

Model target	Target default value	Internal model parameter	Unit	Meaning
p1	1.	phi_lat	π/12	latitudinal phase shift
p2	2.	omega_lat	2*π	latitudinal frequency
p3	3.	phi_lon	π/12	longitudinal phase shift
p4	4.	omega lon	2*π	longitudinal frequency

Tab. 15.5 Parameters for the model world

Mapping between model targets and internal model parameters is performed by simenv\_get\_\*

For reasons of simplification these parameters influence state variables atmo and bios by the product of two sine terms in the following manner:

```
value_lat(lat) = \sin(2*\pi*omega_lat*f(lat)+phi_lat*\pi/12)
value_lon(lon) = \sin(2*\pi*omega_lon*f(lon)+phi_lon*\pi/12)
```

The function f(.) maps lat and/or lon in a way, that yields

Mean avg is calculated in a box around (lat,lon) =  $(0^{\circ},0^{\circ})$ .

#### 15.2.2 Fortran Model

With respect to Example 5.1 the following Fortran code **world\_f.f** could be used to describe the model coupled to SimEnv. SimEnv modifications are marked in **bold**.

```
program world f
c declare SimEnv coupling functions
  integer*4
             simenv_ini_f, simenv_get_f, simenv_get_run_f
  integer*4
              simenv_slice_f, simenv_put_f, simenv_end_f
c declare atmo without dimensions level and time and bios without time
c because they are computed in place and simenv slice is used
  real*4 atmo(0:44,0:89)
              bios(0:35,0:89)
  real*4
  integer*4 atmo g(0:19)
  integer*4
             bios g
   integer*4
              run int
   character*6 run char
  istatus = simenv ini f()
c check return code for the model coupling functions at least here
  if(istatus.ne.0) call exit (1)
c only if necessary:
  istatus = simenv get run(run int,run char)
  p1 = 1.
  p2 = 2.
  p3 = 3.
  p4 = 4.
  istatus = simenv get f('p1',p1,p1)
  istatus = simenv get f('p2',p2,p2)
  istatus = simenv_get_f('p3',p3,p3)
  istatus = simenv get f('p4',p4,p4)
c compute dynamics of atmo and bios over space and time,
c of atmo g over time, all dependent on p1,p2,p3,p4
  do idecade = 0,19
      do level= 0,3
         istatus = simenv_slice_f('atmo',3,level,level)
         istatus = simenv slice f('atmo',4,idecade,idecade)
         istatus = simenv put f('atmo',atmo)
      istatus = simenv slice f('bios',3,idecade,idecade)
      istatus = simenv put f('bios',bios)
   enddo
   istatus = simenv put f('atmo g',atmo g)
c compute dynamics of bios g
   istatus = simenv_put_f('bios_g',bios_g)
   istatus = simenv end f()
   end
                                                          Example file:world_f.f
```

**Example 15.1** Model interface for Fortran models - model world\_f.f

#### 15.2.3 C Model

With respect to Example 5.1 the following C code **world\_c.c** could be used to describe the model coupled to SimEnv. SimEnv modifications are marked in **bold**.

```
#include <stdio.h>
#include <stdlib.h>
#include <string.h>
/* declare SimEnv coupling functions (compile with -I$SE HOME) */
#include <simenv.h>
/* declare atmo without dimensions level and time and bios without time*/
/* because they are computed in place and simenv slice is used */
static float atmo[45][90];
static float bios[36][90];
static int atmo g[20];
static int
              bios g;
main (void)
   float p1,p2,p3,p4;
   int run int;
   char run char[6];
  int level,idecade,istatus,idim;
  istatus = simenv ini c();
/* check return code of model coupling functions at least here */
  if(istatus != 0) return 1;
/* only if necessary: */
  istatus = simenv get run c(&run int,run char);
  p1 = 1.;
  p2 = 2.;
  p3 = 3.;
  p4 = 4.;
  istatus = simenv_get_c('p1',&p1,&p1);
  istatus = simenv get c('p2',&p2,&p2);
  istatus = simenv_get_c('p3',&p3,&p3);
  istatus = simenv_get_c('p4',&p4,&p4);
/* compute dynamics of atmo and bios over space and time, */
/* of atmo g over time, all dependent on p1,p2,p3,p4 */
   for (idecade=0; idecade<=19; idecade++)</pre>
   {...
     for (level=0; level<=3; level++)</pre>
       idim=3;
       istatus = simenv slice c('atmo',&idim,&level,&level);
       istatus = simenv slice c('atmo',&idim,&idecade,&idecade);
       istatus = simenv_put_c('atmo',(char *) &atmo);
     idim=3;
     istatus = simenv slice c('bios',&idim,&idecade,&idecade);
     istatus = simenv put c('bios',(char *) &bios);
   istatus = simenv put c('atmo g',(char *) &atmo g);
```

```
/* compute dynamics of bios_g */
    ...
    istatus = simenv_put_c('bios_g', ,(char *) &bios_g);
    istatus = simenv_end_c();
    return 0;
}
Example file: world_c.c
```

**Example 15.2** Model interface for C models – model world\_c.c

#### 15.2.4 C++ Model

With respect to Example 5.1 the following C++ code **world\_cpp.cpp** could be used to describe the model coupled to SimEnv. SimEnv modifications are marked in **bold**.

```
#include <stdio.h>
#include <stdlib.h>
/* declare SimEnv coupling functions (compile with -I$SE HOME) */
#include <simenv.h>
class World
/st declare atmo without dimensions level and time and bios without time ^{\star}/
/* because they are computed in place and simenv slice is used */
  public: float atmo[45][90];
  public: float bios[36][90];
  public: int
                    atmo g[20];
  public: int
                    bios g;
  private: int
                    level, idecade, istatus, idim;
  public: void computeAtmo(float p1 ,float p2, float p3, float p4)
/* compute dynamics of atmo over space and time, */
/* and of atmo g over time, all dependent on p1,p2,p3,p4 */
     for (idecade=0; idecade<=19; idecade++)</pre>
       for (level=0; level<=3; level++)</pre>
       {...
         idim=3;
         istatus = simenv slice c('atmo', &idim, &level, &level);
         istatus = simenv slice c('atmo', &idim, &idecade, &idecade);
         istatus = simenv put c('atmo',(char *) &atmo);
       }
     }
   }
public: void computeBios(float p1, float p2, float p3, float p4)
/* compute dynamics of bios over space and time, */
/* and of bios g all dependent on p1,p2,p3,p4 */
     for (idecade=0; idecade<=19; idecade++)</pre>
     { . . .
       idim=3:
       istatus = simenv slice c('bios',&idim,&idecade,&idecade);
       istatus = simenv_put_c('bios',(char *) &bios);
/* compute dynamics of bios g */
   }
}
```

```
main (void)
   int run int, istatus;
   char run char[6];
   istatus = simenv_ini_c();
/* check return code of model coupling functions at least here */
   if(istatus != 0) return 1;
/* only if necessary: */
  istatus = simenv_get_run_c(&run_int,run_char);
   float p1 = 1., float p2 = 2., float p3 = 3., float p4 = 4.;
   istatus = simenv_get_c('p1',&p1,&p1);
   istatus = simenv_get_c('p2',&p2,&p2);
   istatus = simenv_get_c('p3',&p3,&p3);
   istatus = simenv_get_c('p4',&p4,&p4);
  World world;
  world.computeAtmo(p1,p2,p3,p4);
   istatus = simenv_put_c('atmo_g',(char *) &(world.atmo_g));
  world.computeBios(p1,p2,p3,p4);
  istatus = simenv_put_c('bios_g',(char *) &(world.bios_g));
  istatus = simenv_end_c();
  return 0;
}
                                                      Example file: world_cpp.cpp
```

**Example 15.3** Model interface for C++ models – model world\_cpp.cpp

## 15.2.5 Python Model

With respect to Example 5.1 the following Python code **world\_py.py** could be used to describe the model coupled to SimEnv. SimEnv modifications are marked in **bold**.

```
#!/usr/local/bin/python
import string
import os
from simenv import *
from math import *
from Numeric import *
atmo=zeros([45,90,4,20], Float)
bios=zeros([36,90,20], Float)
atmo g=zeros([20], Float)
simenv_ini_py()
# only if necessary:
run_int = int(simenv_get_run_py())
p1=1.
p2 = 2.
p3 = 3.
p4=4.
p1 = float(simenv_get_py('p1',p1))
p2 = float(simenv_get_py('p2',p2))
p3 = float(simenv_get_py('p3',p3))
p4 = float(simenv_get_py('p4',p4))
# compute dynamics of atmo and bios over space and time,
# of atmo g over time, all dependent on p1,p2,p3,p4
for idecade in range (20):
  for level in range(4):
atmo=reshape (atmo, 45*90*4*20,))
simenv_put_py('atmo',atmo)
bios=reshape(atmo, 45*90*20,))
simenv_put_py('bios',bios)
simenv put py('atmo g',atmo g)
# compute dynamics of bios g
simenv put py('bios g',bios g)
simenv end py()
                                                         Example file:world_py.py
```

**Example 15.4** Model interface for Python models – model world\_py.py

## 15.2.6 Model Interface at Shell Script Level

Assume any experiment. Assume model executable world\_sh to target values p1 to p4 as arguments from the command line.

