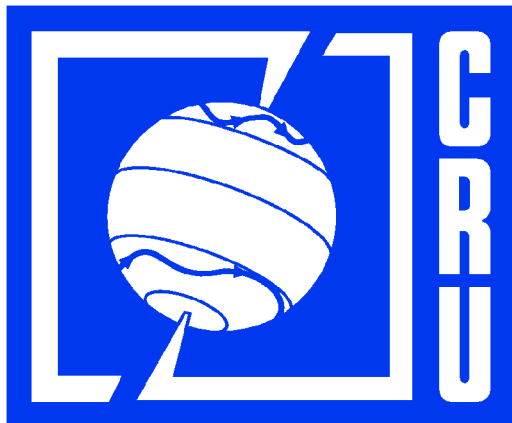


Integrated Modelling of Climate Change and its Impacts



Thomas Kleinen
Climatic Research Unit
14/03/2007

Outline

1. The paradigms of integrated assessment modelling
 - Policy evaluation modelling
 - Policy optimisation modelling
 - Policy guidance modelling
2. Uncertainty in integrated assessment
 - The probabilistic Tolerable Windows Approach
3. Modelling impacts of climate change
 - Changes in flooding probability



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The general problem

- Aim of Integrated Assessment (IA):
 - Consider the **entire** chain of cause-and-effect of climate change



The general problem

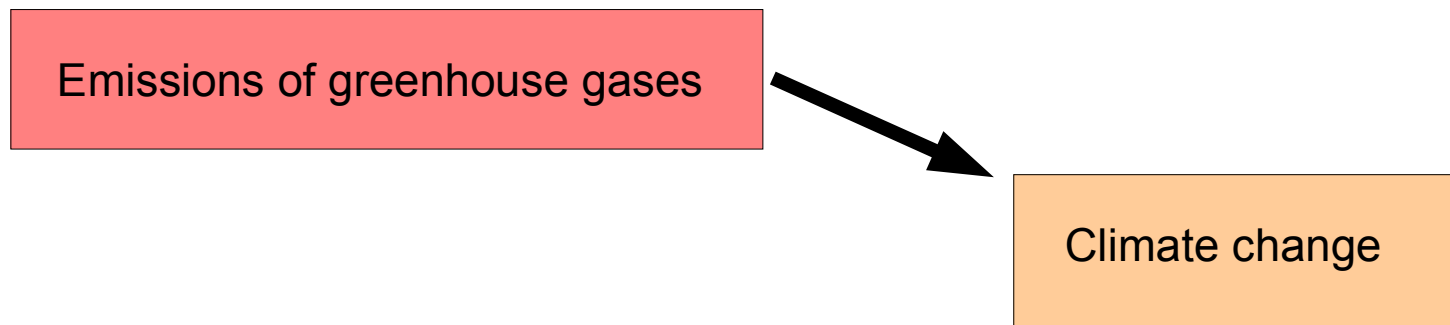
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Emissions of greenhouse gases



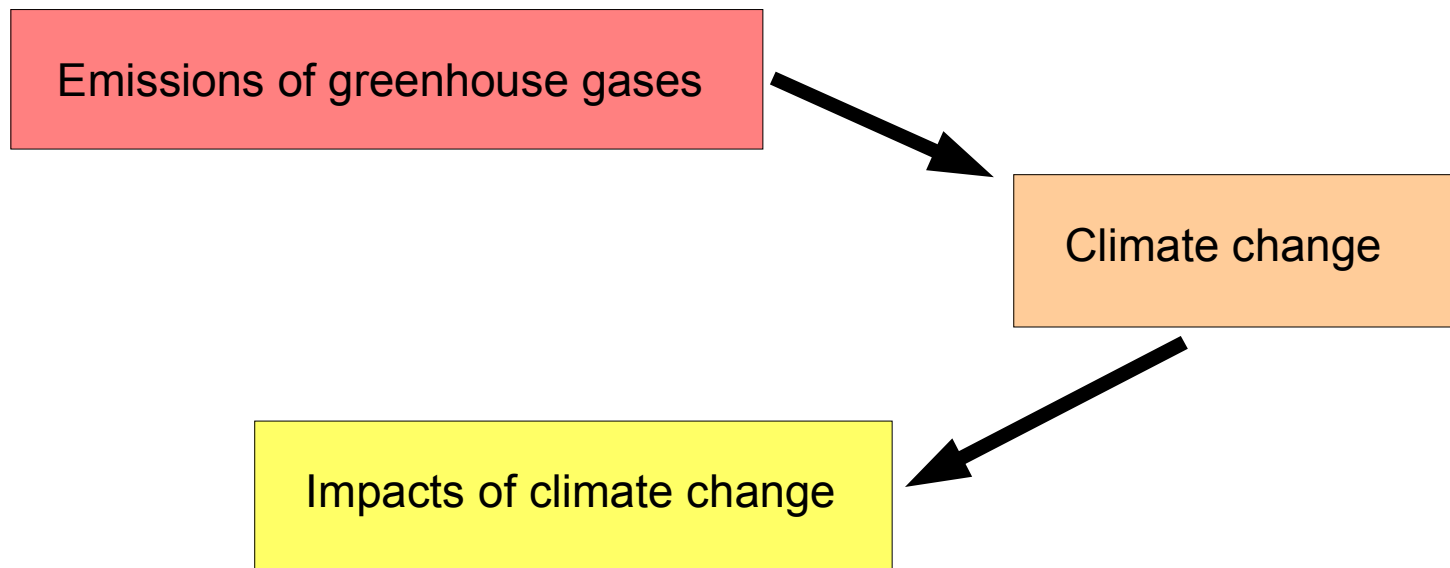
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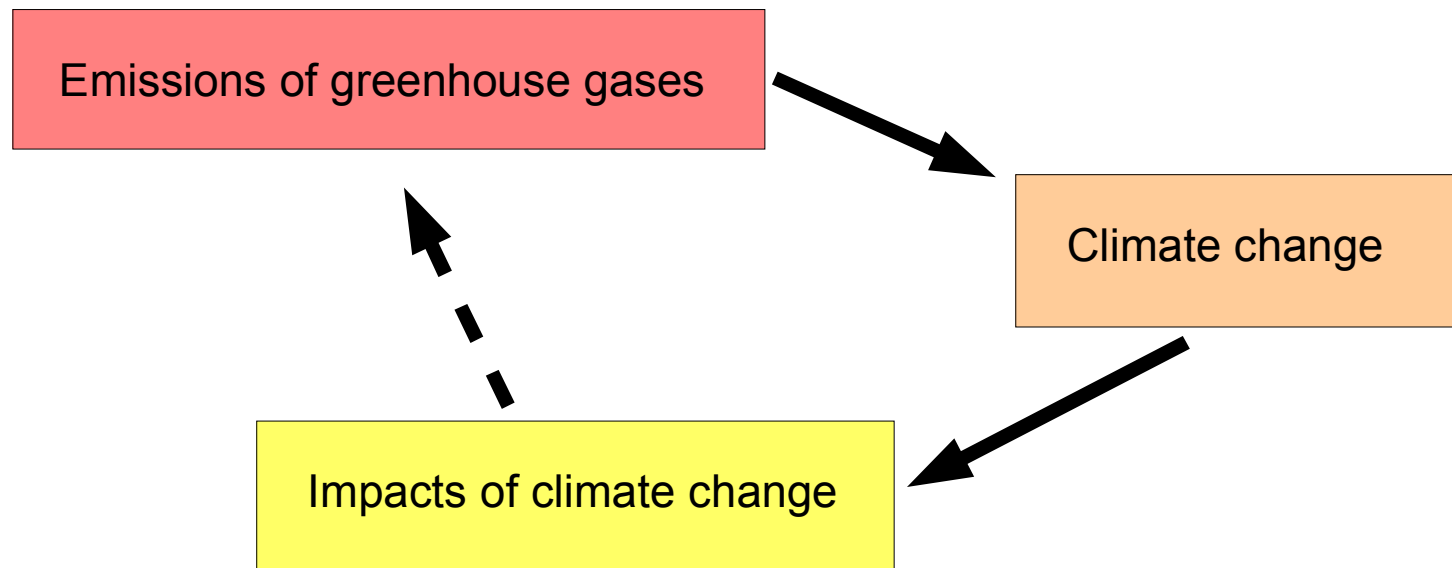
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- Aim of Integrated Assessment (IA):
 - Consider the **entire** chain of cause-and-effect of climate change



- Assessment conducted in **integrated** framework

Three paradigms

$$\dot{x} = f(x, t)$$



Three paradigms

$$\dot{x} = f(x, t; u)$$



Three paradigms

- Mathematically integrated assessment is a *control problem*

$$\dot{\mathbf{x}} = \mathbf{f}(\mathbf{x}, t; \mathbf{u})$$

- Evolution of system state \mathbf{x} also depends on *control vector* \mathbf{u}
- Three general approaches to handle this kind of problem:



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 - *Policy evaluation modelling*
 - 2) Determine “best” control path \mathbf{u}
 - *Policy optimisation modelling*
 - 3) Determine sets of control paths that conform to additional criteria
 - *Policy guidance modelling*



Policy evaluation modelling

- The general approach:
 - Predefine control path, e.g. GHG emissions, investment decisions, R&D
 - Evaluate consequences
- Example: IMAGE family of IA models, e.g. Rotmans et al. 1990

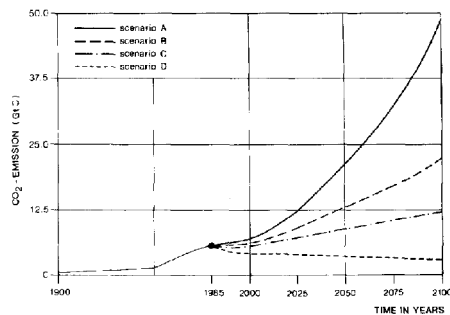


Fig. 2. Emission of CO₂.

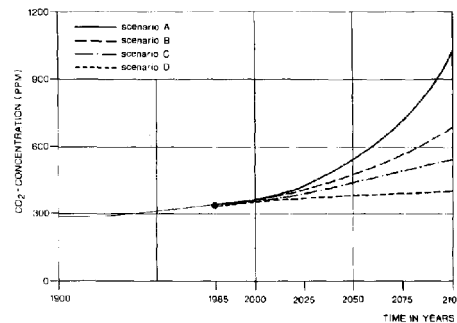


Fig. 8. Concentration of CO₂.

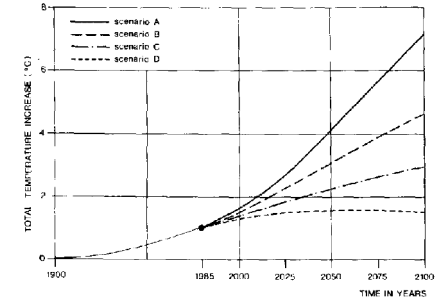


Fig. 13. Total temperature increase.

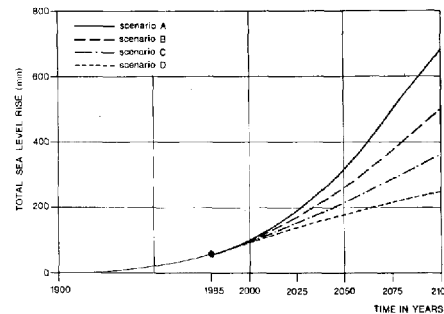
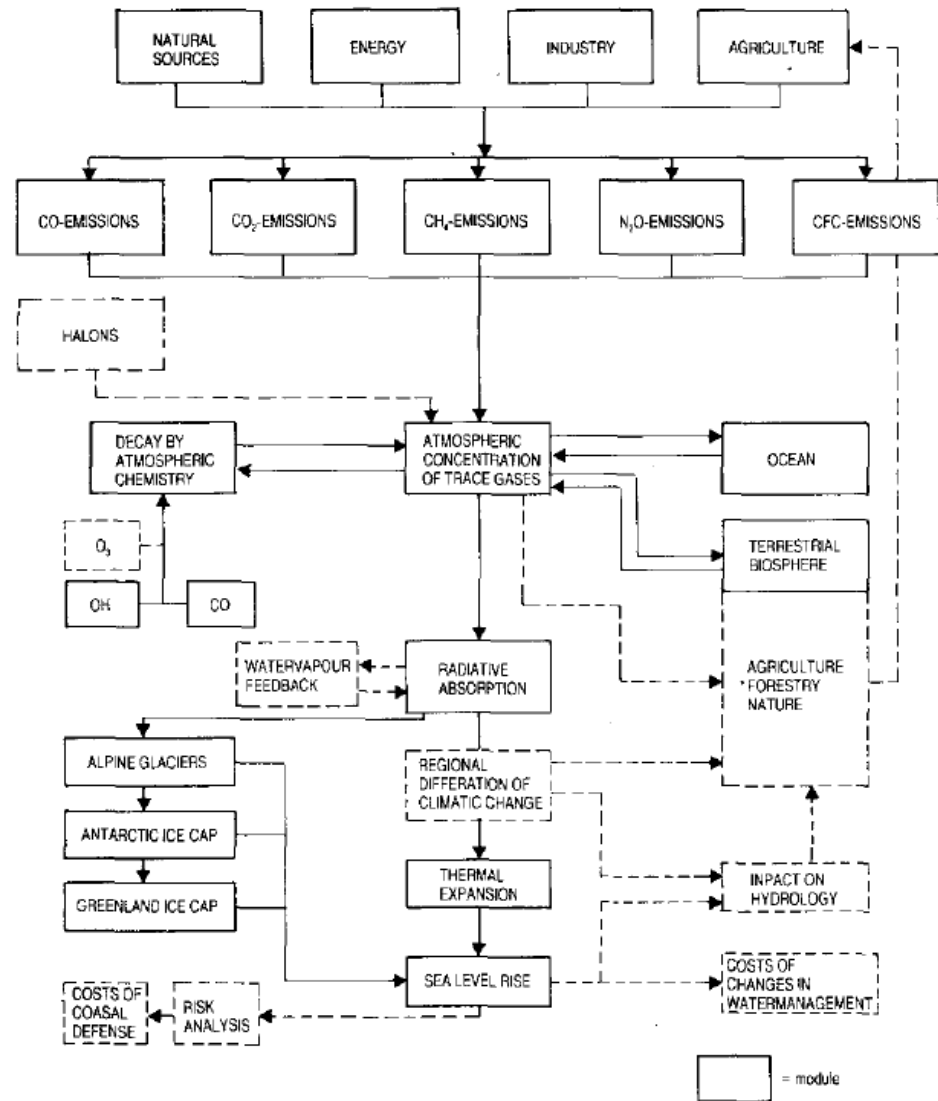
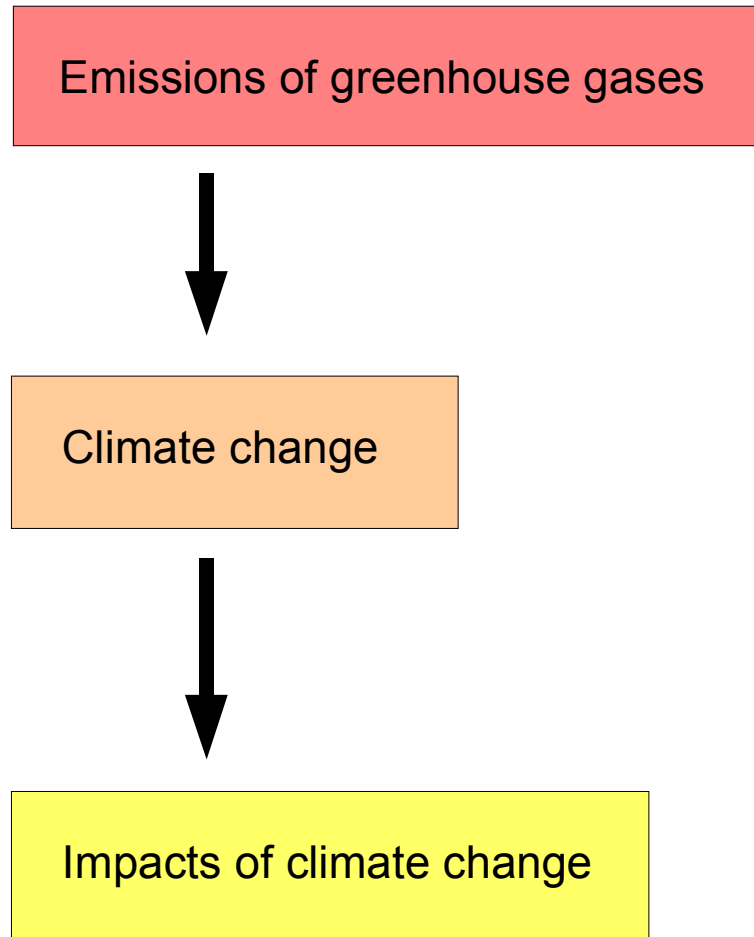


Fig. 15. Total sea level rise.