The shell script **world\_sh.run** with an interface at shell script level to run the model world\_sh and to transform model output to SimEnv could look like:

```
# always perform at begin
. $SE HOME/simenv ini sh
# create temporary directory run<run char> to perform the model
# and model output transformation from native to SimEnv structure there
. $SE HOME/simenv get run sh
mkdir run$run char
cd run$run char
# get adjustments for p1 ... p4
target name='p1'
target def val=1.
. $SE HOME/simenv get sh
target name='p2'
target def_val=2.
. $SE HOME/simenv get sh
target name='p3'
target def val=3.
. $SE HOME/simenv get sh
target name='p4'
target def val=4.
. $SE HOME/simenv get sh
# run the model
cp ../land sea mask.coarsed .
../world sh $p1 $p2 $p3 $p4
# read model results and output them to SimEnv
../world shput
# clear and remove directory
rm -f *
cd ..
rmdir run$run char
# always perform at end
. $SE HOME/simenv end sh
                                                       Example file: world_sh.run
```

**Example 15.5** Model interface at shell script level – model shell script world\_sh.run

#### 15.2.7 GAMS Model

The SimEnv version comes with a coupled GAMS model **gams\_model.gms** and all associated files that fully correspond with the GAMS example model at http://www.gams.com/docs/gams/Tutorial.pdf. Modifications for SimEnv are marked in **bold**.

```
SETS
        canning plants / SEATTLE, SAN-DIEGO /
  Ι
                         / NEW-YORK, CHICAGO, TOPEKA / ;
PARAMETERS
  A(I) capacity of plant i in cases
         SEATTLE
                     350
         SAN-DIEGO
                     600
  B(J) demand at market j in cases
         NEW-YORK 325
         CHICAGO
                     300
         TOPEKA
                     275 / ;
* - Before using parameter (here: dem ny and dem ch) as SimEnv experiment
   targets they have to be declared as GAMS parameters with their
   default values from above.
* - then insert $include <model> simenv get.inc
  simenv get.inc is generated automatically based on <model>.edf
* - and assign adjusted targets to model variables
  PARAMETERS
  dem ny /325.0/;
  dem ch /300.0/;
  $include gams_model_simenv_get.inc
  A("SEATTLE") = dem ny;
  A("SAN-DIEGO") = dem ch;
TABLE D(I, J) distance in thousands of miles
               NEW-YORK CHICAGO TOPEKA
  SEATTLE
                  2.5
                               1.7
                                            1.8
  SAN-DIEGO
                   2.5
                                1.8
                                             1.4 ;
SCALAR F freight in dollars per case per thousand miles /90/
* get the model status as a model output
  modstat is set to transport.modelstat ;
PARAMETER C(I, J) transport cost in thousands of dollars per case;
  C(I,J) = F * D(I,J) / 1000;
VARIABLES
  X(I,J) shipment quantities in cases
          total transportation costs in thousands of dollars;
POSITIVE VARIABLE X ;
EOUATIONS
             define objective function
   SUPPLY(I) observe supply limit at plant i
  DEMAND(J) satisfy demand at market j ;
             Z = E = SUM((I,J), C(I,J)*X(I,J));
COST ..
             SUM(J, X(I,J)) = L = A(I);
SUPPLY(I) ..
DEMAND(J) .. SUM(I, X(I, J)) = G = B(J);
MODEL TRANSPORT /ALL/ ;
SOLVE TRANSPORT USING LP MINIMIZING Z ;
```

- \* After solving the equations \$include simenv put.inc
- \* has to be inserted.
- \* simenv\_put.inc is generated automatically by SimEnv
- \* based on <model>.edf and <model>.gdf
- \* Additional GAMS commands are possible after the last modification modstat = transport.modelstat 
  \$include gams\_model\_simenv\_put.inc

Example file:gams\_model.gms

**Example 15.6** Model interface for GAMS models – model gams\_model.gms

## 15.2.8 Post-Processor User-Defined Operator

Implementation of the user-defined operator mat\_mul in the file usr\_opr\_mat\_mul.f:

```
function icheck user def operator()
  declare fields to hold extents and coordinates
   dimension iext1(9), iext2(9)
   dimension ico blk1(9), ico blk2(9)
   dimension ico beg1(9), ico beg2(9)
c get dimensionality idimens, extents iext,
c coordinate block number ico blk and
c begin number for coordinates ico_beg in coordinate block ico_blk
  idimens1=iget dim arg(1,iext1)
   idimens2=iget dim arg(2,iext2)
   iok=iget co arg(1,ico blk1,ico beg1)
  iok=iget co arg(2,ico blk2,ico beg2)
c get check modus for coordinates
   ichk modus=iget co chk modus()
   if(idimens1.ne.2.or.idimens2.ne.2) then
 wrong dimensionalities
      ierror=1
   else
      if(iext1(2).ne.iext2(1)) then
  wrong extents
         ierror=2
      else
         if(ico blk1(2).eq.ico blk2(1)) then
  coordinates identical
            if (ico beg1(2).eq.ico beg2(1)) then
               iret=31
            else
               iret=33
            endif
         else
   differing coordinates
            iret=32
            if (ichk modus.eq.1) then
  check only for weak coordinate
               do j=0, iext1(2)-1
  get coordinate values
                  iretv1=iget_coord_val(
                         ico blk1(2),ico beg1(2)+j,value1)
                  iretv2=iget_coord_val(
                         ico blk2(1),ico beg2(1)+j,value2)
  iret=33: differing coordinate values
                  if(value1.ne.value2) iret=33
               enddo
            endif
         endif
         ierror=0
         if (ichk modus.eq.2) then
            if(iret.gt.31) ierror=3
         elseif(ichk modus.eq.1) then
            if(iret.gt.32) ierror=3
         endif
      endif
   endif
```

```
if(ierror.eq.0) then
     iext1(2) = iext2(2)
     ico blk1(2) = ico blk2(2)
     ico beg1(2)=ico beg2(2)
     iok=iput_struct_res(0,idimens1,iext1,ico_blk1,ico_beg1)
   endif
c return error code
   icheck user def operator=ierror
   return
   end
   function icompute_user_def_operator(res)
c SimEnv operator results are always of type real*4
   real*4 res(1)
c auxiliary variables
   integer*4 iext1(9),iext2(9)
   real*8 value8
c get dimensionality idimens and extents iext for both arguments
   idimens=iget dim arg(1,iext1)
   idimens=iget dim arg(2,iext2)
c perform matrix multiplication
   m=0
   do k=1, iext2(2)
      ioffs2=(k-1)*iext2(1)
      do i=1, iext1(1)
         ioffs1=i
   res(i,k) = sum(arg1(i,l) * arg2(l,k))
C
         value8=0.
         indi defined=0
         do l=1, iext1(2)
            ia1=ioffs1+(l-1)*iext1(1)
            ia2=ioffs2+l
            fac1=arg1(ia1)
            fac2=arg2(ia2)
            if(is undef(fac1)+is undef(fac2).eq.0) then
               indi defined=1
               value8=value8+fac1*fac2
            endif
         enddo
         m=m+1
         if (indi defined.eq.0) then
            res(m) = set undef()
            res(m)=clip undef(value8)
      enddo
   enddo
 return error code
   icompute user def operator=0
   return
   end
                                                   Example file: usr_opr_mat_mul.f
```

**Example 15.7** Post-processor user-defined operator module – operator mat\_mul

### 15.2.9 Post-Processor Result Import Interface

Implementation of an interface to import post-processor output from SimEnv in ASCII format:

```
program read ieee file
  real*4, pointer, dimension(:) :: coord values
  real*4, pointer, dimension(:) :: result values
  integer*4 name length, idummy, idim, iext(9), length values
  character result name*512, coord name*20
  open(unit=1,file='world f.inf03.ascii',form='formatted',status='old')
  open(unit=2,file='world f.res03.ascii',form='formatted',status='old')
iostat=0
  do while (iostat.eq.0)
      read(1,'(i8)',iostat=iostat) name length
      if (iostat.eq.0) then
         backspace(1)
         read(1,11) idummy,result_name(1:name_length)
11
         format(i8,a<name length>)
         read(1,'(10i8)') idim, (iext(i), i=1,9)
         length result=1
         do i=1, idim
            result no. res<i>:
            length result=length result*iext(i)
            read(1,'(a20)') coord name
            allocate(coord values(iext(i)))
            ibeg=1
            do while (ibeg.le.iext(i))
               iend=min0(ibeg+9,iext(i))
               read(1,'(10g12.6)') (coord_values(i),i=ibeg,iend)
            enddo
            further processing of coordinate values
С
            deallocate (coord values)
         enddo
         allocate(result values(length result))
         ibeq=1
         do while (ibeg.le.length result)
            iend=min0(ibeg+9,length result)
            read(2,'(10g12.6)') (result values(i), i=ibeg, iend)
         enddo
         further processing of result values
\sim
         deallocate (result values)
      endif
   enddo
   close(unit=1)
   close(unit=2)
  end
```

**Example 15.8** ASCII compliant post-processor result import interface

# 15.3 Compilation of Post-Processor Built-in Operators and Operator Arguments

# 15.3.1 Post-Processor Built-in Operators (in Thematic Order)

arggeneral numerical argumentconst\_argconstant value argumentchar\_argcharacter argument

Name	Meaning	
Elemental operators		
arg1 + arg2	addition	
arg1 - arg2	subtraction	
arg1 * arg2	multiplication	
arg1 / arg2	division	
arg1 ** arg2	exponentiation	
+ arg	identity	
- arg	negation	
( arg )	parentheses	
Basic op	perators	
abs(arg)	absolute value	
dim(arg1,arg2)	positive difference	
exp(arg)	exponentiation	
int(arg)	truncation value	
log(arg)	natural logarithm	
log10(arg)	decade logarithm	
mod(arg1,arg2)	remainder	
nint(arg)	round value	
sign(arg)	sign of value	
sqrt(arg)	square root	
Trigonor	metric operators	
sin(arg)	sine	
cos(arg)	cosine	
tan(arg)	tangent	
cot(arg)	cotangent	
asin(arg)	arc sine	
acos(arg)	arc cosine	
atan(arg)	arc tangent	
acot(arg)	arc cotangent	
sinh(arg)	hyperbolic sine	
cosh(arg)	hyperbolic cosine	
tanh(arg)	hyperbolic tangent	
coth(arg)	hyperbolic cotangent	
, ,	Miscellaneous operators	
classify(const_arg,arg)	classification of arg into const_arg classes	
clip(char_arg,arg)	clip arg according to char arg	
cumul(char_arg,arg)	cumulates arg according to char_arg	
January, arg,	carrialated any according to orial_any	