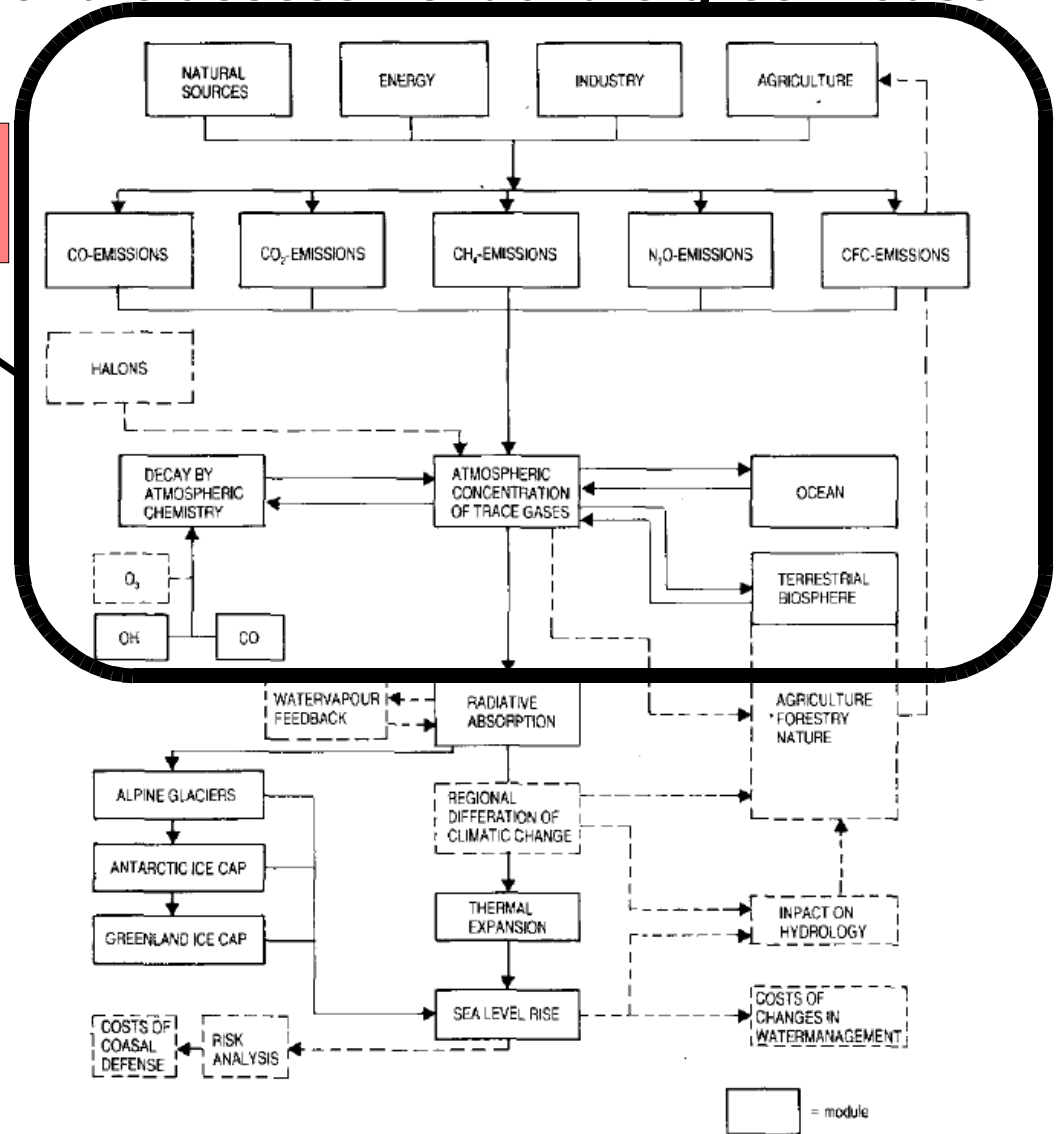
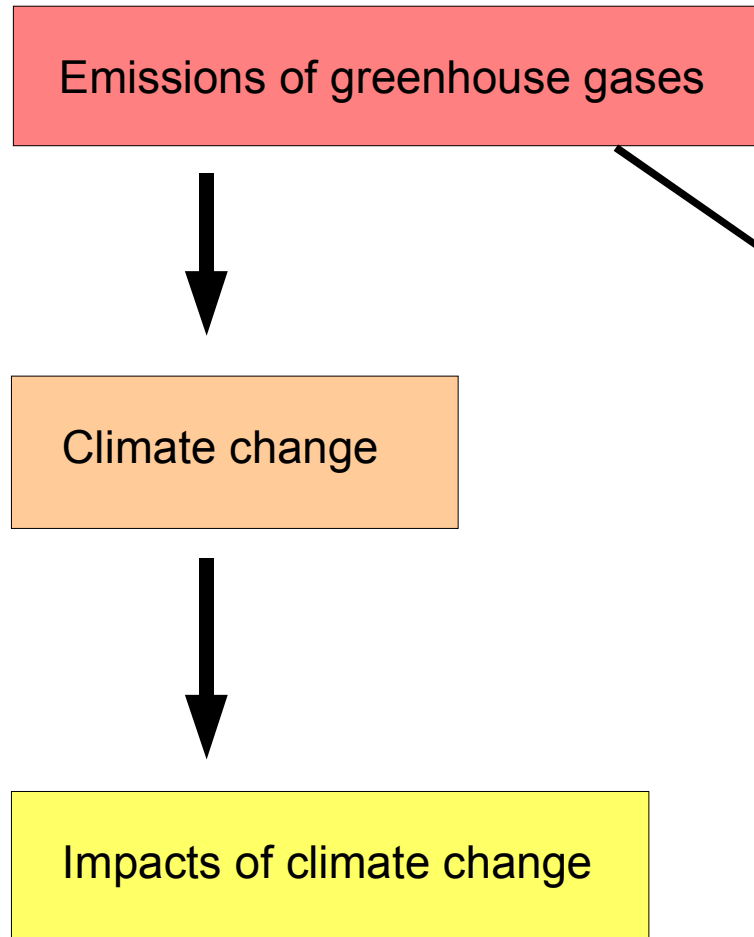
The IMAGE model

- IMAGE = Integrated model for the assessment of the greenhouse effect



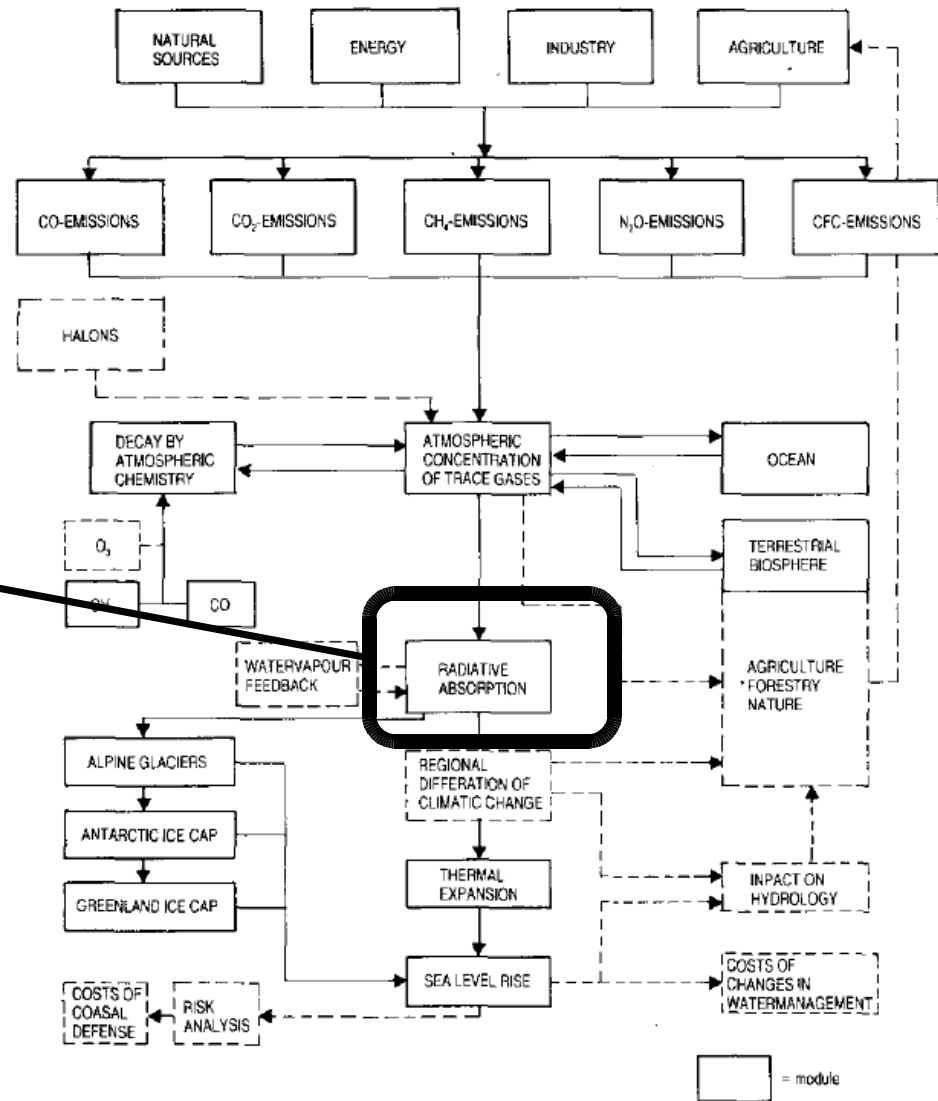
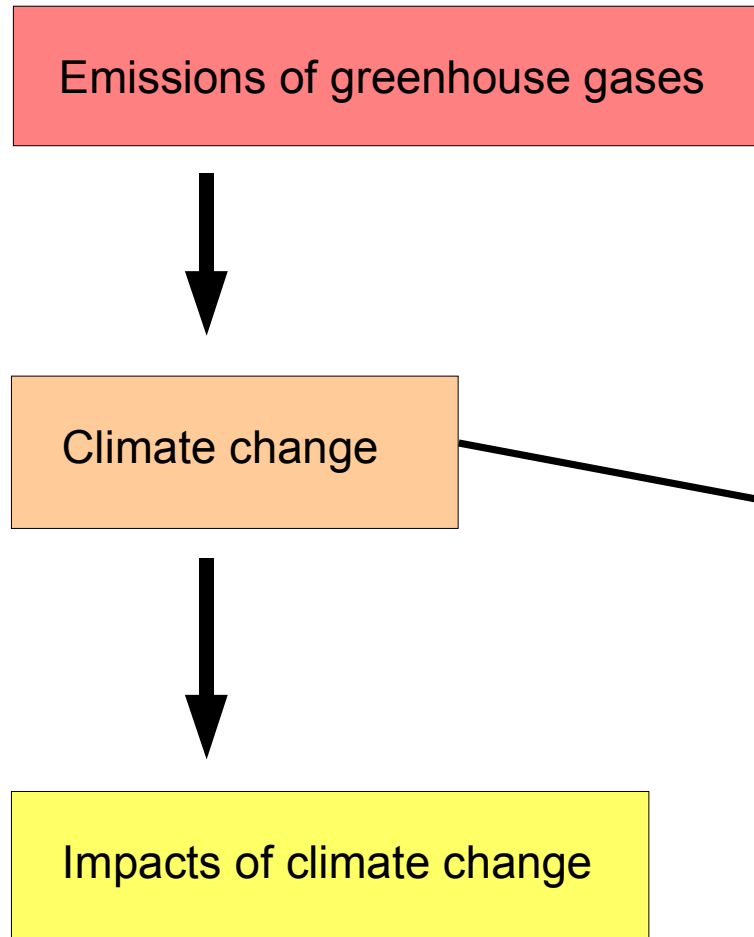
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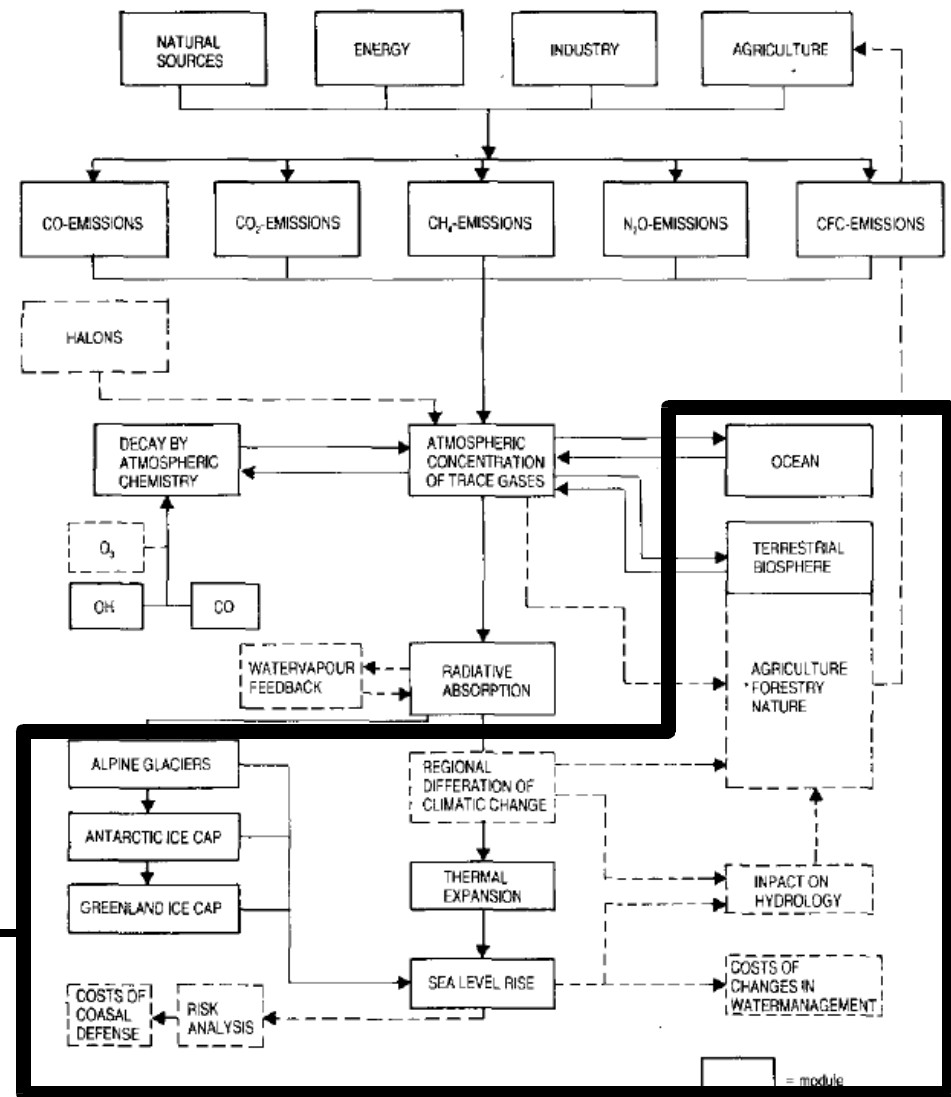
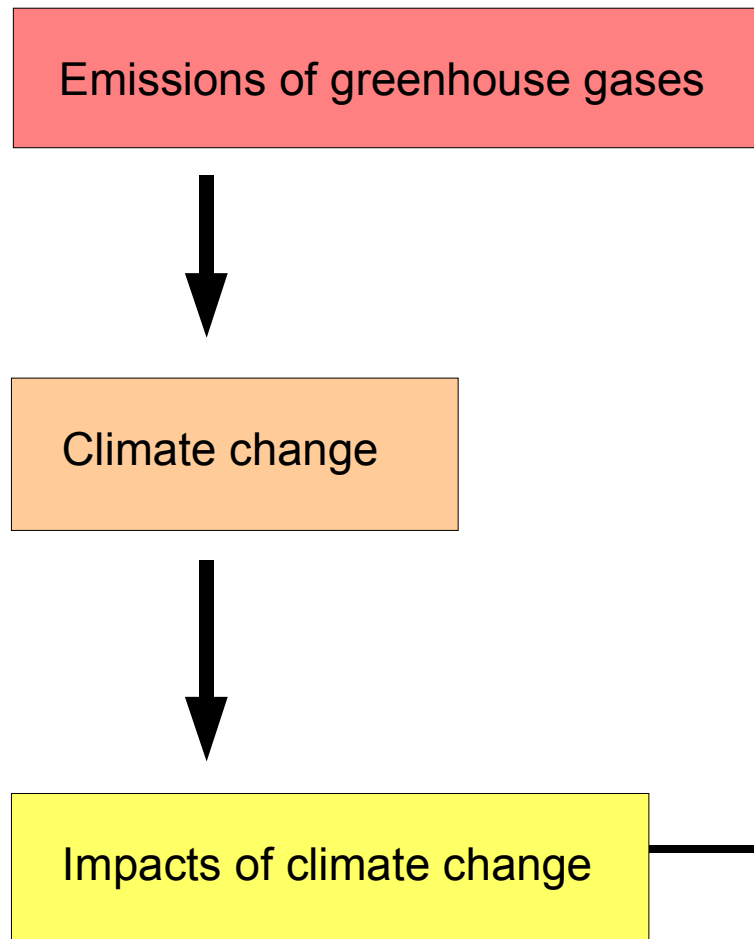
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Policy evaluation modelling

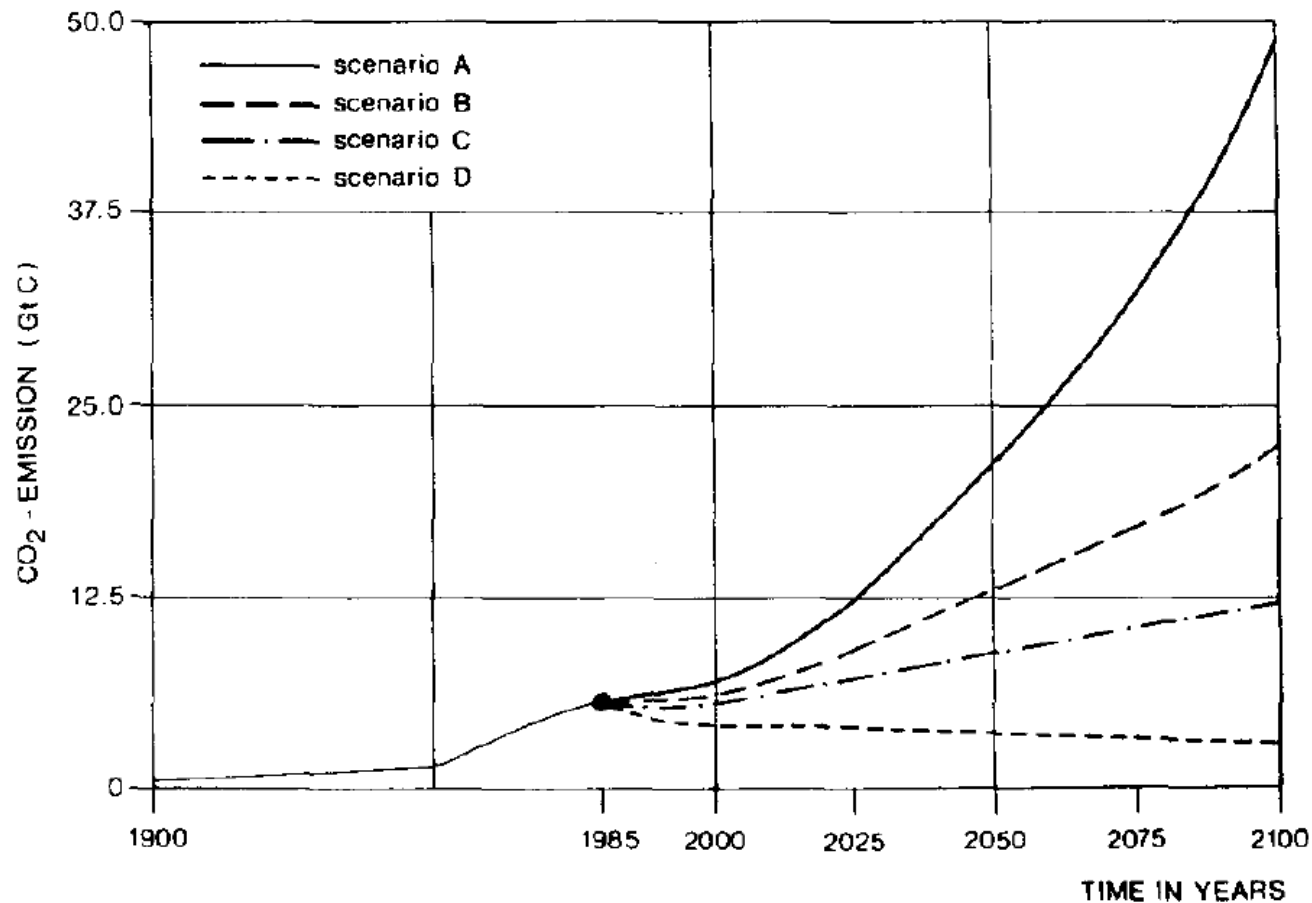


Fig. 2. Emission of CO₂.

Rotmans et al. 1990

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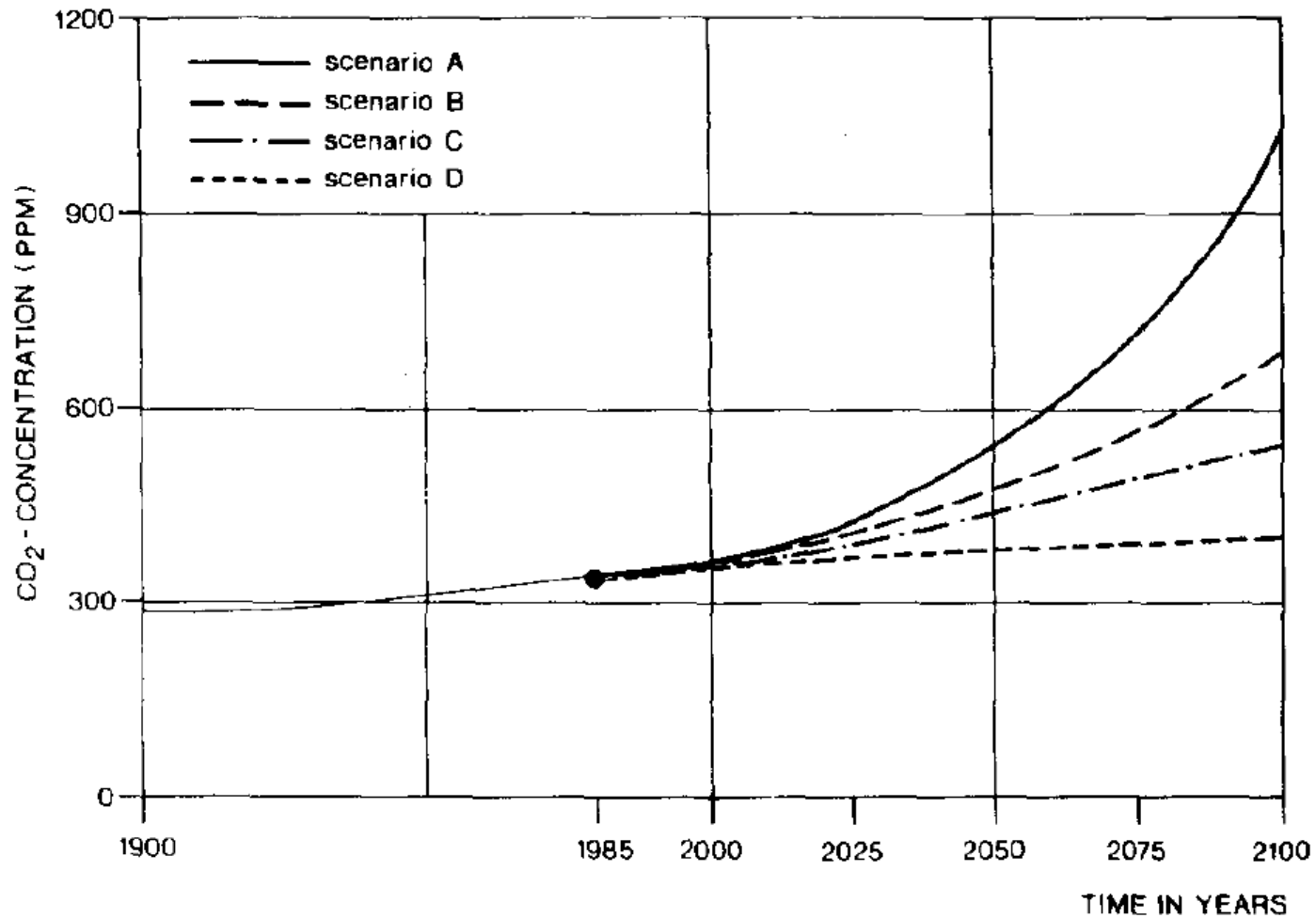


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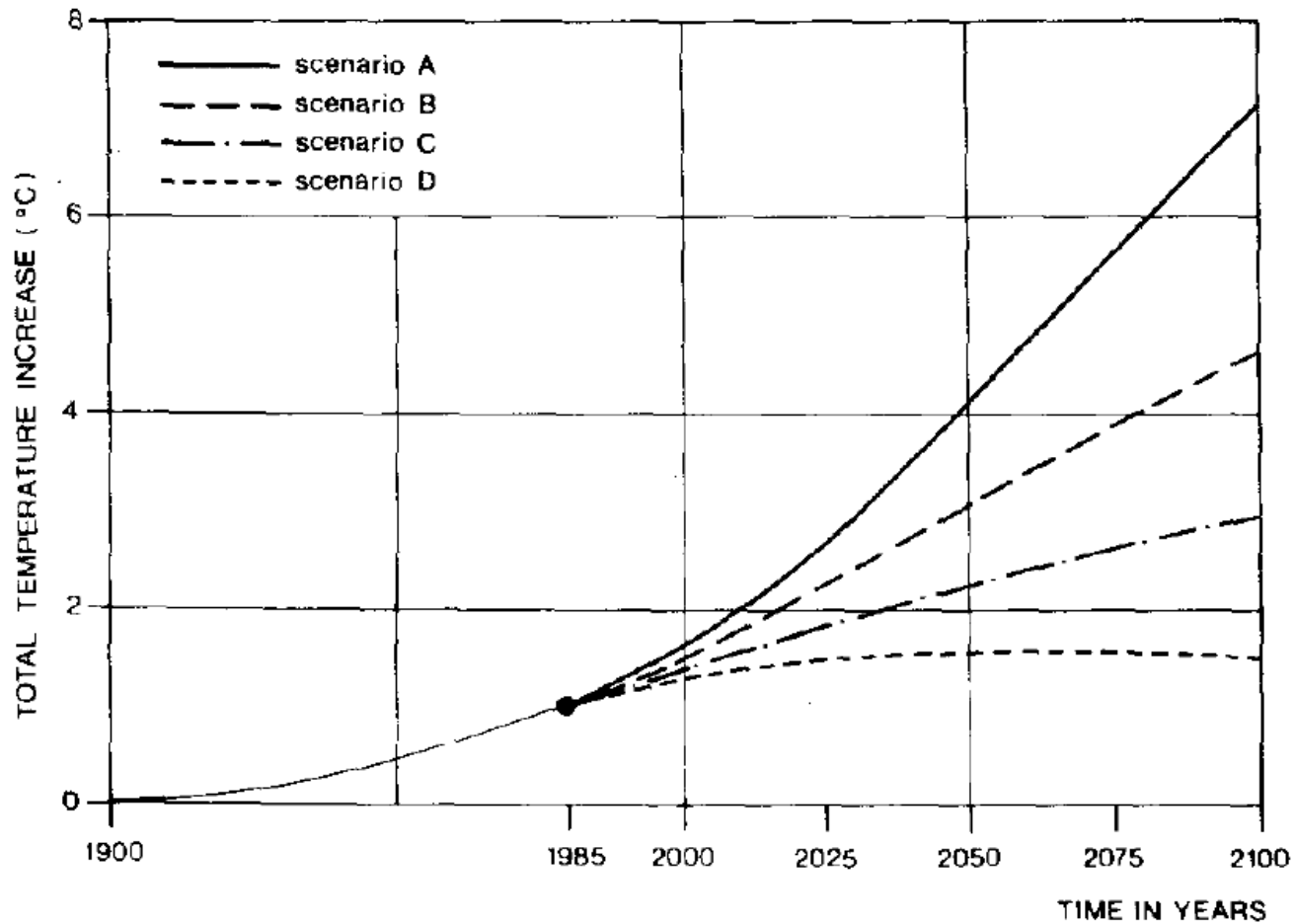


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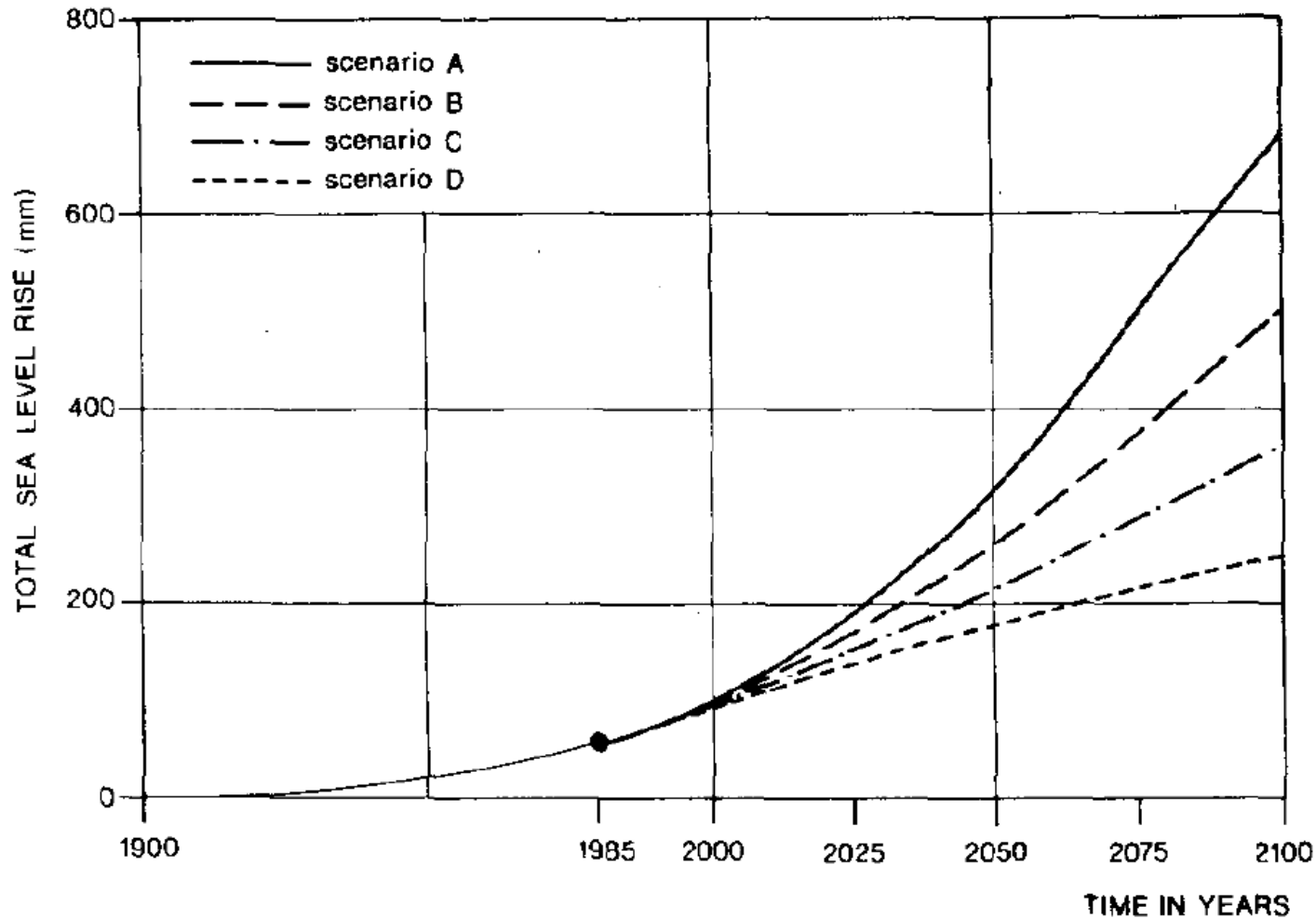


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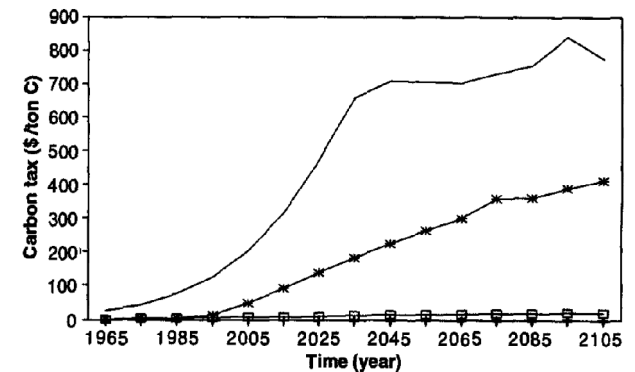
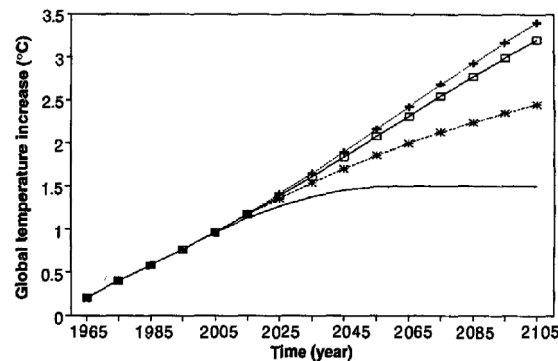
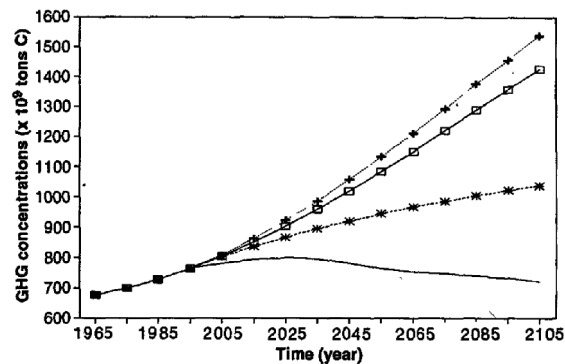
Policy evaluation modelling II