Name	Meaning
flip(char_arg,arg)	flip arg according to char_arg
get_experiment(char_arg1, char_arg2,char_arg3,arg)	include an other experiment
get_table_fct(char_arg,arg)	table function with linear interpolation of table char_arg for position arg
if(char_arg,arg1,arg2,arg3)	general purpose conditional if-construct
mask(char_arg,arg1,arg2)	mask elements of argument arg1
matmul(arg1,arg2)	matrix multiplication
nr_of_runs()	number of single runs of the current experiment
rank(char_arg,arg)	rank of arg according to char_arg
run(char_arg,arg)	values of arg for a single run selected by char_arg
transpose(char_arg,arg)	transpose arg according to char_arg
undef()	undefined element
	n and moment operators for arguments
min(arg)	argument minimum of values
max(arg)	argument maximum of values
sum(arg)	argument sum of values
avg(arg)	argument arithmetic mean of values
var(arg)	argument variance of values
avgg(arg)	argument geometric mean of values
avgh(arg)	argument harmonic mean of values
avgw(arg1,arg2)	argument weighted mean of values
hgr(const_arg,arg)	argument histogram of values
count(char_arg,arg)	count number of values
minprop(arg)	index of the element where the minimum is reached the first time
maxprop(arg)	index of the element where the maximum is reached the first time
	gregation and moment operators for arguments
min_n(arg1,,argn)	minimum per element
max_n(arg1,,argn)	maximum per element
minprop_n(arg1,,argn)	argument position (1 n) where the minimum is reached the first time
maxprop_n(arg1,,argn)	argument position (1 n) where the maximum is reached the first time
	-related aggregation and moment operators for arguments
min_l(char_arg,arg)	dimension-related argument minima of values
max_l(char_arg,arg)	dimension-related argument maxima of values
sum_l(char_arg,arg)	dimension-related argument sums of values
avg_l(char_arg,arg)	dimension-related argument arithmetic means of values
var_l(char_arg,arg)	dimension-related argument variances of values
avgg_l(char_arg,arg)	dimension-related argument geometric means of values
avgh_l(char_arg,arg)	dimension-related argument harmonic means of values
avgw_l(char_arg,arg1,arg2)	dimension-related argument weighted means of values
hgr_l(char_arg,const_arg,arg)	dimension-related argument histograms of values
count_l(char_arg1,char_arg2, arg)	dimension-related count numbers of values
minprop_l(char_arg,arg)	dimension-related argument position (1 n) where the minimum is reached the first time
maxprop_l(char_arg,arg)	dimension-related argument position (1 n) where the maximum is reached the first time
Multi-run o	perators (behavioural analysis)
behav(char_arg,arg)	general purpose operator for navigating and aggregating in the experiment space



Name	Meaning	
Multi-run operators (Monte Carlo analysis)		
cnf(const_arg,arg)	positive distance of confidence line from mean avg_e(arg)	
cor(arg1,arg2)	correlation coefficient between arg1 and arg2	
cov(arg1,arg2)	covariance between arg1 and arg2	
ens(arg)	whole Monte Carlo run ensemble	
krt(arg)	kurtosis (4 <sup>th</sup> moment)	
med(arg)	median	
qnt(const_arg,arg)	quantile of arg	
reg(arg1,arg2)	linear regression coefficient to forecast arg2 from arg1	
rng(arg)	range = max_e(arg) - min_e(arg) skewness (3 <sup>rd</sup> moment)	
skw(arg)	skewness (3 <sup>rd</sup> moment)	
stat(const_arg1,const_arg2, const_arg3,const_arg4,arg5)	basic statistical summaries	
min_e(arg)	run ensemble minimum	
max_e(arg)	run ensemble maximum	
sum_e(arg)	run ensemble sum	
avg_e(arg)	run ensemble mean	
var_e(arg)	run ensemble variance	
avgg_e(arg)	run ensemble geometric mean	
avgh_e(arg)	run ensemble harmonic mean	
avgw_e(arg1,arg2)	run ensemble weighted mean	
hgr_e(const_arg,arg)	heuristic probability density function	
count_e(char_arg,arg)	run ensemble count number of values	
minprop_e(arg)	run number where the minimum is reached the first time	
maxprop_e(arg)	run number where the maximum is reached the first time	
Multi-run operators (local sensitivity analysis)		
sens_abs(char_arg,arg)	absolute sensitivity measure	
sens_rel(char_arg,arg)	relative sensitivity measure	
sym_abs(char_arg,arg)	absolute symmetry measure	
sym_rel(char_arg,arg)	relative symmetry measure	
lin_abs(char_arg,arg)	absolute linearity measure	
lin_rel(char_arg,arg)	relative linearity measure	

 Tab. 15.6
 Post-processor built-in operators (in thematic order)

# 15.3.2 Post-Processor Built-in Operators (in Alphabetic Order)

arggeneral numerical argumentconst\_argconstant value argumentchar\_argcharacter argument

Name	Meaning
- arg	negation
(arg)	parentheses
+ arg	identity
abs(arg)	absolute value
acos(arg)	arc cosine
acot(arg)	arc cotangent
arg1 - arg2	subtraction
arg1 * arg2	multiplication
arg1 ** arg2	exponentiation
arg1 / arg2	division
arg1 + arg2	addition
asin(arg)	arc sine
atan(arg)	arc tangent
avg(arg)	argument arithmetic mean of values
avg_e(arg)	run ensemble mean
avg_l(char_arg,arg)	dimension-related argument arithmetic means of values
avgg(arg)	argument geometric mean of values
avgg_e(arg)	run ensemble geometric mean
avgg_l(char_arg,arg)	dimension-related argument geometric means of values
avgh(arg)	argument harmonic mean of values
avgh_e(arg)	run ensemble harmonic mean
avgh_l(char_arg,arg)	dimension-related argument harmonic means of values
avgw(arg1,arg2)	argument weighted mean of values
avgw_e(arg1,arg2)	run ensemble weighted mean
avgw_l(char_arg,arg1,arg2)	dimension-related argument weighted means of values
behav(char_arg,arg)	general purpose operator for navigating and aggregating in the experiment space
classify(const_arg,arg)	classification of arg into const_arg classes
clip(char_arg,arg)	clip arg according to char_arg
cnf(const_arg,arg)	positive distance of confidence line from mean avg_e(arg)
cor(arg1,arg2)	correlation coefficient between arg1 and arg2
cos(arg)	cosine
cosh(arg)	hyperbolic cosine
cot(arg)	cotangent
coth(arg)	hyperbolic cotangent
count(char arg,arg)	count number of values
count_e(char_arg,arg)	run ensemble count
count_l(char_arg1,char_arg2, arg)	dimension-related count numbers of values
cov(arg1,arg2)	covariance between arg1 and arg2
cumul(char_arg,arg)	cumulates arg according to char_arg
dim(arg1,arg2)	positive difference
ens(arg)	whole Monte Carlo run ensemble
exp(arg) exponentiation	
flip(char_arg,arg)	flip arg according to char_arg

get_experiment(char_arg3, and) get_table_fct(char_arg3, and) get_table_fct(char_arg3, and) hgr(const_arg,arg) hgr_e(const_arg,arg) hcuristic probability density function hcuristic probability	Name	Meaning
get_table_fct(char_arg,arg)   table function with linear interpolation of fable char_arg for position arg herr (const. arg,arg)   for_e(const. arg,arg)   for_e(const. arg,arg)   for_e(const. arg,arg)   filear arg. arg. arg. arg. arg. arg. arg. ar		include an other experiment
hgr(const_arg.arg) hgr (const_arg.arg) heuristic probability density function hgr (const_arg.arg) lif(char arg.arg1,arg2,arg3) lif(char arg.arg1,arg2,arg3) lin_abs(char_arg.arg) lin_abs(char_arg.arg) lin_abs(char_arg.arg) lin_abs(char_arg.arg) lin_abs(char_arg.arg) lin_abs(char_arg.arg) log(arg) log(arg		table function with linear interpolation of table char arg for position arg
Ingr.e(const.arg.arg)   heuristic probability density function   heuristic probability function   heuristic probability density function   heuristic probability function   heuristic probability function   heuristic probability density function   heuristic probability measure		
ingr_(Ichar_arg.const_arg.arg) dimension-related argument histograms of values   int(arg) truncation value   krt(arg) kurtosis (4 <sup>th</sup> moment)   ilin_abs(char_arg.arg)   ilin_abs(char_arg.arg.arg)   ilin_abs(char_arg.arg.arg)   ilin_abs(char_arg.arg)   ilin_abs(char_arg.a		ů ů
int(arg) truncation value (with a grang 1 arg 2, arg 3) general purpose conditional if-construct (int(arg) truncation value (with a grang) (withouts) (4" moment) (with a grang) (withouts) (4" moment) (with a grang) (		····
int(arg) kurlosis (4 <sup>th</sup> moment) lin_abs(char_arg,arg) absolute linearity measure lin_rel(char_arg,arg) relative linearity measure log(arg) natural logarithm log10(arg) decade logarithm mask(char_arg,arg1,arg2) mask elements of argument arg1 max(arg) argument maximum of values max_(arg) run ensemble maximum max_(char_arg,arg) index of the element where the maximum is reached the first time maxprop_e(arg) run ensemble minimum maxprop_l(char_arg,arg) index of the element where the maximum is reached the first time maxprop_l(char_arg,arg) argument position (1 n) where the maximum is reached the first time maxprop_l(arg1,argn) argument position (1 n) where the maximum is reached the first time maxprop_l(arg1,argn) argument position (1 n) where the maximum is reached the first time maxprop_l(arg1,argn) argument position (1 n) where the maximum is reached the first time maxprop_l(arg1,argn) median min(arg1) argument minimum of values min_l(char_arg,arg) dimension-related argument minima of values min_l(argn) argument minimum of values min_l(argn) run ensemble minimum minprop_l(argn) index of the element where the minimum is reached the first time minprop_l(argn) index of the element where the minimum is reached the first time minprop_l(argn) index of the element where the minimum is reached the first time minprop_l(argn) index of the element where the minimum is reached the first time minprop_l(argn) index of the element where the minimum is reached the first time minprop_l(argn) argument position (1 n) where the minimum is reached the first time minprop_l(argn) argument position (1 n) where the minimum is reached the first time minprop_l(argn) argument position (1 n) where the minimum is reached the first time minprop_l(argn) argument position (1 n) where the minimum is reached the first time minprop_l(argn) argument position (1 n) where the minimum is reached the first time minprop_l(argn) argument position (1 n) where the minimum is reached the first time minprop		
krt(arg)         kurtosis (4*** moment)           lin_abs(char_arg,arg)         absolute linearity measure           lin_fel(char_arg,arg)         relative linearity measure           log(arg)         natural logarithm           log(arg)         decade logarithm           mask(char_arg,arg1,arg2)         matrix multiplication           max/arg1         argument maximum of values           max_e(arg)         run ensemble maximum           max_l(char_arg,arg)         dimension-related argument maximus of values           max_mx_(rag1,argn)         maximum per element           maxprop(arg)         index of the element where the maximum is reached the first time           maxprop_(arg1)         index of the element where the maximum is reached the first time           maxprop_(arg1,argn)         argument maximum to reached the first time           maxprop_(arg1,argn)         argument position (1 n) where the maximum is reached the first time           min_(arg1,argn)         median           min_(arg1,argn)         median           min_(arg1,argn)         miniminum           min_(arg1,argn)         miniminum           min_(arg1,argn)         miniminum per element           minprop_(arg1)         index of the element where the minimum is reached the first time           minprop_		0 1 1
lin_abs(char_arg,arg)         absolute linearity measure           lin_rel(char_arg,arg)         relative linearity measure           log(10(arg))         relative linearity measure           log(10(arg))         decade logarithm           mask(char_arg,arg1,arg2)         maximul maximum           max(arg)         maximul maximum of values           max_e(arg)         run ensemble maximum           max_al(char_arg,arg)         dimension-related argument maxima of values           max_n(arg1,,argn)         maximum per element           maxprop(arg)         index of the element where the maximum is reached the first time           maxprop_(lchar_arg,arg)         dimension-related argument position (1 n) where the maximum is reached the first time           maxprop_n(arg1,,argn)         argument position (1 n) where the maximum is reached the first time           med(arg)         median           min_e(arg)         argument minimum of values           min_e(arg)         run ensemble minimum           min_e(arg)         unesemble minimum           min_f(arg1,,argn)         minimum per element           min_relarg1,,argn)         minimum per element           maxprop_n(arg1,,argn)         minimum of values           min_e(arg)         run element           min_f(arg1,,argn)		kurtosis (4 <sup>th</sup> moment)
In_rel(char_arg,arg)   relative linearity measure		
log(arg) log10(arg) decade logarithm mask(char_arg,arg1,arg2) matmul(arg1,arg2) matmul(arg1,arg2) matck elements of argument arg1 matmul(arg1,arg2) max (arg) max (arg) max (arg) max (char_arg,arg) dimension-related argument maximum is reached the first time maxprope_(arg) maxprop_(char_arg,arg) min(arg) min(arg) min(arg) min_(arg1,argn) minimum min_(char_arg,arg) min_(arg1,argn) minimum minprop_(char_arg,arg) minprop_(char_arg		
Iog10(arg)   decade logarithm   mask(char_arg.arg1,arg2)   mark elements of argument arg1   mask (char_arg.arg1,arg2)   mark multiplication   max_(arg)   argument maximum of values   max_(arg, arg.arg)   dimension-related argument maximum is reached the first time   maxprop_(arg)   max_(arg,arg,arg)   max_max_m(arg1,,argn)   maximum per element   maxprop_(arg)   maxprop_(char_arg,arg)   maxprop_(char_arg,arg)   dimension-related argument maximum is reached the first time   maxprop_(char_arg,arg)   munumber where the maximum is reached the first time   maxprop_(char_arg,arg)   dimension-related argument position (1 n) where the maximum is reached the first time   median   argument position (1 n) where the maximum is reached the first time   median   argument minimum of values   min_(arg)   min_(arg)   min_(arg, arg, arg)   dimension-related argument minima of values   min_(arg, arg, arg)   min_maxprop_(char_arg, arg)   min_maxprop_(char_arg, arg)   min_maxprop_(char_arg, arg)   min_maxprop_(char_arg, arg)   min_maxprop_(char_arg, arg, arg, arg, arg, arg, arg, arg,		•
mask(char_arg,arg1,arg2)         mask elements of argument arg1           matulu(arg1,arg2)         matrix multiplication           max(arg)         argument maximum of values           max_e(arg)         run ensemble maximum           max_(lchar_arg,arg)         dimension-related argument maxima of values           max_n(arg1,argn)         maximum per element           maxprop(arg)         index of the element where the maximum is reached the first time           maxprop (arg1,argn)         minesion-related argument position (1 n) where the maximum is reached the first time           maxprop_n(arg1,,argn)         argument position (1 n) where the maximum is reached the first time           med(arg)         median           min(arg1)         argument minimum of values           min_e(arg)         run ensemble minimum           min_e(arg)         minimum per element           min_e(arg)         minimum per element           minprop(arg)         index of the element where the minimum is reached the first time           minprop_e(arg)         run unmber where the minimum is reached the first time           minprop_n(arg1,argn)         dimension-related argument position (1 n) where the minimum is reached the first time           minprop_n(arg1,argn)         dimension-related argument position (1 n) where the minimum is reached the first time		
matnul(arg1,arg2) matrix multiplication max(arg) argument maximum of values max_e(arg) run ensemble maximum max_l(char_arg,arg) dimension-related argument maximum seached the first time maxprop_e(arg) run number where the maximum is reached the first time maxprop_e(arg) run number where the maximum is reached the first time maxprop_e(arg) run number where the maximum is reached the first time maxprop_n(arg1,argn) argument position (1 n) where the maximum is reached the first time maxprop_n(arg1,argn) argument position (1 n) where the maximum is reached the first time med(arg) median min(arg) argument minimum of values min_e(arg) run ensemble minimum min_l(char_arg,arg) dimension-related argument minima of values mininprop(arg) index of the element where the minimum is reached the first time minprop_e(arg) index of the element where the minimum is reached the first time minprop_l(char_arg,arg) dimension-related argument position (1 n) where the minimum is reached the first time minprop_n(arg1,,argn) argument position (1 n) where the minimum is reached the first time minprop_n(arg1,,argn) argument position (1 n) where the minimum is reached the first time minprop_n(arg1,,argn) argument position (1 n) where the minimum is reached the first time minprop_n(arg1,,argn) argument position (1 n) where the minimum is reached the first time minprop_n(arg1,,argn) argument position (1 n) where the minimum is reached the first time minprop_n(arg1,,argn) argument position (1 n) where the minimum is reached the first time minprop_n(arg1,,argn) argument position (1 n) where the minimum is reached the first time minprop_n(arg1,,argn) argument position (1 n) where the minimum is reached the first time minprop_n(arg1,,argn) argument position (1 n) where the minimum is reached the first time minprop_n(arg1,,argn) argument position (1 n) where the minimum is reached the first time for a run number of argument of argument of argument argument argu		
max(arg) argument maximum of values max_e(arg) run ensemble maximum max_l(char_arg,arg) dimension-related argument maxima of values max_n(arg1,,argn) maximum per element maxprop(arg) index of the element where the maximum is reached the first time maxprop_e(arg) run number where the maximum is reached the first time maxprop_l(char_arg,arg) dimension-related argument position (1 n) where the maximum is reached the first time maxprop_n(arg1,argn) argument position (1 n) where the maximum is reached the first time med(arg) median min(arg1) argument minimum of values min_e(arg) run ensemble minimum min_l(char_arg,arg) dimension-related argument minima of values min_n(arg1,argn) minimum per element minprop(arg) index of the element where the minimum is reached the first time minprop_e(arg) run number where the minimum is reached the first time minprop_n(arg1,argn) argument position (1 n) where the minimum is reached the first time minprop_n(arg1,argn) argument position (1 n) where the minimum is reached the first time mod(arg1,arg2) remainder mod(arg1,arg2) remainder nrun(arg1,argn) round value nr_of runs() number of single runs of the current experiment qunt(const_arg,arg) rank of arg according to char_arg reg(arg1,arg2) linear regression coefficient to forecast arg2 from arg1 reg(arg1,arg2) range = max_e(arg) - min_e(arg) run(char_arg,arg) sens_el(char_arg,arg) sens_el(char_arg,arg) sens_elected argument position (1 n) selected by char_arg sens_abs(char_arg,arg) sign of value sin(arg) sign of values sin(arg) sign of values sin(arg) selected sign select		