- Advantages:
 - Allows use of process-based models well established in natural sciences
 - High resolution possible, very detailed assessment
 - Any impact(s) that can be described by a model can be considered
- Disadvantages:
 - Search for policy recommendation by trial and error



Policy optimisation modelling

- Aim: determine *optimal* control path
- Two flavors: cost-benefit analysis (CBA) and cost-effectiveness analysis (CEA)
- CEA: Determine cost-efficient controls to reach target
- CBA: Determine control path that maximizes global welfare while considering costs and benefits of climate change
- Example: DICE / RICE models, Nordhaus 1992



The DICE model

- DICE = **D**ynamic **i**ntegrated **c**limate **e**conomy

Emissions of greenhouse gases



Climate change



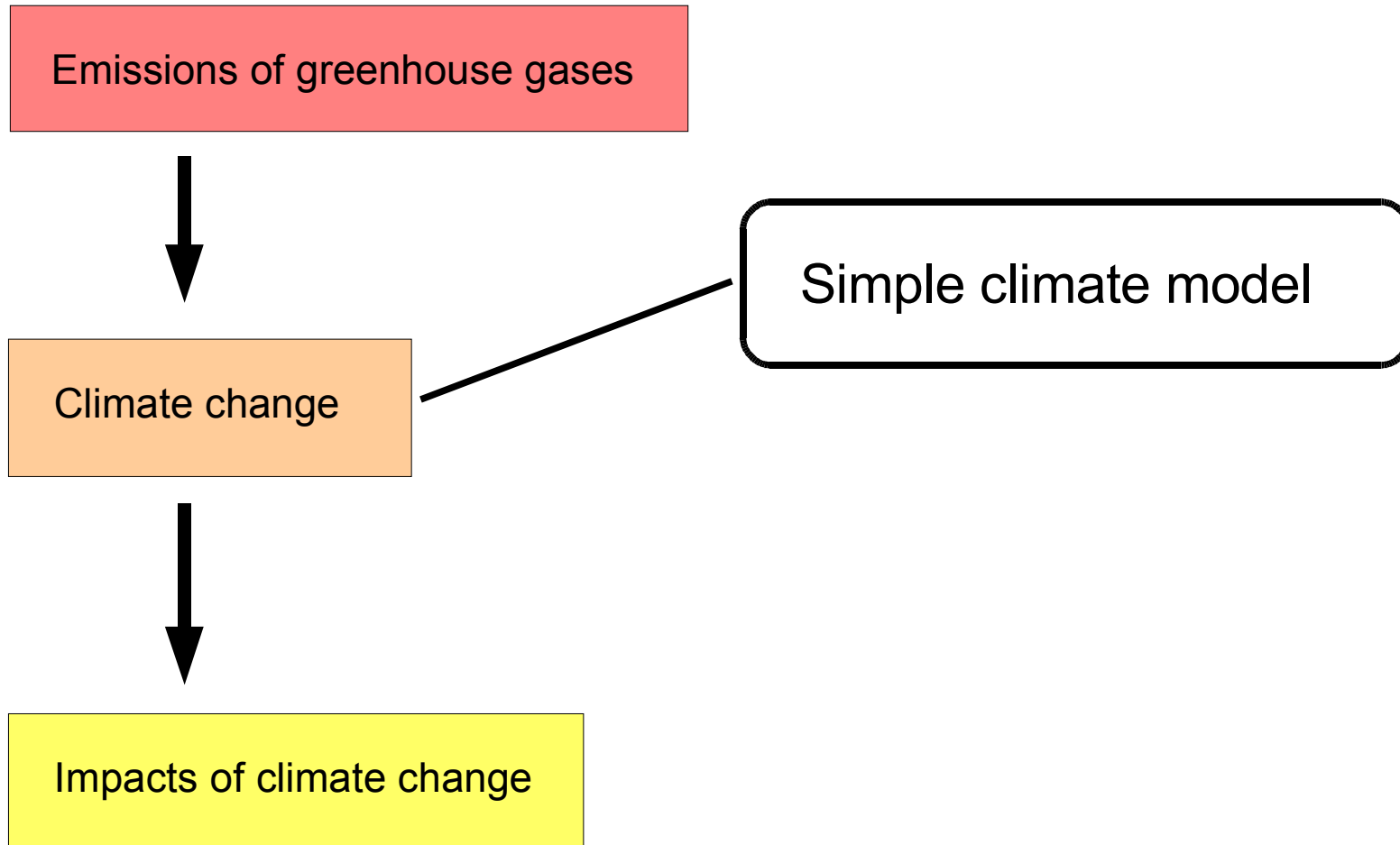
Impacts of climate change

- Economic submodel
 - Economic utility over time
 - Utility dependent on economic output
 - Again dependent on capital, labor, investment
 - Emissions dependent on Output
 - Cost of GHG reductions: simple function, obtained from studies



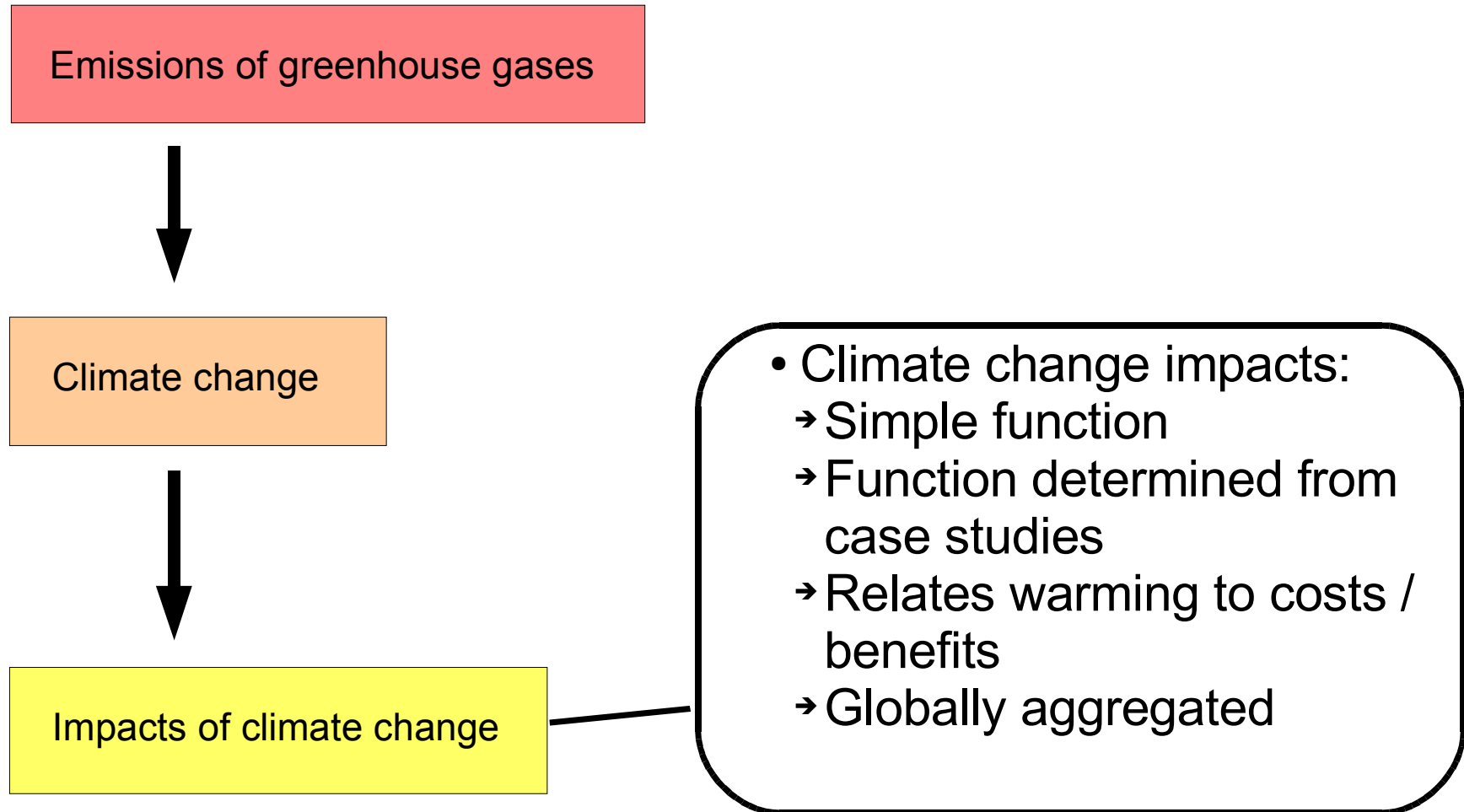
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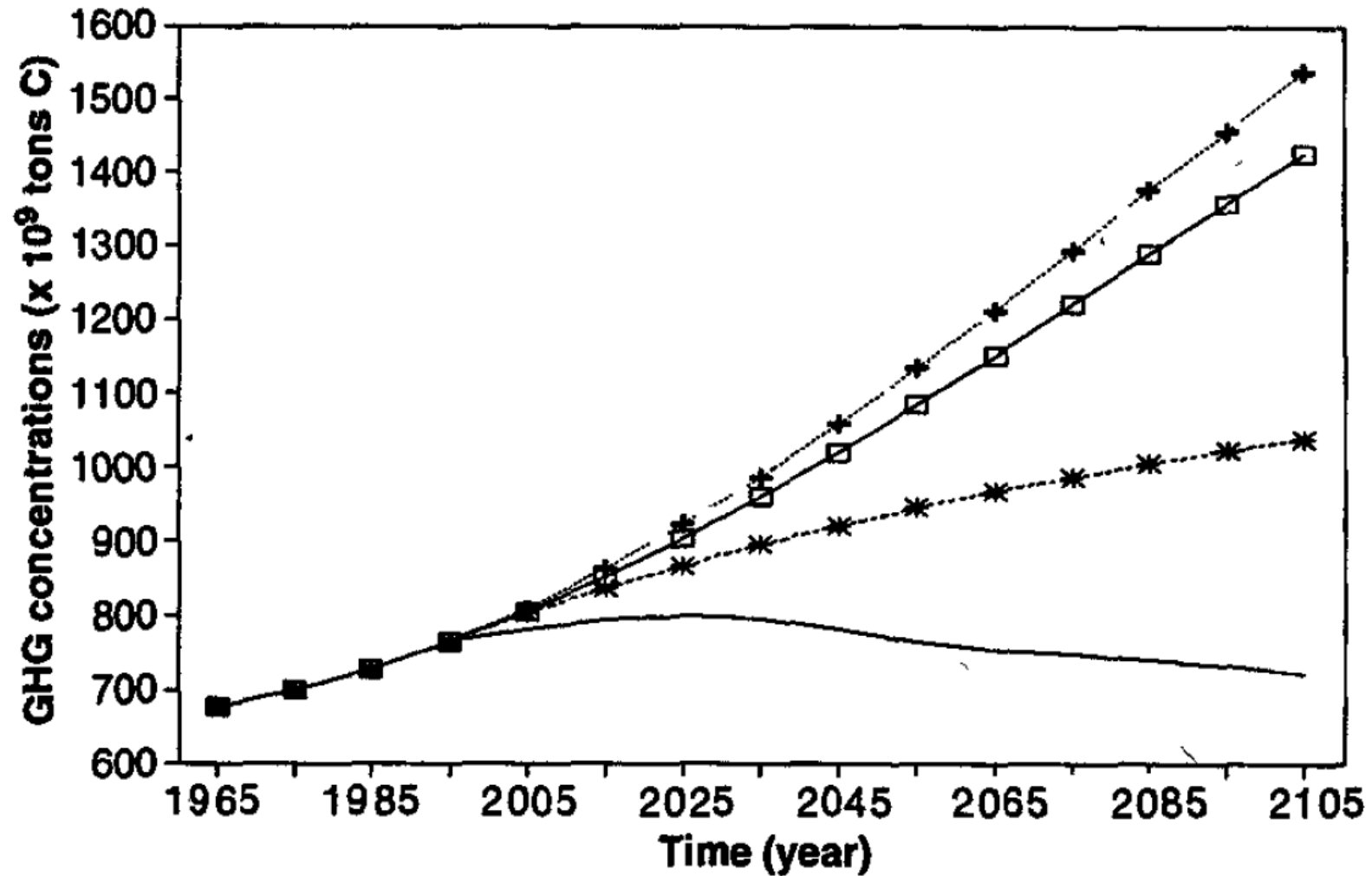


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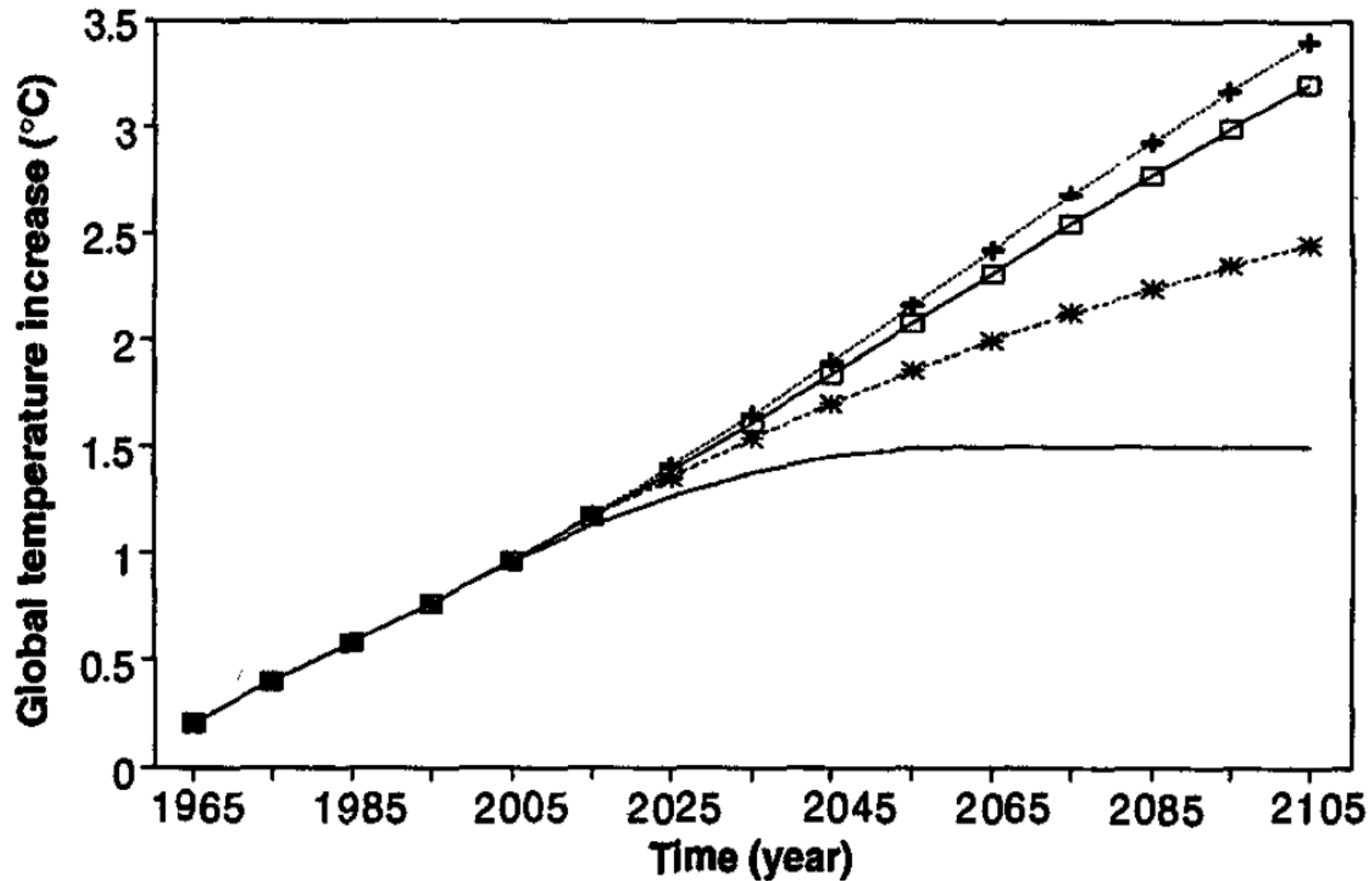
Nordhaus 1992



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Policy optimisation modelling