max_e(arg) run ensemble maximum  max_l(char_arg,arg) dimension-related argument maxima of values  max_n(arg1,argn) maximum per element  maxprop_e(arg) index of the element where the maximum is reached the first time  maxprop_e(arg) run number where the maximum is reached the first time  maxprop_l(char_arg,arg) dimension-related argument position (1 n) where the maximum is reached the first time  maxprop_n(arg1,argn) argument position (1 n) where the maximum is reached the first time  maxprop_n(arg1,argn) argument minimum of values  min_e(arg) run ensemble minimum  min_l(char_arg,arg) dimension-related argument minima of values  min_n(arg1,argn) minimum per element  minprop(arg1,argn) minimum per element  minprop e(arg) run number where the minimum is reached the first time  minprop_l(char_arg,arg) dimension-related argument position (1 n) where the minimum is reached the first time  minprop_l(arg1,argn) dimension-related argument position (1 n) where the minimum is reached the first time  minprop_n(arg1,argn) argument position (1 n) where the minimum is reached the first time  minprop_n(arg1,argn) argument position (1 n) where the minimum is reached the first time  minprop_n(arg1,argn) argument position (1 n) where the minimum is reached the first time  minprop_n(arg1,argn) argument position (1 n) where the minimum is reached the first time  minprop_n(arg1,argn) argument position (1 n) where the minimum is reached the first time  minprop_n(arg1,argn) argument position (1 n) where the minimum is reached the first time  minprop_n(arg1,argn) argument position (1 n) where the minimum is reached the first time  minprop_n(arg1,argn) argument position (1 n) where the minimum is reached the first time  minprop_n(arg1,argn) argument position (1 n) where the minimum is reached the first time  minprop_n(arg1,argn) argument position (1 n) where the minimum is reached the first time  minprop_n(arg1,argn) argument posit	, , ,	·
max_l(char_arg,arg)         dimension-related argument maxima of values           max_n(arg1,,argn)         maximum per element           maxprop(arg)         index of the element where the maximum is reached the first time           maxprop_e(arg)         run number where the maximum is reached the first time           maxprop_l(char_arg,arg)         dimension-related argument position (1 n) where the maximum is reached the first time           maxprop_n(arg1,argn)         argument position (1 n) where the maximum is reached the first time           med(arg)         median           min(arg)         argument minimum of values           min_e(arg)         run ensemble minimum           min_l(char_arg,arg)         dimension-related argument minima of values           min_n(arg1,,argn)         minimum per element           minprop_e(arg)         run number where the minimum is reached the first time           minprop_l(char_arg,arg)         dimension-related argument position (1 n) where the minimum is reached the first time           minprop_n(arg1,,argn)         argument position (1 n) where the minimum is reached the first time           minprop_n(arg1,,argn)         argument position (1 n) where the minimum is reached the first time           minprop_n(arg1,,argn)         argument position (1 n) where the minimum is reached the first time           minprop_n(arg1,,argn)         argume		
max_n(arg1,argn)         maximum per element           maxprop(arg)         index of the element where the maximum is reached the first time           maxprop_e(arg)         run number where the maximum is reached the first time           maxprop_n(arg1,argn)         argument position (1 n) where the maximum is reached the first time           maxprop_n(arg1,argn)         argument position (1 n) where the maximum is reached the first time           med(arg)         median           min_e(arg)         argument minimum of values           min_e(arg)         run ensemble minimum           min_e(arg1,argn)         minimum per element           min_e(arg1,argn)         minimum per element           min_rop(arg1,argn)         minimum per element where the minimum is reached the first time           minprop (arg1,argn)         run number where the minimum is reached the first time           minprop_l(char_arg,arg)         argument position (1 n) where the minimum is reached the first time           minprop_n(arg1,argn)         argument position (1 n) where the minimum is reached the first time           minprop_n(arg1,argn)         argument position (1 n) where the minimum is reached the first time           minprop_n(arg1,argn)         argument position (1 n) where the minimum is reached the first time           minprop_n(arg1,argn)         argument position (1 n) where the mi		
maxprop(arg) index of the element where the maximum is reached the first time maxprop_e(arg) run number where the maximum is reached the first time maxprop_l(char_arg,arg) dimension-related argument position (1 n) where the maximum is reached the first time maxprop_n(arg1,,argn) mayprop_n(arg1,,argn) med(arg) med(arg) min(arg) argument minimum of values min_e(arg) min_l(char_arg,arg) min_l(char_arg,arg) min_l(char_arg,arg) min_l(char_arg,arg) min_m(arg1,,argn) minimum per element minprop(arg) minprop_e(arg) minprop_l(char_arg,arg) minprop_l(char_arg,arg) minprop_n(arg1,,argn) minprop_n(arg1,,argn) minprop_n(arg1,,argn) minprop_n(arg1,,argn) minprop_n(arg1,,argn) minprop_n(arg1,,argn) argument position (1 n) where the minimum is reached the first time minprop_n(arg1,,argn) mod(arg1,arg2) mod(arg1,arg2) mod(arg1,arg2) mod(arg1,arg2) mod(arg1,arg2) mod(arg1,arg2) mod(arg1,arg3) maxprop_n(arg1,,argn) mod(arg1,arg3) maxprop_n(arg1,,argn) mod(arg1,arg3) mod(a		
maxprop_e(arg) run number where the maximum is reached the first time dimension-related argument position (1 n) where the maximum is reached the first time argument position (1 n) where the maximum is reached the first time argument position (1 n) where the maximum is reached the first time medicarg) median argument position (1 n) where the maximum is reached the first time medicarg) min (arg) argument minimum of values min_e(arg) run ensemble minimum dimension-related argument minima of values min_n(arg1,,argn) minimum per element minimum is reached the first time minimorpo_e(arg) run number where the minimum is reached the first time minimorpo_e(arg) run number where the minimum is reached the first time minimorpo_n(arg1,,argn) argument position (1 n) where the minimum is reached the first time minimorpo_n(arg1,,argn) argument position (1 n) where the minimum is reached the first time minimorpo_n(arg1,,argn) argument position (1 n) where the minimum is reached the first time minimorpo_n(arg1,,argn) argument position (1 n) where the minimum is reached the first time minimorpo_n(arg1,,argn) argument position (1 n) where the minimum is reached the first time minimorpo_n(arg1,,argn) argument position (1 n) where the minimum is reached the first time minimorpo_n(arg1,arg2) round value  n_of_runs() number of single runs of the current experiment quantile of arg arg.  reg(arg1,arg2) rank of arg according to char_arg  reg(arg1,arg2) linear regression coefficient to forecast arg2 from arg1  rng(arg) range = max_e(arg) - min_e(arg)  run(char_arg,arg) values of arg for a single run selected by char_arg  sens_abs(char_arg,arg) seloute sensitivity measure  sens_abs(char_arg,arg) relative sensitivity measure  sens_abs(char_arg,arg) sine  sinn(arg) square root  basic statistical summaries		'
maxprop_l(char_arg,arg)  maxprop_l(arg1,,argn)  maxprop_n(arg1,,argn)  med(arg)  min(arg)  min(arg)  min_e(arg)  minimum of values  min_e(arg)  minimum per element  minprop(arg)  minimum per element  minprop_e(arg)  minprop_e(arg)  minprop_l(char_arg,arg)  minprop_n(arg1,,argn)  minprop_n(a		
reached the first time maxprop_n(arg1,,argn) argument position (1 n) where the maximum is reached the first time med(arg) median min(arg) argument minimum of values min_e(arg) run ensemble minimum min_l(char_arg,arg) dimension-related argument minima of values min_n(arg1,,argn) minimum per element minprop(arg) run number where the minimum is reached the first time minprop_e(arg) run number where the minimum is reached the first time minprop_l(char_arg,arg) dimension-related argument position (1 n) where the minimum is reached the first time minprop_n(arg1,,argn) argument position (1 n) where the minimum is reached the first time minprop_n(arg1,,argn) argument position (1 n) where the minimum is reached the first time mod(arg1,arg2) remainder mint(arg) round value n_of_runs() number of single runs of the current experiment qut(const_arg,arg) quantile of arg rank(char_arg,arg) rank of arg according to char_arg reg(arg1,arg2) linear regression coefficient to forecast arg2 from arg1 rng(arg) range = max_e(arg) - min_e(arg) run(char_arg,arg) values of arg for a single run selected by char_arg sens_abs(char_arg,arg) absolute sensitivity measure sens_rel(char_arg,arg) sign of value sin(arg) sign of value sin(arg) skewness (3 <sup>rd</sup> moment) sqrt(arg) skewness (3 <sup>rd</sup> moment) sqrt(arg) square root basic statistical summaries  sum(arg) argument sum of values		
med(arg)         median           min(arg)         argument minimum of values           min_e(arg)         run ensemble minimum           min_l(char_arg,arg)         dimension-related argument minima of values           min_n(arg1,,argn)         minimum per element           minprop(arg)         index of the element where the minimum is reached the first time           minprop_(earg)         run number where the minimum is reached the first time           minprop_(char_arg,arg)         dimension-related argument position (1 n) where the minimum is reached the first time           minprop_n(arg1,,argn)         argument position (1 n) where the minimum is reached the first time           mod(arg1,arg2)         remainder           nint(arg)         round value           nr_of_runs()         number of single runs of the current experiment           qnt(const_arg,arg)         quantile of arg           rank(char_arg,arg)         rank of arg according to char_arg           reg(arg1,arg2)         linear regression coefficient to forecast arg2 from arg1           rng(arg)         range = max_e(arg) - min_e(arg)           run(char_arg,arg)         values of arg for a single run selected by char_arg           sens_abs(char_arg,arg)         absolute sensitivity measure           sens_rel(char_arg,arg)         sign of value           s	maxprop_i(onal_arg,arg)	
med(arg)         median           min(arg)         argument minimum of values           min_e(arg)         run ensemble minimum           min_l(char_arg,arg)         dimension-related argument minima of values           min_n(arg1,,argn)         minimum per element           minprop(arg)         index of the element where the minimum is reached the first time           minprop_(earg)         run number where the minimum is reached the first time           minprop_(char_arg,arg)         dimension-related argument position (1 n) where the minimum is reached the first time           minprop_n(arg1,,argn)         argument position (1 n) where the minimum is reached the first time           mod(arg1,arg2)         remainder           nint(arg)         round value           nr_of_runs()         number of single runs of the current experiment           qnt(const_arg,arg)         quantile of arg           rank(char_arg,arg)         rank of arg according to char_arg           reg(arg1,arg2)         linear regression coefficient to forecast arg2 from arg1           rng(arg)         range = max_e(arg) - min_e(arg)           run(char_arg,arg)         values of arg for a single run selected by char_arg           sens_abs(char_arg,arg)         absolute sensitivity measure           sens_rel(char_arg,arg)         sign of value           s	maxprop n(arg1,,argn)	argument position (1 n) where the maximum is reached the first time
min(arg) argument minimum of values min_e(arg) run ensemble minimum min_l(char_arg,arg) dimension-related argument minima of values min_n(arg1,,argn) minimum per element minprop(arg) index of the element where the minimum is reached the first time minprop_e(arg) run number where the minimum is reached the first time minprop_e(arg) dimension-related argument position (1 n) where the minimum is reached the first time minprop_n(arg1,,argn) argument position (1 n) where the minimum is reached the first time mod(arg1,arg2) remainder nint(arg) round value nr_of_runs() number of single runs of the current experiment qnt(const_arg,arg) quantile of arg rank(char_arg,arg) rank of arg according to char_arg reg(arg1,arg2) linear regression coefficient to forecast arg2 from arg1 rng(arg) range = max_e(arg) - min_e(arg) run(char_arg,arg) values of arg for a single run selected by char_arg sens_abs(char_arg,arg) absolute sensitivity measure sens_rel(char_arg,arg) relative sensitivity measure sens_rel(char_arg,arg) sign of value sin(arg) sine sin(arg) sine sin(arg) skewness (3 <sup>rd</sup> moment) sqrt(arg) square root stat(const_arg1,const_arg2, const_arg3,const_arg4,arg5) sum(arg) argument sum of values		
min_e(arg) run ensemble minimum min_l(char_arg,arg) dimension-related argument minima of values min_n(arg1,,argn) minimum per element minprop(arg) index of the element where the minimum is reached the first time minprop_e(arg) run number where the minimum is reached the first time minprop_l(char_arg,arg) dimension-related argument position (1 n) where the minimum is reached the first time minprop_n(arg1,,argn) argument position (1 n) where the minimum is reached the first time mod(arg1,arg2) remainder mint(arg) round value mr_of_runs() number of single runs of the current experiment qnt(const_arg,arg) quantile of arg rank(char_arg,arg) rank of arg according to char_arg reg(arg1,arg2) linear regression coefficient to forecast arg2 from arg1 rng(arg) range = max_e(arg) - min_e(arg) run(char_arg,arg) values of arg for a single run selected by char_arg sens_abs(char_arg,arg) absolute sensitivity measure sens_rel(char_arg,arg) relative sensitivity measure sign(arg) sign of value sin(arg) sign of value sin(arg) sign of value sin(arg) sign of value sin(arg) shewness (3 <sup>rd</sup> moment) sqrt(arg) square root stat(const_arg1,const_arg2, const_arg4,arg5) sum(arg) argument sum of values	, ,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,	argument minimum of values
min_l(char_arg,arg) dimension-related argument minima of values min_n(arg1,,argn) minimum per element minprop(arg) index of the element where the minimum is reached the first time minprop_e(arg) run number where the minimum is reached the first time minprop_l(char_arg,arg) dimension-related argument position (1 n) where the minimum is reached the first time minprop_n(arg1,,argn) argument position (1 n) where the minimum is reached the first time minprop_n(arg1,,argn) argument position (1 n) where the minimum is reached the first time mod(arg1,arg2) remainder mint(arg) round value nr_of_runs() number of single runs of the current experiment qnt(const_arg,arg) quantile of arg rank(char_arg,arg) rank of arg according to char_arg reg(arg1,arg2) linear regression coefficient to forecast arg2 from arg1 rng(arg) range = max_e(arg) - min_e(arg) run(char_arg,arg) values of arg for a single run selected by char_arg sens_abs(char_arg,arg) absolute sensitivity measure sens_rel(char_arg,arg) sign of value sin(arg) sine sin(arg) sine sin(arg) sine sin(arg) skewness (3 <sup>rd</sup> moment) sqt(arg) square root saic statistical summaries sum(arg) argument sum of values	min_e(arg)	
minprop(arg) index of the element where the minimum is reached the first time minprop_e(arg) run number where the minimum is reached the first time minprop_l(char_arg,arg) dimension-related argument position (1 n) where the minimum is reached the first time minprop_n(arg1,,argn) argument position (1 n) where the minimum is reached the first time mod(arg1,arg2) remainder nint(arg) round value nr_of_runs() number of single runs of the current experiment qnt(const_arg,arg) quantile of arg rank (char_arg,arg) rank of arg according to char_arg reg(arg1,arg2) linear regression coefficient to forecast arg2 from arg1 rng(arg) range = max_e(arg) - min_e(arg) run(char_arg,arg) values of arg for a single run selected by char_arg sens_abs(char_arg,arg) absolute sensitivity measure sens_rel(char_arg,arg) relative sensitivity measure sign(arg) sign of value sin(arg) sine sinh(arg) hyperbolic sine skw(arg) skewness (3 <sup>rd</sup> moment) sqrt(arg) square root stat(const_arg1,const_arg2, const_arg3,const_arg4,arg5) sum(arg) argument sum of values		dimension-related argument minima of values
minprop(arg) index of the element where the minimum is reached the first time minprop_e(arg) run number where the minimum is reached the first time minprop_l(char_arg,arg) dimension-related argument position (1 n) where the minimum is reached the first time minprop_n(arg1,,argn) argument position (1 n) where the minimum is reached the first time mod(arg1,arg2) remainder nint(arg) round value nr_of_runs() number of single runs of the current experiment qnt(const_arg,arg) quantile of arg rank (char_arg,arg) rank of arg according to char_arg reg(arg1,arg2) linear regression coefficient to forecast arg2 from arg1 rng(arg) range = max_e(arg) - min_e(arg) run(char_arg,arg) values of arg for a single run selected by char_arg sens_abs(char_arg,arg) absolute sensitivity measure sens_rel(char_arg,arg) relative sensitivity measure sign(arg) sign of value sin(arg) sine sinh(arg) hyperbolic sine skw(arg) skewness (3 <sup>rd</sup> moment) sqrt(arg) square root stat(const_arg1,const_arg2, const_arg3,const_arg4,arg5) sum(arg) argument sum of values	min n(arg1,,argn)	minimum per element
minprop_e(arg) minprop_l(char_arg,arg) dimension-related argument position (1 n) where the minimum is reached the first time minprop_n(arg1,,argn) minprop_n(arg1,,argn) minprop_n(arg1,,argn) minprop_n(arg1,,argn) minprop_n(arg1,,argn) minprop_n(arg1,,argn) mod(arg1,arg2) remainder mod(arg1,arg2) remainder mint(arg) round value nr_of_runs() number of single runs of the current experiment qnt(const_arg,arg) quantile of arg rank of arg according to char_arg reg(arg1,arg2) linear regression coefficient to forecast arg2 from arg1 rng(arg) range = max_e(arg) - min_e(arg) run(char_arg,arg) values of arg for a single run selected by char_arg sens_abs(char_arg,arg) sens_abs(char_arg,arg) relative sensitivity measure sign(arg) sign of value sin(arg) sine sinh(arg) shw(arg) skewness (3 <sup>rd</sup> moment) sqrt(arg) square root stat(const_arg1,const_arg2, const_arg3,const_arg4,arg5) sum(arg) argument sum of values		
minprop_l(char_arg,arg) minprop_n(arg1,,argn) minprop_n(arg1,,argn) mod(arg1,arg2) remainder mint(arg) round value nr_of_runs() number of single runs of the current experiment quantile of arg rank(char_arg,arg) reg(arg1,arg2) linear regression coefficient to forecast arg2 from arg1 rng(arg) run(char_arg,arg) run(char_arg,arg) values of arg for a single run selected by char_arg sens_abs(char_arg,arg) relative sensitivity measure sens_rel(char_arg,arg) sinh(arg) sinh(arg) square root stat(const_arg1,const_arg2, const_arg3,const_arg4,arg5) sum(arg) argument sum of values sirunder argument position (1 n) where the minimum is reached the first time argument position (1 n) where the minimum is reached the first time argument position (1 n) where the minimum is reached the first time argument position (1 n) where the minimum is reached the first time argument position (1 n) where the minimum is reached the first time argument position (1 n) where the minimum is reached the first time argument position (1 n) where the minimum is reached the first time argument position (1 n) where the minimum is reached the first time argument sylver the minimum is reached the first time argument sylver the minimum is reached the first time argument sylver the minimum is reached the first time argument sylver the minimum is reached the first time argument sylver the minimum is reached the first time argument sylver the minimum is reached the first time argument sylver argument sylver the minimum is reached the first time argument sylver argume		run number where the minimum is reached the first time
reached the first time  minprop_n(arg1,,argn) argument position (1 n) where the minimum is reached the first time  mod(arg1,arg2) remainder  nint(arg) round value  nr_of_runs() number of single runs of the current experiment  qnt(const_arg,arg) quantile of arg  rank (char_arg,arg) rank of arg according to char_arg  reg(arg1,arg2) linear regression coefficient to forecast arg2 from arg1  rng(arg) range = max_e(arg) - min_e(arg)  run(char_arg,arg) values of arg for a single run selected by char_arg  sens_abs(char_arg,arg) absolute sensitivity measure  sens_rel(char_arg,arg) relative sensitivity measure  sign(arg) sign of value  sin(arg) sine  sinh(arg) hyperbolic sine  skw(arg) skewness (3 <sup>rd</sup> moment)  sqrt(arg) square root  stat(const_arg1,const_arg2, const_arg3,const_arg4,arg5)  sum(arg) argument sum of values		dimension-related argument position (1 n) where the minimum is
mod(arg1,arg2)remaindernint(arg)round valuenr_of_runs()number of single runs of the current experimentqnt(const_arg,arg)quantile of argrank(char_arg,arg)rank of arg according to char_argreg(arg1,arg2)linear regression coefficient to forecast arg2 from arg1rng(arg)range = max_e(arg) - min_e(arg)run(char_arg,arg)values of arg for a single run selected by char_argsens_abs(char_arg,arg)absolute sensitivity measuresens_rel(char_arg,arg)relative sensitivity measuresign(arg)sign of valuesin(arg)sinesinh(arg)hyperbolic sineskw(arg)skewness (3rd moment)sqrt(arg)square rootstat(const_arg1,const_arg2, const_arg3,const_arg4,arg5)basic statistical summariessum(arg)argument sum of values		
nint(arg)round valuenr_of_runs()number of single runs of the current experimentqnt(const_arg,arg)quantile of argrank(char_arg,arg)rank of arg according to char_argreg(arg1,arg2)linear regression coefficient to forecast arg2 from arg1rng(arg)range = max_e(arg) - min_e(arg)run(char_arg,arg)values of arg for a single run selected by char_argsens_abs(char_arg,arg)absolute sensitivity measuresens_rel(char_arg,arg)relative sensitivity measuresign(arg)sign of valuesin(arg)sinesinh(arg)hyperbolic sineskw(arg)skewness (3rd moment)sqrt(arg)square rootstat(const_arg1,const_arg2, const_arg3,const_arg4,arg5)basic statistical summaries	minprop_n(arg1,,argn)	argument position (1 n) where the minimum is reached the first time
nr_of_runs() number of single runs of the current experiment qnt(const_arg,arg) quantile of arg rank(char_arg,arg) reg(arg1,arg2) reg(arg1,arg2) linear regression coefficient to forecast arg2 from arg1 rng(arg) run(char_arg,arg) values of arg for a single run selected by char_arg sens_abs(char_arg,arg) sens_rel(char_arg,arg) sign of value sin(arg) sine sinh(arg) sine sinh(arg) skw(arg) skw(arg) stat(const_arg1,const_arg2, const_arg3,const_arg4,arg5) sum(arg) argument sum of values	mod(arg1,arg2)	remainder
qnt(const_arg,arg) quantile of arg rank(char_arg,arg) rank of arg according to char_arg reg(arg1,arg2) linear regression coefficient to forecast arg2 from arg1 rng(arg) range = max_e(arg) - min_e(arg) run(char_arg,arg) values of arg for a single run selected by char_arg sens_abs(char_arg,arg) absolute sensitivity measure sens_rel(char_arg,arg) relative sensitivity measure sign(arg) sign of value sin(arg) sine sinh(arg) hyperbolic sine skw(arg) skewness (3 <sup>rd</sup> moment) sqrt(arg) square root stat(const_arg1,const_arg2, const_arg3,const_arg4,arg5) sum(arg) argument sum of values		round value
rank (char_arg,arg) rank of arg according to char_arg reg(arg1,arg2) linear regression coefficient to forecast arg2 from arg1 rng(arg) range = max_e(arg) - min_e(arg) run(char_arg,arg) values of arg for a single run selected by char_arg sens_abs(char_arg,arg) absolute sensitivity measure sens_rel(char_arg,arg) relative sensitivity measure sign(arg) sign of value sin(arg) sine sinh(arg) hyperbolic sine skw(arg) skewness (3 <sup>rd</sup> moment) sqrt(arg) square root stat(const_arg1,const_arg2, const_arg3,const_arg4,arg5) sum(arg) argument sum of values	nr_of_runs()	number of single runs of the current experiment
reg(arg1,arg2) linear regression coefficient to forecast arg2 from arg1 rng(arg) range = max_e(arg) - min_e(arg) run(char_arg,arg) values of arg for a single run selected by char_arg sens_abs(char_arg,arg) absolute sensitivity measure sens_rel(char_arg,arg) relative sensitivity measure sign(arg) sign of value sin(arg) sine sinh(arg) hyperbolic sine skw(arg) skewness (3 <sup>rd</sup> moment) sqrt(arg) square root stat(const_arg1,const_arg2, const_arg3,const_arg4,arg5) sum(arg) argument sum of values	qnt(const_arg,arg)	
rng(arg) range = max_e(arg) - min_e(arg) run(char_arg,arg) values of arg for a single run selected by char_arg sens_abs(char_arg,arg) absolute sensitivity measure sens_rel(char_arg,arg) relative sensitivity measure sign(arg) sign of value sin(arg) sine sinh(arg) hyperbolic sine skw(arg) skewness (3 <sup>rd</sup> moment) sqrt(arg) square root stat(const_arg1,const_arg2, const_arg3,const_arg4,arg5) sum(arg) argument sum of values	rank(char_arg,arg)	
run(char_arg,arg) values of arg for a single run selected by char_arg sens_abs(char_arg,arg) absolute sensitivity measure sens_rel(char_arg,arg) relative sensitivity measure sign(arg) sign of value sin(arg) sine sinh(arg) hyperbolic sine skw(arg) skewness (3 <sup>rd</sup> moment) sqrt(arg) square root stat(const_arg1,const_arg2, const_arg3,const_arg4,arg5) sum(arg) argument sum of values	reg(arg1,arg2)	
sens_abs(char_arg,arg)absolute sensitivity measuresens_rel(char_arg,arg)relative sensitivity measuresign(arg)sign of valuesin(arg)sinesinh(arg)hyperbolic sineskw(arg)skewness (3rd moment)sqrt(arg)square rootstat(const_arg1,const_arg2, const_arg3,const_arg4,arg5)basic statistical summariessum(arg)argument sum of values	rng(arg)	
sens_rel(char_arg,arg)       relative sensitivity measure         sign(arg)       sign of value         sin(arg)       sine         sinh(arg)       hyperbolic sine         skw(arg)       skewness (3 <sup>rd</sup> moment)         sqrt(arg)       square root         stat(const_arg1,const_arg2, const_arg3,const_arg4,arg5)       basic statistical summaries         sum(arg)       argument sum of values	run(char_arg,arg)	values of arg for a single run selected by char_arg
sign(arg)       sign of value         sin(arg)       sine         sinh(arg)       hyperbolic sine         skw(arg)       skewness (3 <sup>rd</sup> moment)         sqrt(arg)       square root         stat(const_arg1,const_arg2, const_arg3,const_arg4,arg5)       basic statistical summaries         sum(arg)       argument sum of values	sens_abs(char_arg,arg)	absolute sensitivity measure
sin(arg)sinesinh(arg)hyperbolic sineskw(arg)skewness (3 <sup>rd</sup> moment)sqrt(arg)square rootstat(const_arg1,const_arg2, const_arg3,const_arg4,arg5)basic statistical summariessum(arg)argument sum of values	sens_rel(char_arg,arg)	relative sensitivity measure
sinh(arg)hyperbolic sineskw(arg)skewness (3 <sup>rd</sup> moment)sqrt(arg)square rootstat(const_arg1,const_arg2, const_arg3,const_arg4,arg5)basic statistical summariessum(arg)argument sum of values	sign(arg)	sign of value
skw(arg)       skewness (3 <sup>rd</sup> moment)         sqrt(arg)       square root         stat(const_arg1,const_arg2, const_arg3,const_arg4,arg5)       basic statistical summaries         sum(arg)       argument sum of values	sin(arg)	sine
sqrt(arg)       square root         stat(const_arg1,const_arg2, const_arg3,const_arg4,arg5)       basic statistical summaries         sum(arg)       argument sum of values	sinh(arg)	
sqrt(arg)square rootstat(const_arg1,const_arg2, const_arg3,const_arg4,arg5)basic statistical summariessum(arg)argument sum of values	skw(arg)	skewness (3 <sup>rd</sup> moment)
const_arg3,const_arg4,arg5) sum(arg) argument sum of values		square root
sum(arg) argument sum of values		basic statistical summaries
		argument sum of values
1 - 1 - 1 - 1 - 1 - 1 - 1 - 1 - 1 - 1 -	sum_e(arg)	run ensemble sum