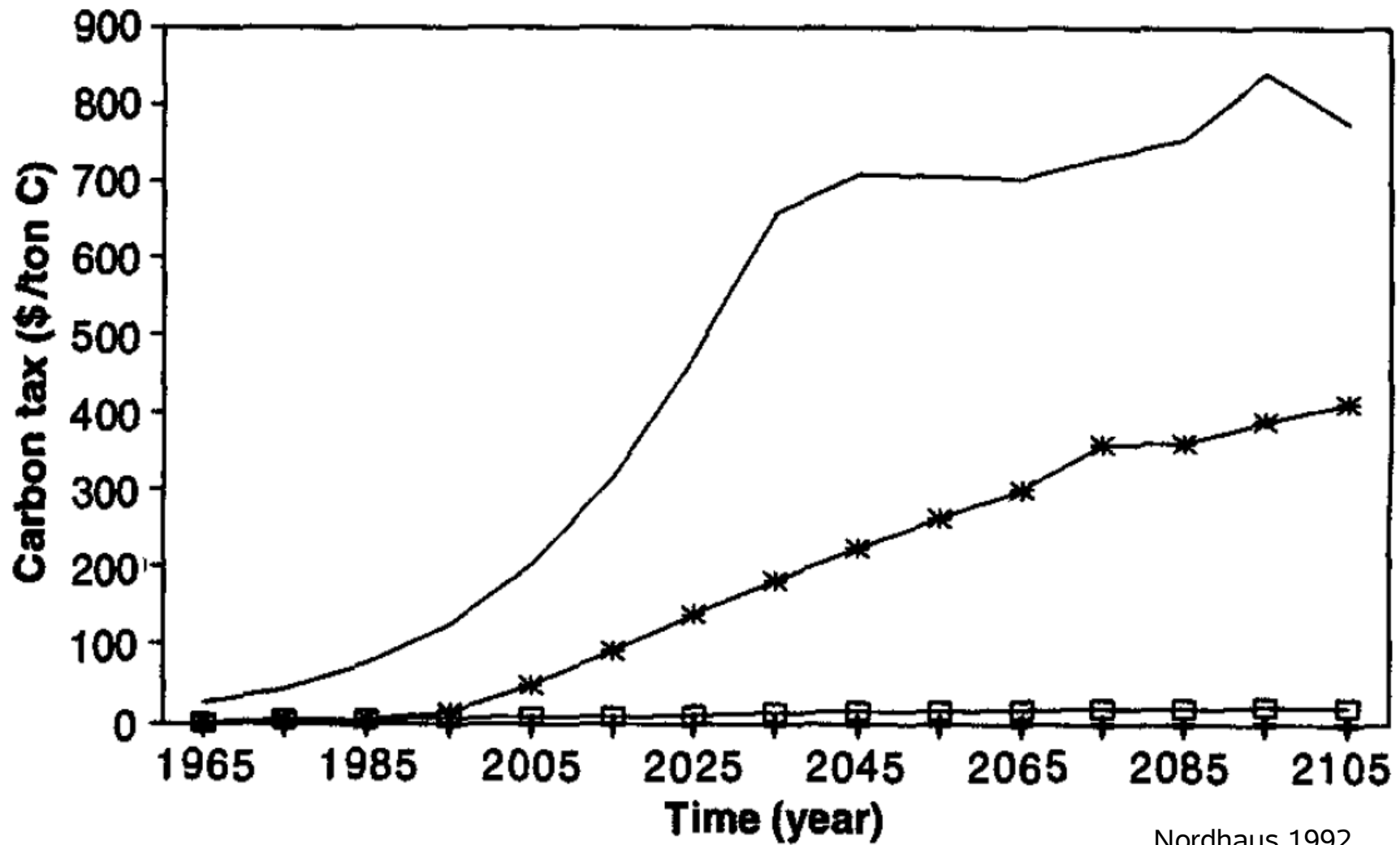
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Nordhaus 1992



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optimisation modelling II

- Advantages:
 - Comparison in single metric
 - Allows determination of policy recommendations
- Disadvantages:
 - Global aggregation masks winners and losers of climate change
 - Cost / benefit studies mainly for industrialized countries
 - Costing of non-market impacts very uncertain, possibly ethically non-desirable
 - Discounting leads to low valuation of future impacts



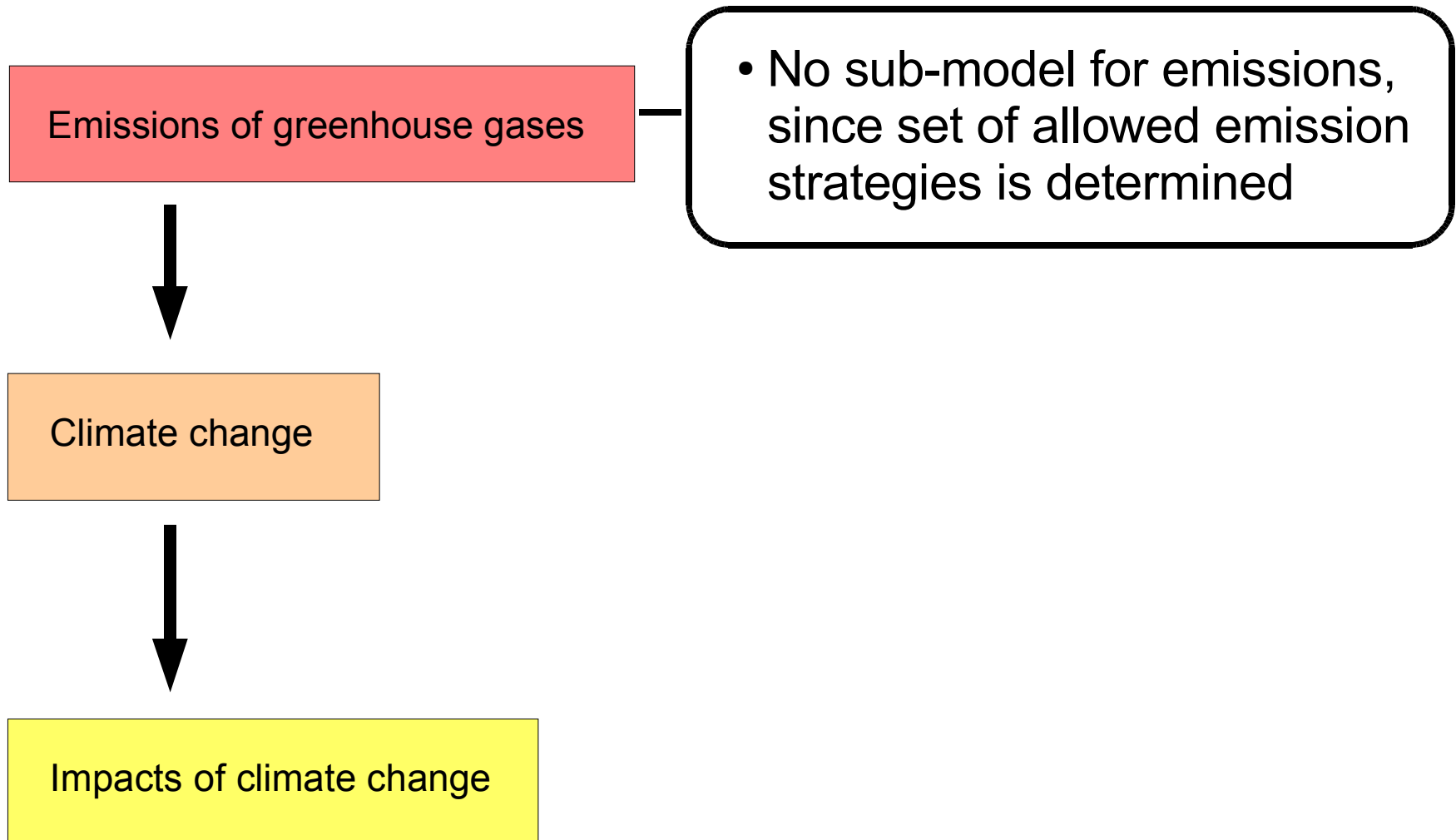
Policy guidance modelling

- Aim: determine control strategies that are compatible with climate change policy objectives
- General approach:
 - Introduce additional constraints (“guardrails”) to exclude undesirable consequences of climate change or undesirable climate protection strategies
 - Determine set of emission strategies that violate none of the introduced guardrails
- Example: Tolerable Windows Approach (TWA), Bruckner et al. 1999



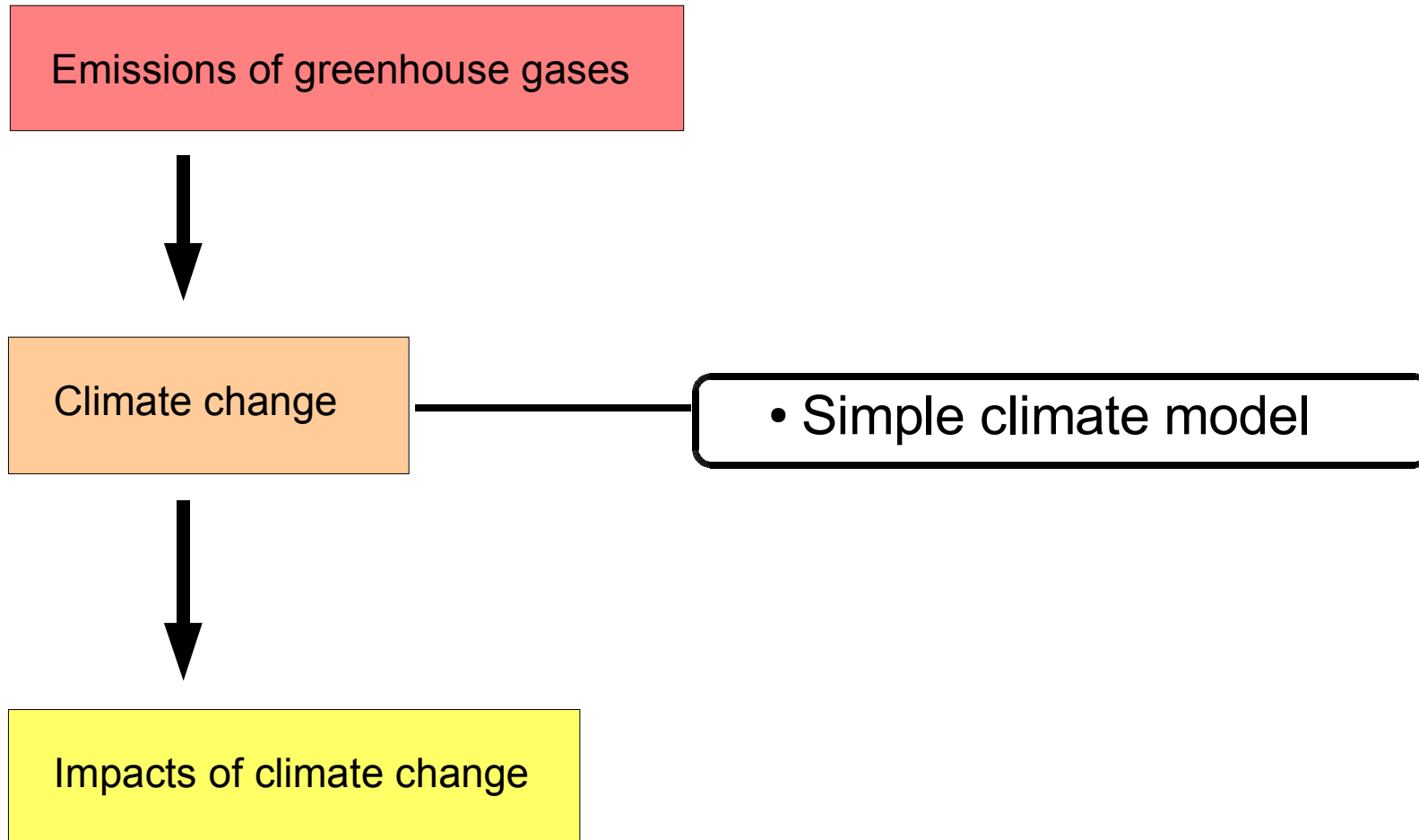
Tolerable Windows Approach

- TWA = Tolerable windows approach



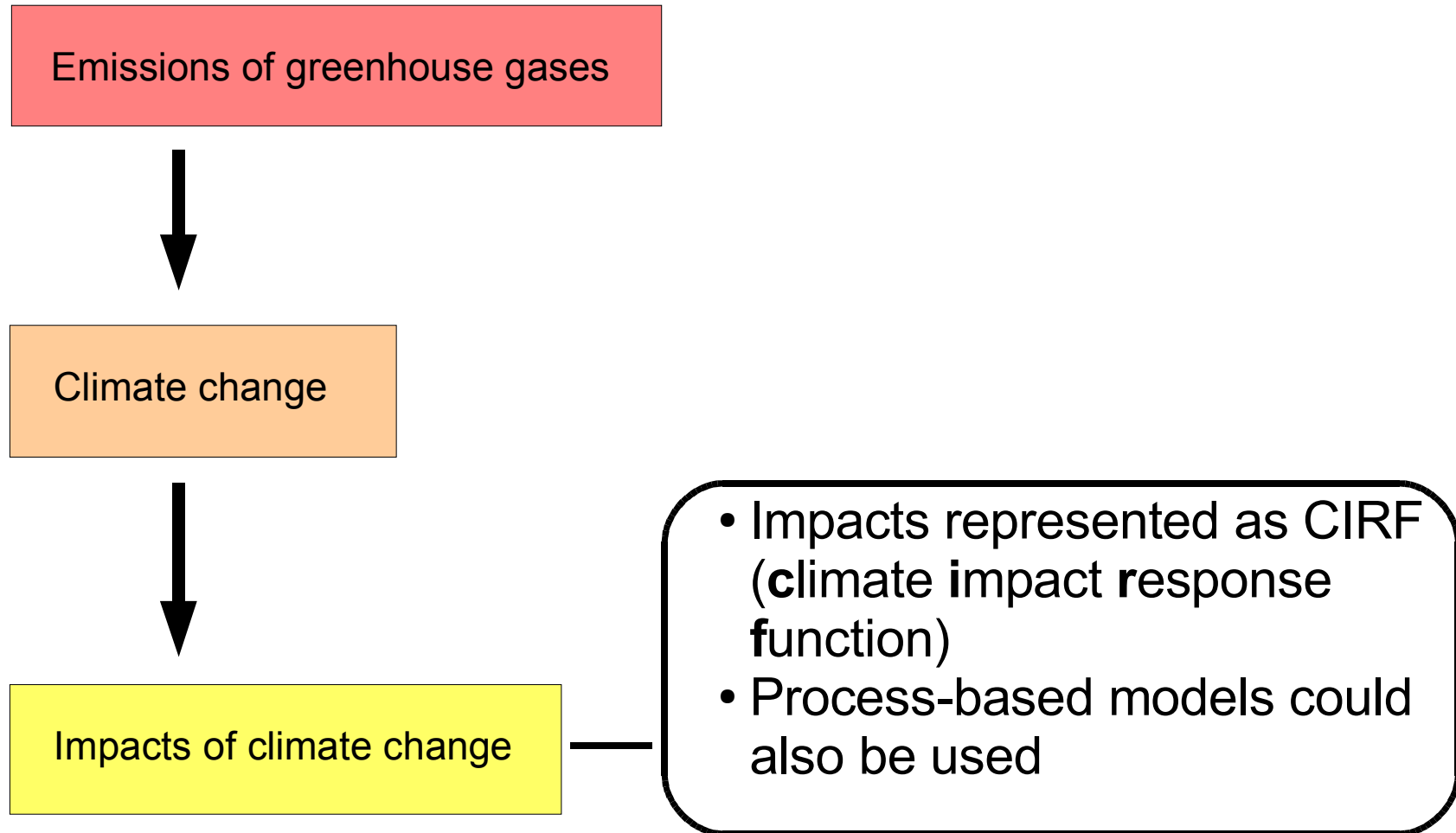
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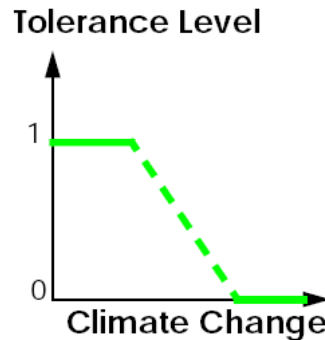


Normative Assessment:

TWA schematically

- In the TWA, assessment starts with “guardrails”
- Guardrails define tolerable climate change impacts / GHG reductions
- Analysis subsequently determines set of admissible protection strategies

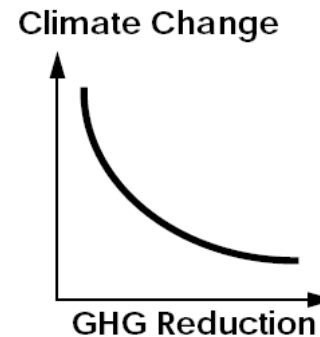
Climate Impact (CI):



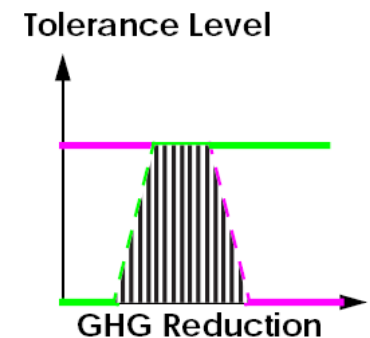
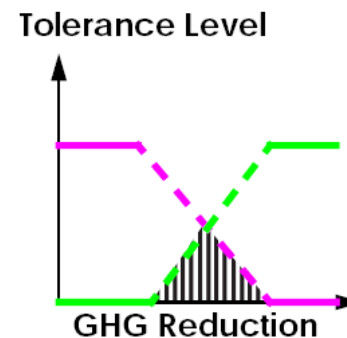
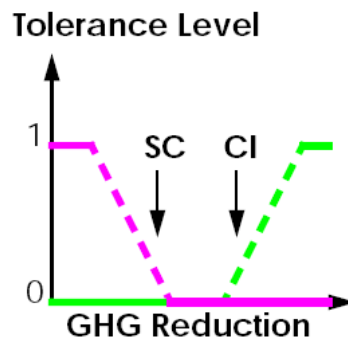
Socio-Economic Consequences (SC):