Name	Meaning
sum_l(char_arg,arg)	dimension-related argument sums of values
sym_abs(char_arg,arg)	absolute symmetry measure
sym_rel(char_arg,arg)	relative symmetry measure
tan(arg)	tangent
tanh(arg)	hyperbolic tangent
transpose(char_arg,arg)	transpose arg according to char_arg
undef()	undefined element
var(arg)	argument variance of values
var_e(arg)	run ensemble variance
var_l(char_arg,arg)	dimension-related argument variances of values

 Tab. 15.7
 Post-processor built-in operators (in alphabetical order)



## 15.3.3 Character Arguments of Post-Processor Built-in Operators

Tab. 15.8 summarises for built-in operators character argument values. User-defined operators can not have pre-defined character argument values.

Operator	Argument number	Argument value (without quotation marks, pre-defined values are case-insensitive)	
avg_l	1	sequence of digits 0 and 1	(**)
avgg_l	1	sequence of digits 0 and 1 (	
avgh_I	1	sequence of digits 0 and 1 (**	
avgw_l	1	sequence of digits 0 and 1 (*	
behav	1	(not pre-defined, case insensitive) (*	
clip	1	(not pre-defined, case insensitive)	
count	1	[ all   def   undef ]	
count_e	1	all   def   undef ]	
count_I	1	sequence of digits 0 and 1 (**)	
count_I	2	[ all   def   undef ]	
cumul	1	sequence of digits 0 and 1 (**)	
get_experiment	1	(not pre-defined, case sensitive)	
get_experiment	2	(not pre-defined, case insensitive)	
get_experiment	3	(not pre-defined, case sensitive)	
get_table_fct	1	(not pre-defined, case sensitive)	
flip	1	sequence of digits 0 and 1	(**)
hgr_l	1	sequence of digits 0 and 1	(**)
if	1	[ <   <=   >   >=   =   !=   def   undef ]	
lin_abs	1	(not pre-defined, case insensitive)	(*)
lin_rel	1	(not pre-defined, case insensitive) (*)	
mask	1	[< <= > >= = !=]	
max_l	1	sequence of digits 0 and 1 (**)	
maxprop_l	1	sequence of digits 0 and 1 (**)	
min_I	1	sequence of digits 0 and 1	(**)
minprop_I	1	sequence of digits 0 and 1	(**)
rank	1	[ tie_plain   tie_min   tie_avg ]	
run	1	[ run number   not pre-defined ]	
sens_abs	1	(not pre-defined, case insensitive)	(*)
sens_rel	1	(not pre-defined, case insensitive)	(*)
sum_l	1	sequence of digits 0 and 1	(**)
sym_abs	1	(not pre-defined, case insensitive) (*)	
sym_rel	1	(not pre-defined, case insensitive) (*)	
transpose	1	sequence of digits 1 to 9 (**)	
var_l	1	sequence of digits 0 and 1	(**)

Tab. 15.8Character arguments of post-processor built-in operators

- (\*) Character argument can be empty
- (\*\*) The length of the character string argument from a sequence of digits corresponds with the dimensionality of the non-character argument under investigation.

16-Jun-2004

# 15.3.4 Constant Arguments of Post-Processor Built-in Operators

Tab. 15.8 summarises for built-in operators constant argument values. User-defined operators can not have pre-defined constant argument values.

Operator	Argument number	Argument type	Argument value
classify	1	integer	[0 ≥2]
cnf	1	real	[0.001   0.01   0.05   0.1 ]
hgr	1	integer	[0 ≥4]
hgr_e	1	integer	[0 ≥4]
hgr_l	2	integer	[0 ≥4]
qnt	1	real	0. ≤ arg1 ≤ 100.
stat	1	real	0. ≤ arg1 ≤ 100.
stat	2	real	0. ≤ arg2 ≤ 100.
stat	3	real	[0.001   0.01   0.05   0.1 ]
stat	4	real	[0.001   0.01   0.05   0.1 ]

 Tab. 15.9
 Constant arguments of post-processor built-in operators



# 15.4 Glossary

The glossary defines terms in that sense they are used in this User's Guide. An arrow  $\rightarrow$  refers to another term in the glossary.