Scientific Analysis:

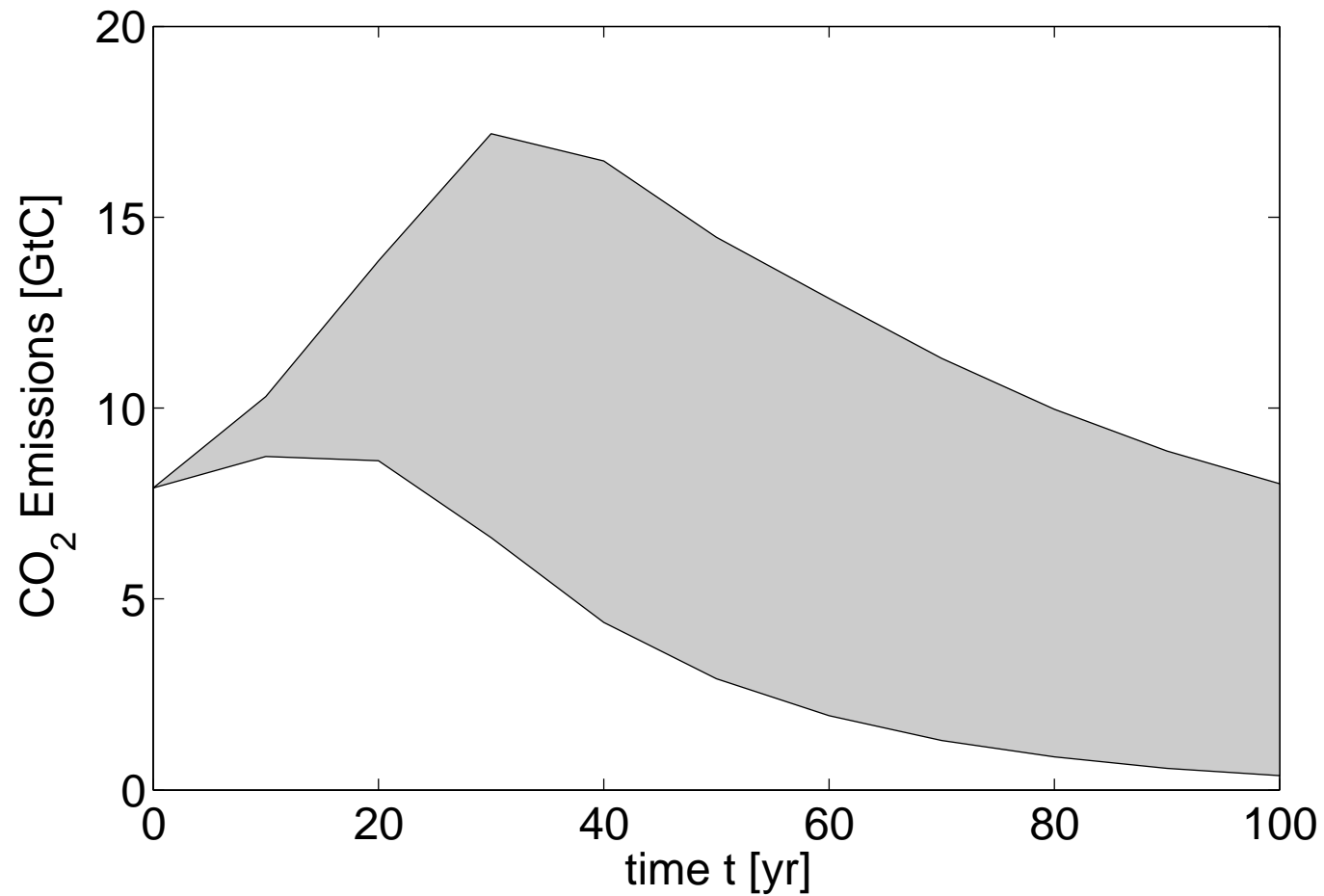


Determination of all admissible climate protection paths:



Emission corridor

Emission corridor for $\Delta T \leq 2.5^\circ\text{C}$



Summary

- Three paradigms in integrated assessment
- Distinguished by handling of control vector:
 - Prescribed for policy evaluation modelling
 - Optimized in policy optimisation modelling
 - Set compatible with constraints determined in policy guidance modelling
- Approaches are complementary
- Neither takes uncertainty into account explicitly



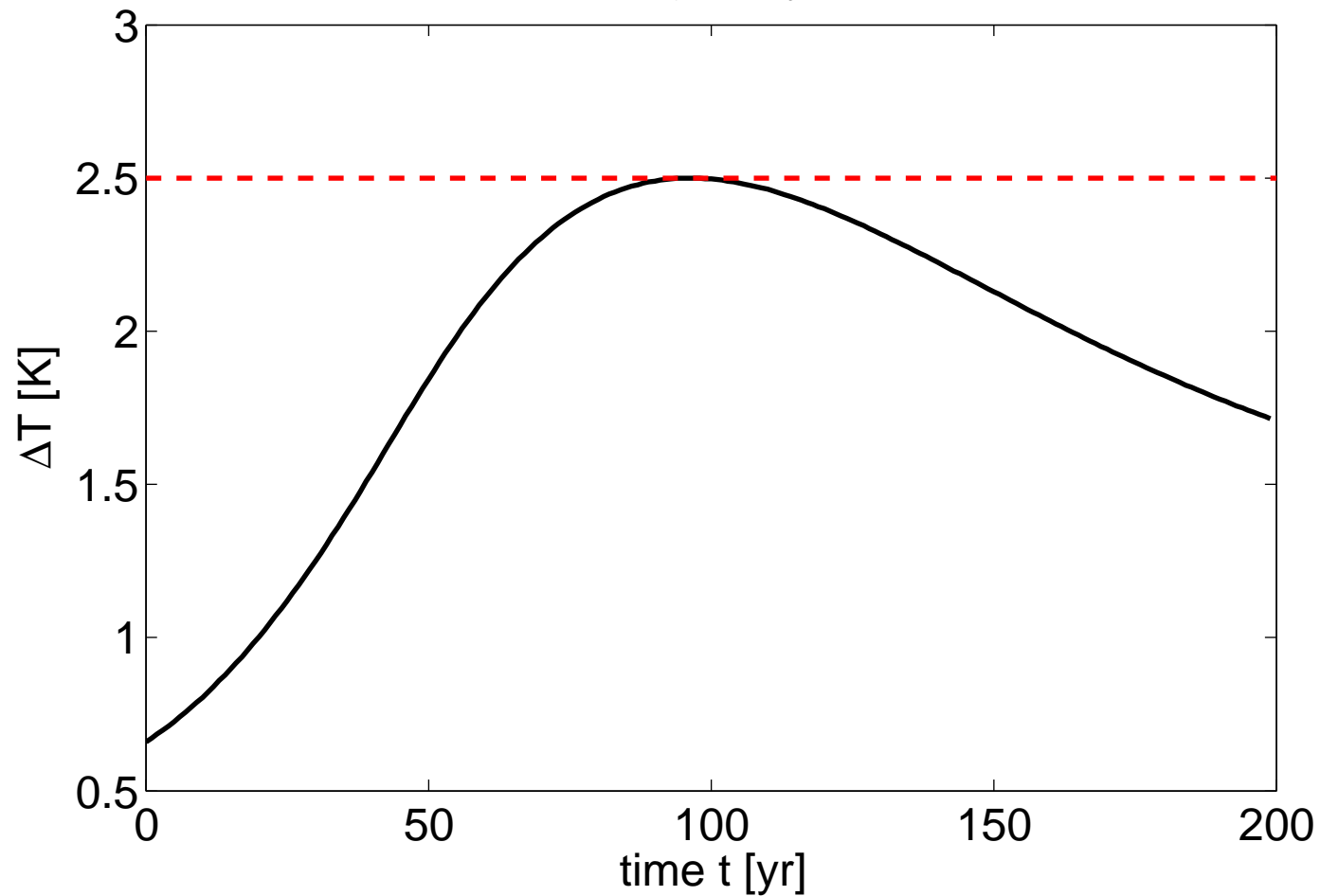
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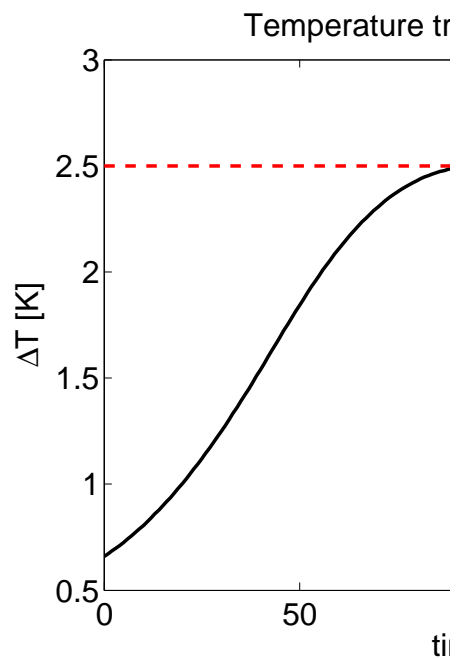


Temperature trajectories

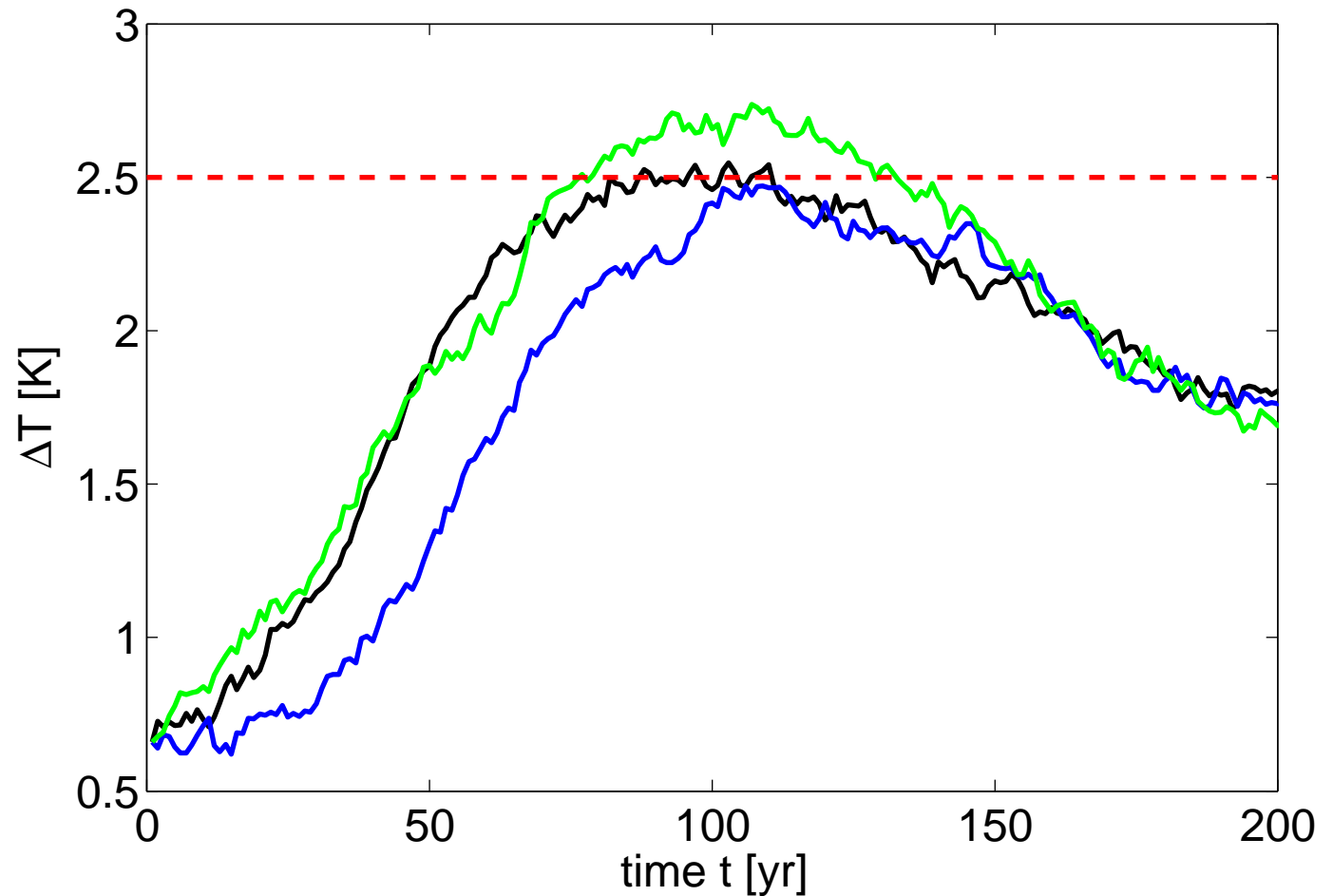
Temperature trajectory: deterministic



Temperature trajectories with nat. variability

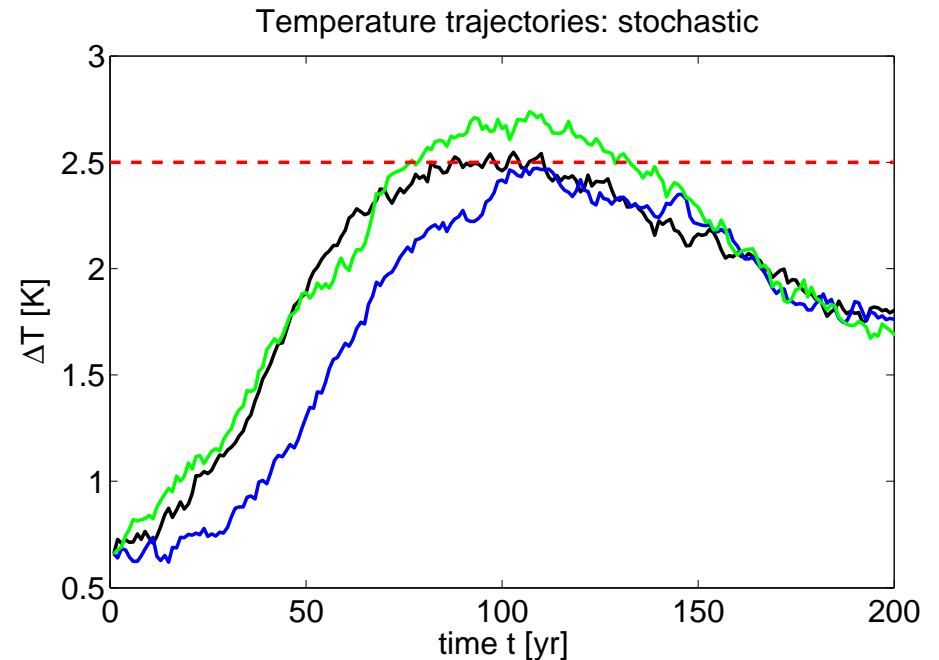
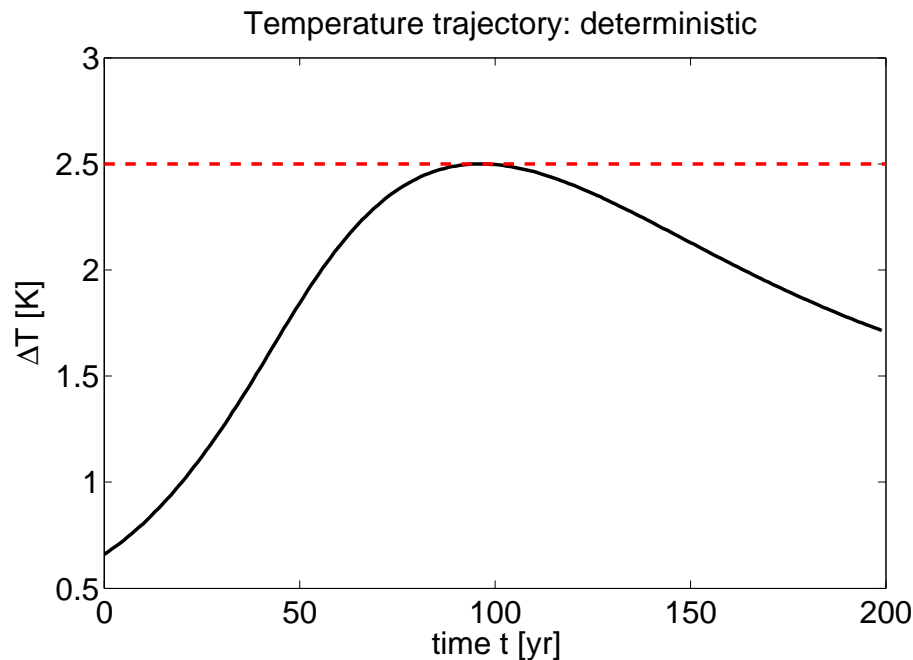