- **Adjustment**: Numerical modification of a  $\rightarrow$  target during an  $\rightarrow$  experiment. Adjustments are related to an  $\rightarrow$  experiment type and are described in the experiment description  $\rightarrow$  user-defined file.
- **ASCII**: The American Standard Code for Information and Interchange developed by the American National Standards Institute (http://www.ansi.org) is used in SimEnv to store information in → user-defined files.
- **Behavioural analysis**: → Experiment type to inspect behaviour of a → model in a space, spanned up by → targets. The target space is scanned in a deterministic manner, applying pre-defined → adjustments of the targets with a flexible scanning strategy for target sub-spaces.
- **Coordinate coord**: In the model description → user-defined file each → dimension of a → variable with a → dimensionality greater than 0 a coordinate is assigned to. A coordinate has a unique name and strictly monotonic ordered coordinate values. The number of coordinate values corresponds with the → extent for this dimension. Consequently, each model variable with a dimensionality greater than 0 resides at a assigned (multi-dimensional) → grid.
- **Coupling**: → model interface
- **Data type**: The type of a → variable as declared in the → model and the corresponding model description → user-defined file. SimEnv data types are byte, short, int, float, and double.
- **Default value**: The nominal (standard) numerical value of an experiment → target. The default value is specified in the experiment description → user-defined file and for → the model interface at the language level also in the model code.
- **Dimension**: → dimensionality
- Dimensionality dim: The number of dimensions of a model → variable or of an → operand in model output post-processing. In the model description → user-defined file each variable a dimensionality is assigned to that corresponds with the dimensionality of the related model output field in the model source code. Dimensionality 0 corresponds to a scalar, dimensionality 1 to a vector, dimensionality 2 to a matrix.
- **Environment variable**: At UNIX operating system level the so called environment is set up as an array of operating-system and user-defined environment variables that have the form Name=Value. The Value of a Name can be addressed by \$Name. In SimEnv directory and path strings or parts of it with environment variables are forbidden.
- **Experiment**: Performing simulation runs with a → model in a co-ordinated manner by applying → experiment types and running the model in a run ensemble, i.e., a series of single simulation runs.
- **Experiment target**: → target
- Experiment type: Pre-defined multi-run simulation experiment. In the process of experiment preparation (defining an experiment by describing it in the experiment description → user-defined file) → targets are assigned to an experiment type and experiment-specific → adjustments and other information are assigned to the targets. Currently available experiment types are → behavioural analysis, → Monte Carlo analysis, and → local sensitivity analysis.
- Extent ext: The number of values for a dimension (from the → dimensionality) of a model → variable or of an → operand in model output post-processing. Extents are always greater than 1. Model variables and operands of dimensionality 0 do not have an extent.
- Fortran column-wise storage model: A rule how to map a multi-dimensional data field to a 1-dimensional vector. A data field field(1:ext<sub>1</sub>, 1:ext<sub>2</sub>, ..., 1:ext<sub>dim</sub>) of  $\rightarrow$  dimensionality dim and  $\rightarrow$  extents ext<sub>1</sub>, ext<sub>2</sub>, ..., ext<sub>dim</sub> is mapped in Fortran in the following way on a 1-dimensional vector vector(1:ext<sub>1</sub>\* ext<sub>2</sub>\* ... \* ext<sub>dim</sub>)

```
\begin{split} & \text{ipointer=0} \\ & \text{do } i_{\text{dim}} = 1 \text{ , ext}_{\text{dim}} \\ & \dots \\ & \text{do } i_2 = 1 \text{ , ext}_2 \\ & \text{do } i_1 = 1 \text{ , ext}_1 \\ & \text{ ipointer = ipointer + 1} \\ & \text{ vector(ipointer) = field(i_1 \text{ , } i_2 \text{ , ... , } i_{\text{dim}})} \\ & \text{ enddo} \\ & \text{enddo} \\ & \text{enddo} \end{split}
```

**GAMS**: The **G**eneral **A**lgebraic **M**odeling **S**ystem (http://www.gams.com) is a high-level modeling system for mathematical programming problems. It consists of a language compiler and a stable of integrated high-performance solvers. GAMS is tailored for complex, large scale modeling applications, and allows to build large maintainable models that can be adapted quickly to new situations.

**Grid**: Regular topological structure for a model → variable or an → operator result in post-processing, spanned up as the Cartesian product of the assigned → coordinates to the variable or the operator result.

IEEE: SimEnv can use on demand for storage of model and post-processor output the Institute of Electrical and Electronics Engineers (http://www.ieee.org) standard number 754 for binary storage of floating point numbers.

**Load Leveler:** The load leveler LoadL is a job management system that handles compute resources at IBM's p655 cluster at PIK.

**Local sensitivity analysis**: → Experiment type with incremental → adjustments of → targets in the neighbourhood of the → default values of the targets. A local sensitivity analysis in SimEnv is always performed independently for all targets involved. During post-processing sensitivity, linearity, and symmetry measures can be determined.

**Macro**: An abbreviation for a unique expression, formed from a chain of operands and operators to apply during post-processing. Macros can be embedded into other expressions and are plugged into the expression during its evaluation. Macros are described in the macro description → user-defined file.

**Model**: A model is a deterministic or stochastic algorithm, implemented in one or a number of computer programs that transforms a sequence of input values (→ targets) into a sequence of output values (→ variables). Normally, inputs are parameters, driving forces, initial values, or boundary values to the model, outputs are state variables of the model. For many cases, the model will be state deterministic, time and space dependent. For SimEnv, the model, its targets and variables are coupled in the process of → interfacing the model to SimEnv.

**Model coupling**: → model interface

**Model interface**: Interfacing a → model to SimEnv means coupling it to SimEnv and enabling finally experimenting with a model within SimEnv. There are coupling interfaces at programming language level for C/C++, Fortran, → Python, and → GAMS. Additionally, models can be interfaced at the shell script level by using shell script syntax elements. For all interface techniques the model has to be wrapped into a shell script.

**Model output post-processing operator**: → operator

Model variable: → variable

Monte Carlo analysis: → Experiment type with pre-single run perturbations of experiment → targets. Each perturbed target a → probability density function pdf with function parameters is assigned to. During the → experiment → adjustments of the targets are realizations from the pdf's using random number techniques. In experiment post-processing statistical measures can be derived from model output of the run ensemble. A prominent statistical measure is the heuristic pdf (histogram) of a model → variable and its relation to the pdf's of the targets.



- NetCDF: Network Common Data Form is an interface for array-oriented data access and a library that provides an implementation of the interface. The NetCDF library also defines a machine-independent format for representing scientific data. Together, the interface, library, and format support the creation, access, and sharing of scientific data. The NetCDF software was developed at the Unidata Program Center in Boulder, Colorado (http://www.unidata.ucar.edu). NetCDF is freely available. SimEnv follows for model output and post-processing output storage the NetCDF Climate and Forecast (CF) metadata convention 1.0-beta4 (http://www.cgd.ucar.edu/cms/eaton/cf-metadata/index.html) and extends it.
- **OpenDX:** The **Open D**ata Explorer OpenDX (http://www.opendx.org) is a uniquely full-featured open source project and software package for the visualization of scientific, engineering and analytical data: Its open system design is built on a standard interface environment. The data model provides users with great flexibility in creating visualizations. OpenDX is based on IBM's Visualization Data Explorer.
- Operand: Argument of an → operator in SimEnv model output post-processing. An operand can be a model → variable, an experiment → target, a constant, a character string, → a macro and an operator.
- Operator: Computational algorithm how to transform the values of a sequence of → operands into a sequence of operator results during model output post-processing. An operator transforms → dimensionality, → extents, and → coordinates from the operands into the corresponding information for the operator result. There are built-in elemental, basic, and advanced operators as well as built-in operators related to specific → experiment types. Additionally, SimEnv offers specification of user-defined operators according to an operator interface. User-defined operators are announced to the system in the operator description → user-defined file.
- Optimization: → Experiment type to minimize a cost function (objective function) over a bounded → target space. In SimEnv a simulated annealing strategy (check chapter 4.5 for explanation) is used to optimize the cost function that is formed from model → variables. Often the cost function represents a distance between model output and reference data to find an optimal point in the target space that fits best the model behaviour with respect to the reference data.

#### **Parallel Operating Environment:** → POE

- **POE:** The **P**arallel **O**perating **E**nvironment POE on IBM's p655 cluster at PIK supplies services to allocate nodes, assign jobs to nodes and launch jobs.
- **Post-processing:** The work step of processing model output data from the whole run ensemble after performing a simulation → experiment. SimEnv post-processing enables navigation in the → target space that is sampled by an experiment as well as construction of additional model output data by application of chains of → operators on model output and reference data.
- **Probability density function pdf**: A probability density function serves to represent a probability distribution in terms of integrals. A probability distribution assigns to every interval of real numbers a probability.
- **Python**: Python (http://www.python.org) is a portable, interpreted, interactive, object-oriented programming language. It incorporates modules, exceptions, dynamic typing, and very high level dynamic data types, and classes.
- **Target**: Element of the input set of a → model. Targets are manipulated numerically during an → experiment. Targets can be addressed in model output post-processing and they have there a → dimensionality of zero.

#### **Target adjustment**: → adjustment

- **User-defined files**: A set of → ASCII files to describe → model-, → experiment-, → operator-, → macro-, and → GAMS model specific information and to determine general SimEnv settings. All user-defined files follow the same syntax rules.
- Variable: Element of the output set of a → model that is stored in a SimEnv model or post-processor output format. Variables are defined in the model as well as in the appropriate model description → user file. Each variable has a unique → data type, a → dimensionality, → extents and an assigned → grid. Normally, variable consists of a series of values, forming fields.

White spaces: → ASCII characters space (blank) and horizontal tabulator used in → user-defined files or within expressions in model output post-processing.

Working directory: The directory, a SimEnv service was started from.