Temperature trajectories: stochastic



Temperature trajectories with nat. variability

- Consideration of natural variability possible in stochastically modified climate model
- Result: observing guardrail dependent on *realisation* of stochastic process => non-zero probability that guardrail is exceeded



Uncertainty

- Uncertainty ever present factor in entire chain of cause-and effect of climate change
- Sensible classification for our purposes by **causes of uncertainty**:
 - 1) Uncertainty caused by the freedom of human decisions
 - 2) Uncertainty caused by natural variability
 - 3) Uncertainty caused by insufficient knowledge
- TWA partly anticipates 1) since human decisions are not *predicted*, but the *manoeuvring space for human decisions* is determined instead
- 2) and 3) subject of the *probabilistic TWA*



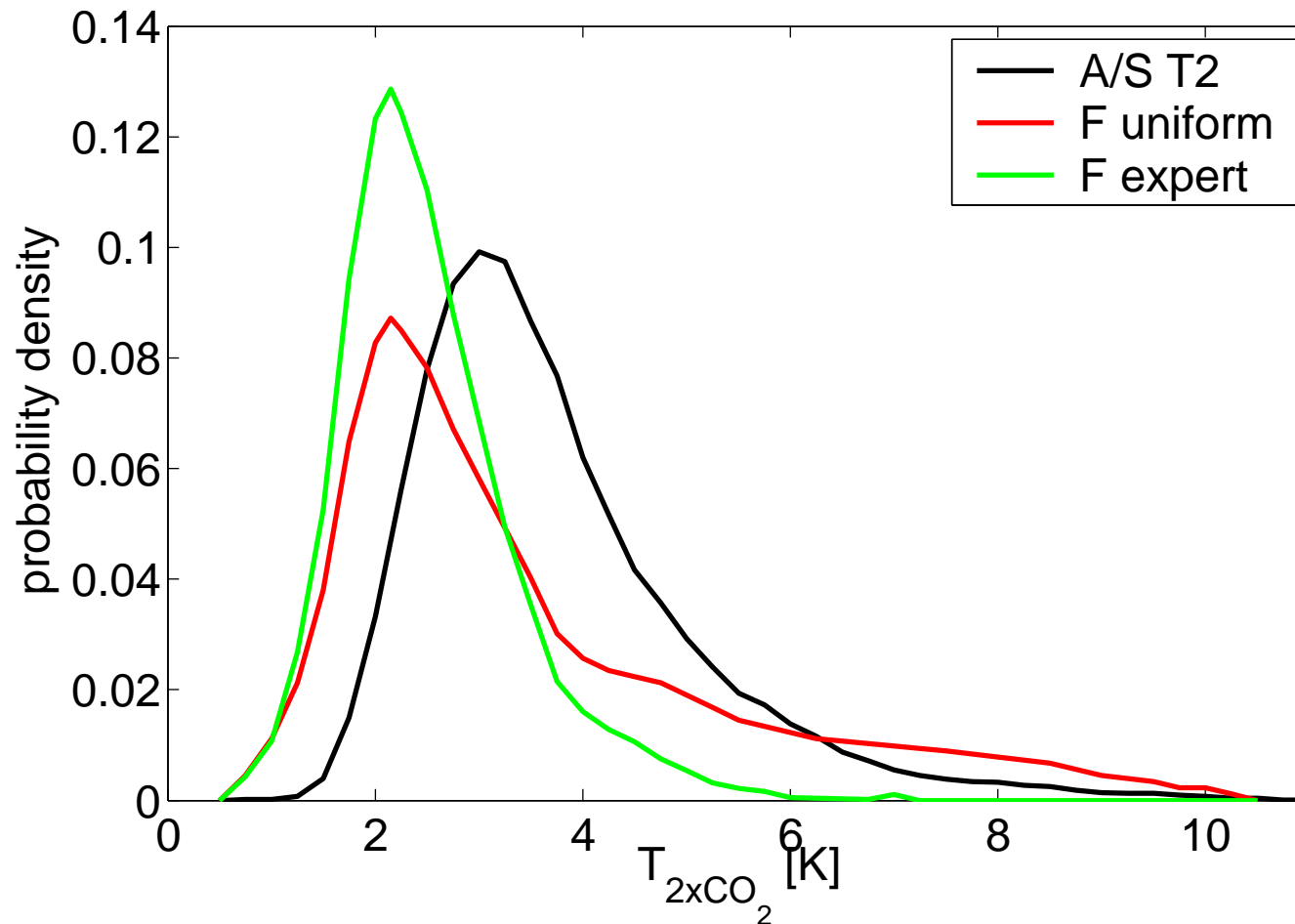
Uncertainty in climate sensitivity

- *Climate sensitivity* is one of the key uncertain factors for future climate change
- Climate sensitivity T_{2xCO_2} warming to be expected for doubling of preindustrial CO_2 - concentration
- IPCC: $T_{2xCO_2} \in [1.5^\circ C, 4.5^\circ C]$
- Other authors: probability distributions for T_{2xCO_2} , i.e. from expert elicitations, comparisons of historical climate with model results



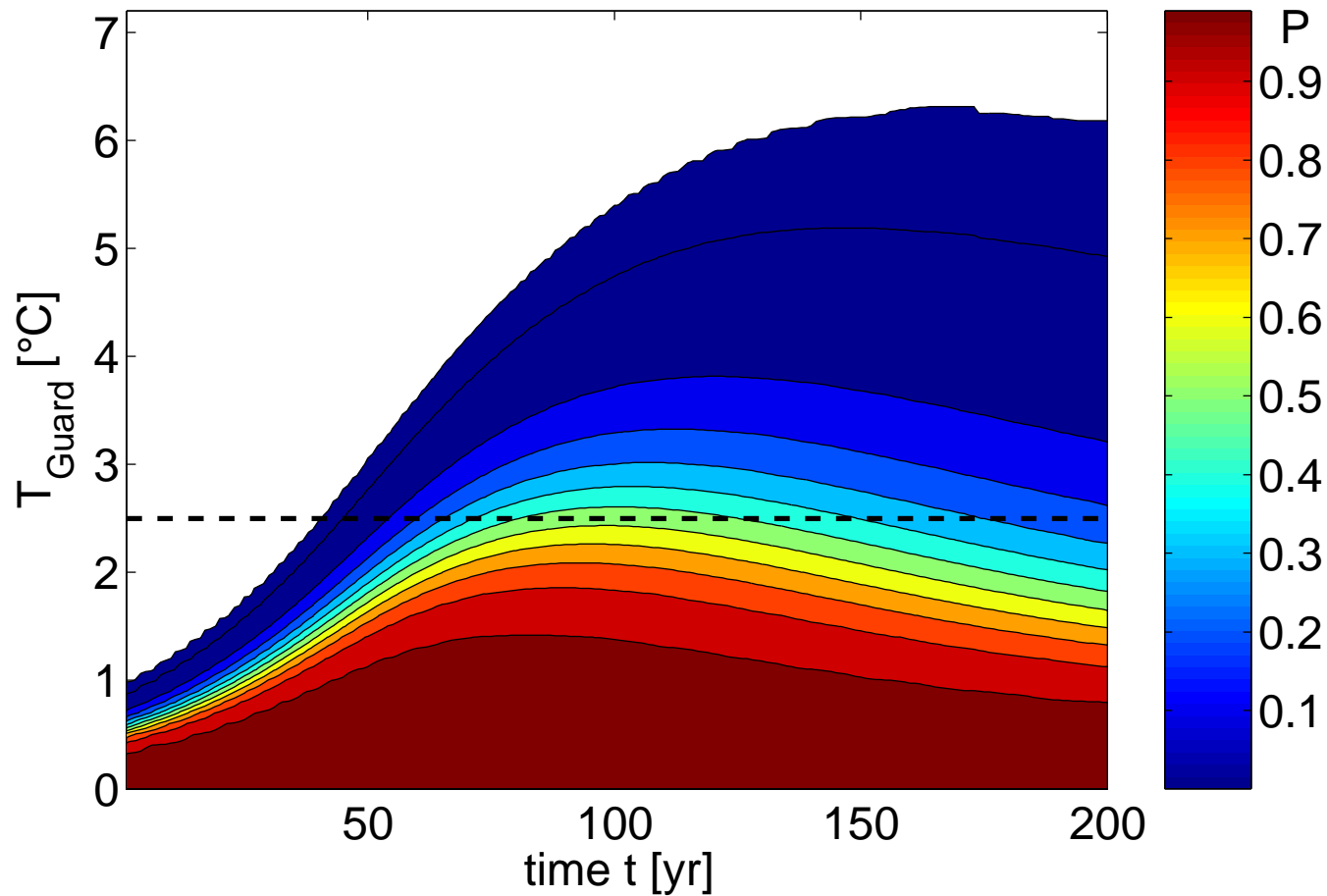
Probability distributions climate sensitivity

Probability density functions for T_{2xCO_2}



Consequences of uncertainty climate sensitivity

$P(\Delta T(t) > T_{\text{Guard}})$, uncertain clim. sens.



The probabilistic TWA

- Uncertainties imply: Extension of TWA necessary
- Deterministic guardrail for impact I defined as

$$I \leq I_{Guard} \Rightarrow P(I \leq I_{Guard}) \in \{0,1\}$$

- If probabilistic uncertainty considered:

$$P(I \leq I_{Guard}) \in [0,1]$$

- Therefore additional **probability guardrail** necessary

$$I \leq I_{Guard} \Rightarrow P(I \leq I_{Guard}) \geq P_{Guard}$$

Solution algorithm

- Problem to be solved: generally **stochastic differential inclusion**

$$d\boldsymbol{\xi} \in \mathbb{F}(\boldsymbol{\xi}, dt \oplus d\boldsymbol{W})$$

$$\text{with } \mathbb{F} := \{ \boldsymbol{f}(\boldsymbol{\xi}, t; \boldsymbol{u}) dt + \boldsymbol{g}(\boldsymbol{\xi}, t; \boldsymbol{u}) d\boldsymbol{W} \mid \boldsymbol{u} \in \mathcal{U} \}$$

$$\text{under } P(\boldsymbol{h}(\boldsymbol{\xi}, t; \boldsymbol{u}) \leq \mathbf{0}) \geq P_{Guard} \quad \forall t \in [0, t_e]$$

- Determination of the upper (lower) boundary of emission corridors:

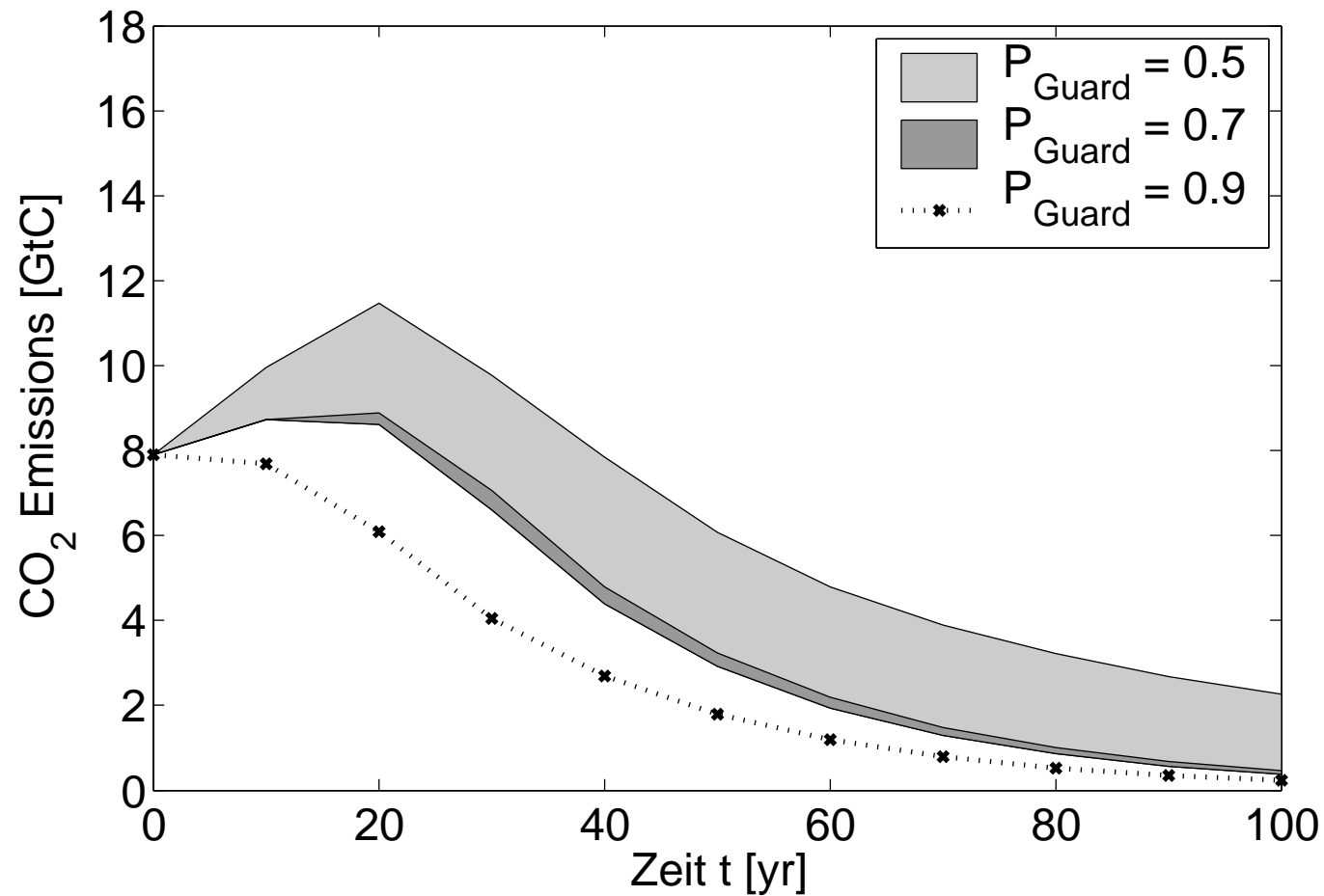
$$\forall t_i \in \{t_1, \dots, t_n\}: \max(\min) E(t_i)$$

$$\text{under } P(\boldsymbol{h}(\boldsymbol{\xi}, t; \boldsymbol{u}) \leq \mathbf{0}) \geq P_{Guard} \quad \forall t \in [0, t_e]$$

- Standard algorithms for constrained optimisation can be used
- P -guardrails can be evaluated using Monte-Carlo approach

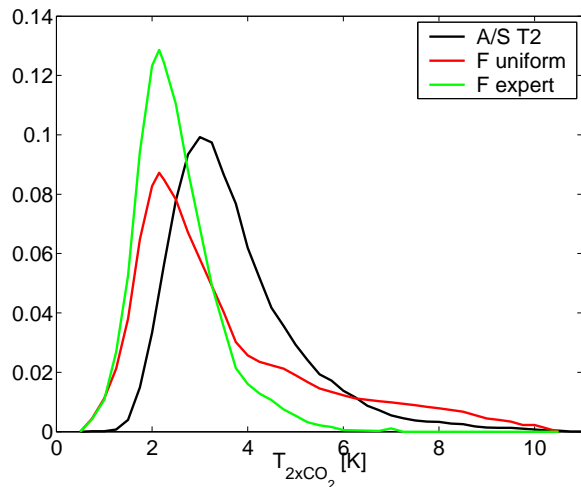
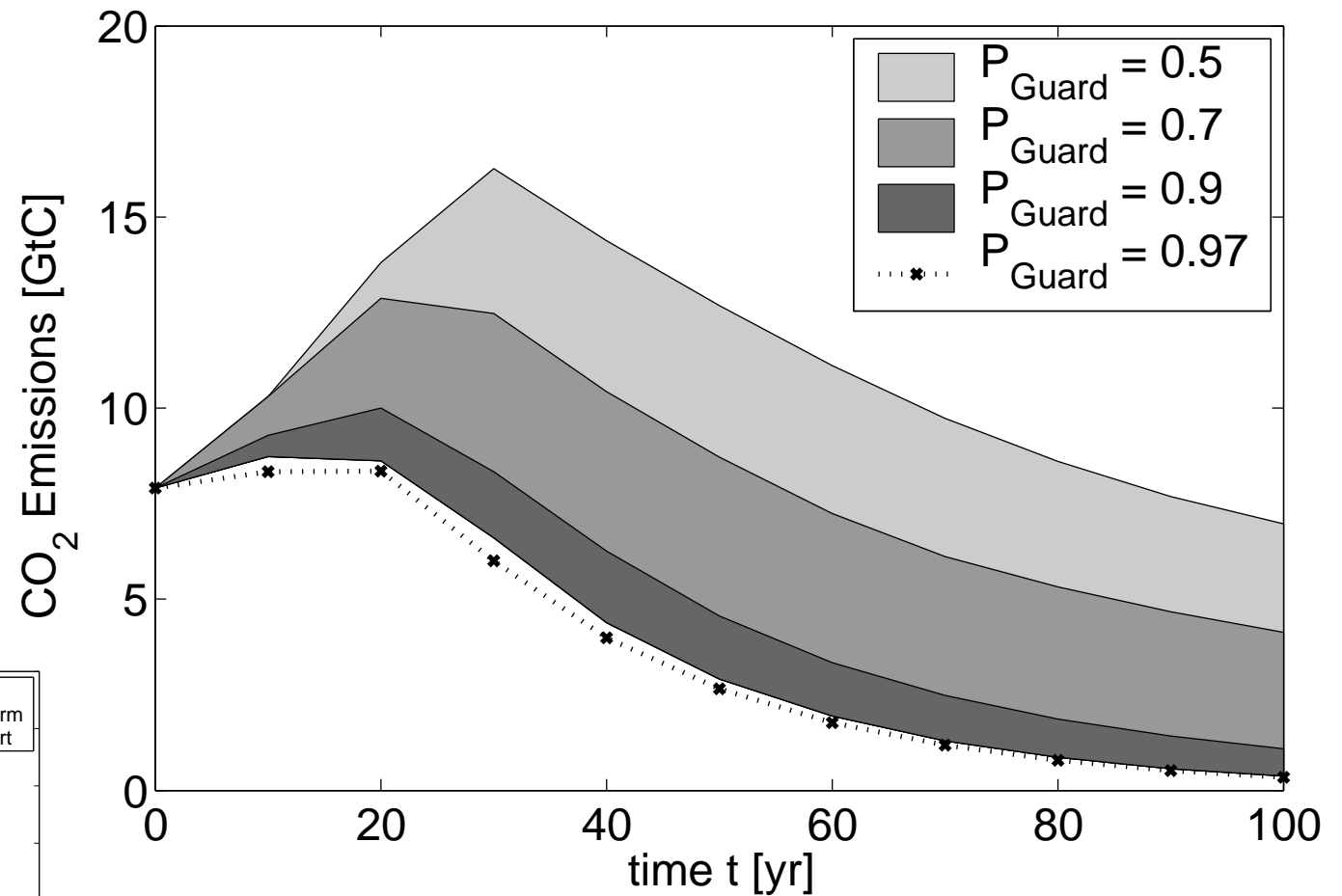
Results: uncertain climate sensitivity I

A&S, $T_{\text{Guard}} = 2\text{K}$, $P(T \leq T_{\text{Guard}}) \geq 0.5, 0.7, 0.9$



Results: uncertain climate sensitivity II

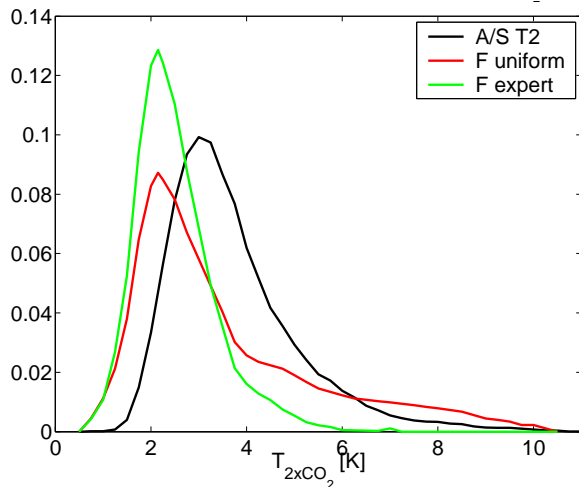
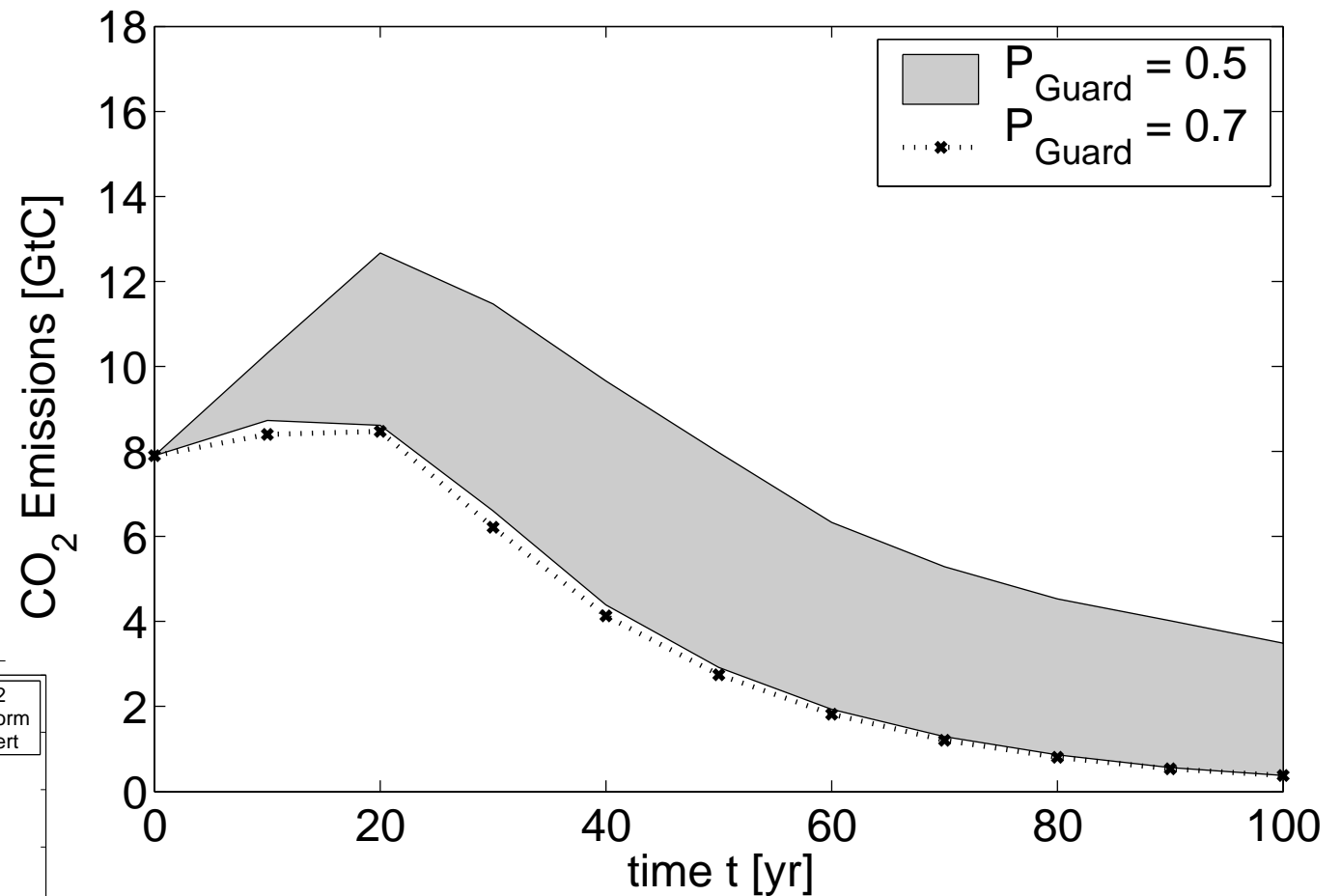
Forest expert, $T_{\text{Guard}} = 2\text{K}$, $P(T \leq T_{\text{Guard}}) \geq 0.5, 0.7, 0.9, 0.97$



Integrated Modelling of Climate Change
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Results: uncertain climate sensitivity III

Forest uniform, $T_{\text{Guard}} = 2\text{K}$, $P(T \leq T_{\text{Guard}}) \geq 0.5, 0.7$



Integrated Modelling of Climate Change
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Summary

- Uncertainty ever-present in IA modelling
- TWA can be extended to probabilistic approach
- Allows consideration of uncertainty through natural variability and through uncertain parameters
- EU target of max. 2°C warming very ambitious
- GHG emissions need to be reduced quickly and strongly



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Changes in flooding probability

- Aim: develop representation of changes in flooding probability (large river basins) in integrated assessment model
- Requirements:
 - Global scale
 - Low computational cost
- Model needs:
 - Downscaling scheme from ΔT_{GM} to $\Delta P, \Delta E$ on river basin scale
 - Representation of natural variability in P, E
 - Hydrological model to aggregate change in P, E to river basin scale
- Resolution chosen: $\Delta x = 0.5^\circ$, $\Delta t = 1 \text{ month}$
- Min. basin size: $2.5 \times 10^4 \text{ km}^2$



Downscaling scheme

- IA models typically determine ΔT_{GM} only
- Changes in mean climate: pattern scaling

- Changed mean climate

$$\bar{T}(r, m, t) = T_C(r, m) + k \Delta T_{GM}(t) \times T_P(r, m)$$

$$\bar{P}(r, m, t) = P_C(r, m) \times (1 + k \Delta T_{GM}(t) \times P_P(r, m))$$

- Natural variability: deviation patterns from CRU-TS (PIK modification) data
- Representation of nat. variability

$$T(r, m, t) = T_C(r, m) + k \Delta T_{GM}(t) \times T_P(r, m) + T'(r, m, t')$$

$$P(r, m, t) = (P_C(r, m) \times (1 + k \Delta T_{GM}(t) \times P_P(r, m))) \times P'(r, m, t')$$

Hydrological model

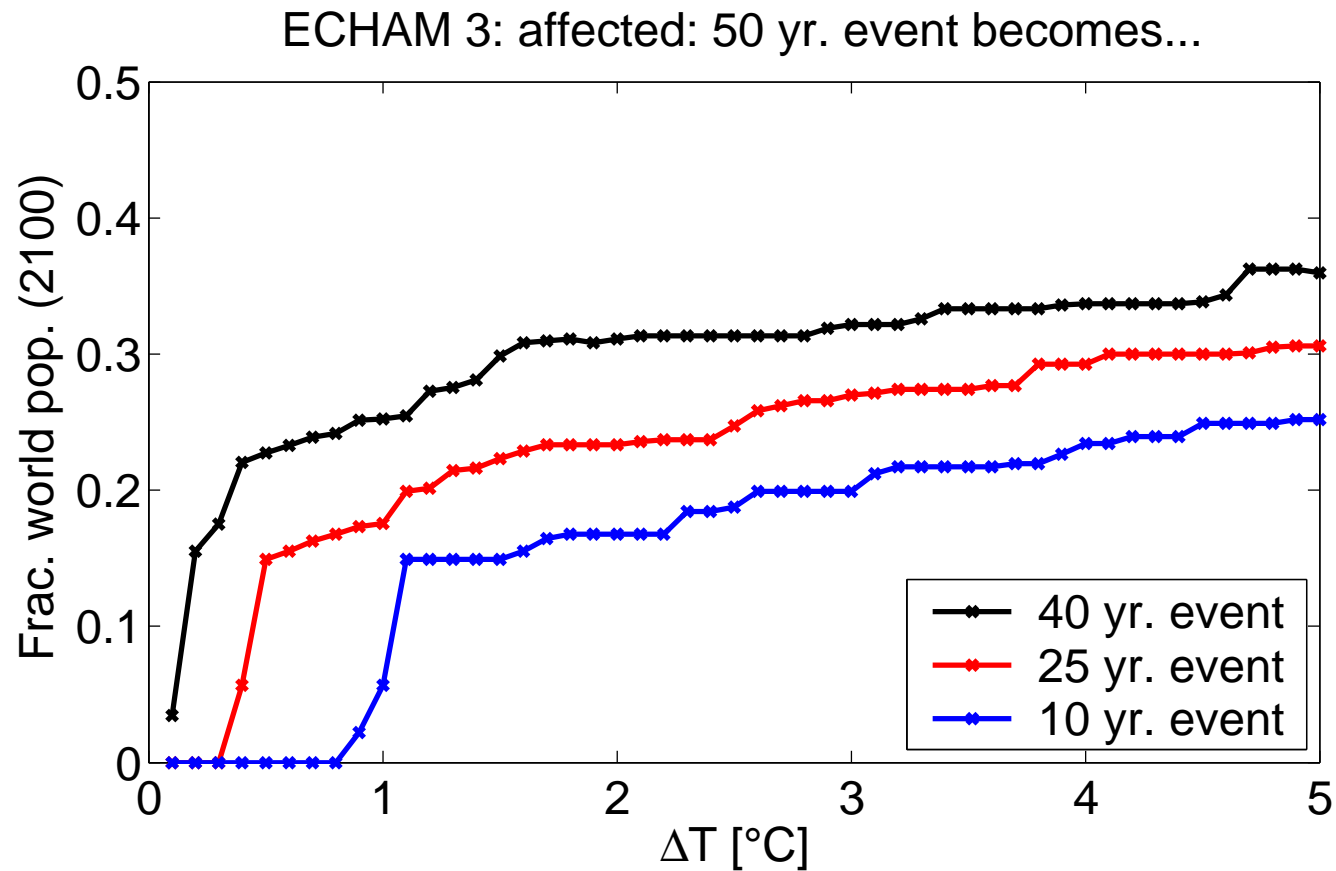
- Most simple model possible:
 - Determine P, E at all grid points belonging to river basin
 - Sum up total $R = P - E - \Delta S$ ($\Delta S=0$) over all grid points
- Model validation using gauge records and historical CRU-TS(PIK) data:

Model performance is comparably good (or rather: bad) as performance of other models on these scales.
- Aggregation measure for setting of guardrails:

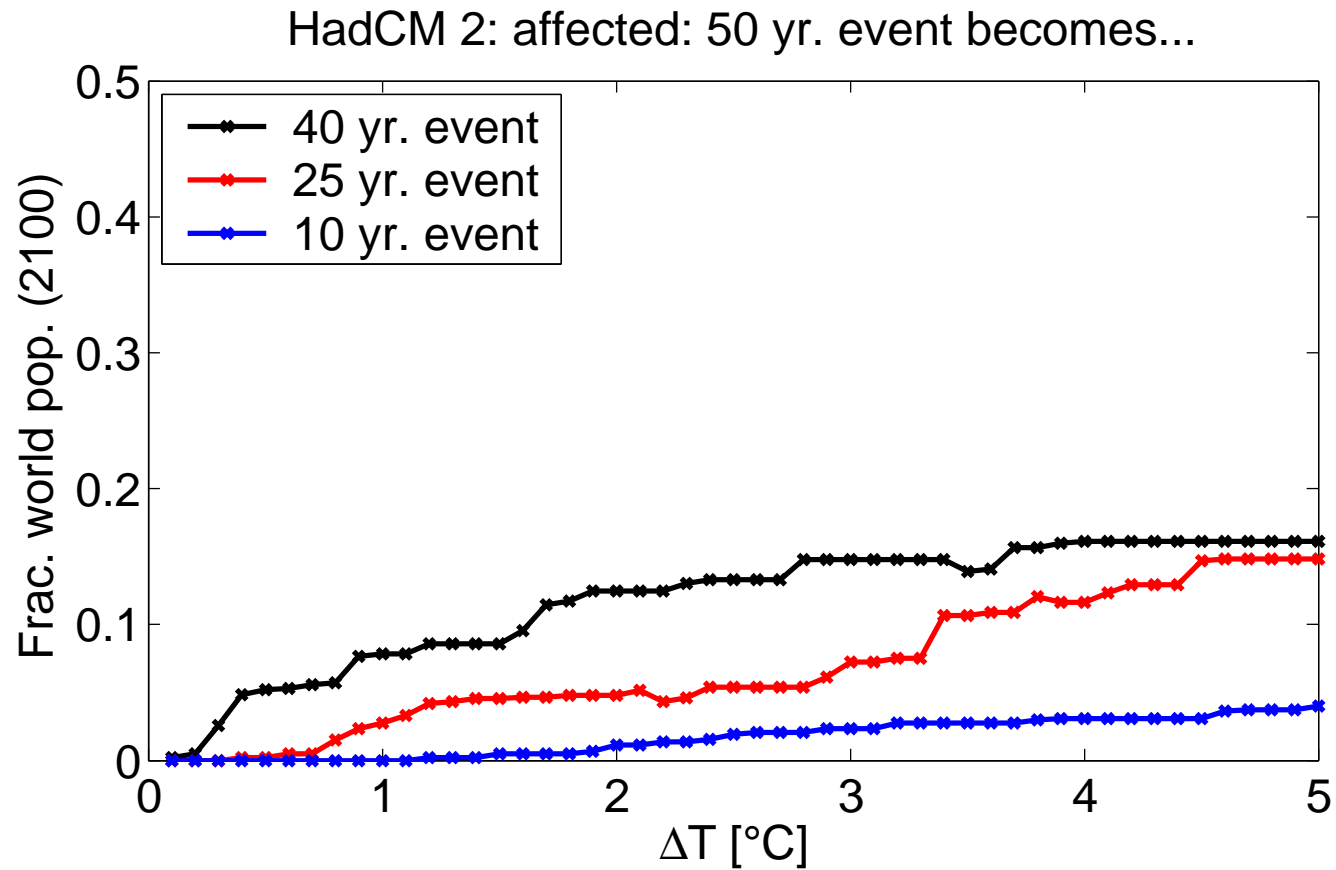
Population (2100) affected by positive change in probability of 50 year flood event Q_{50yr}



Climate Impact Response Function: $\Delta P(Q_{50yr})$

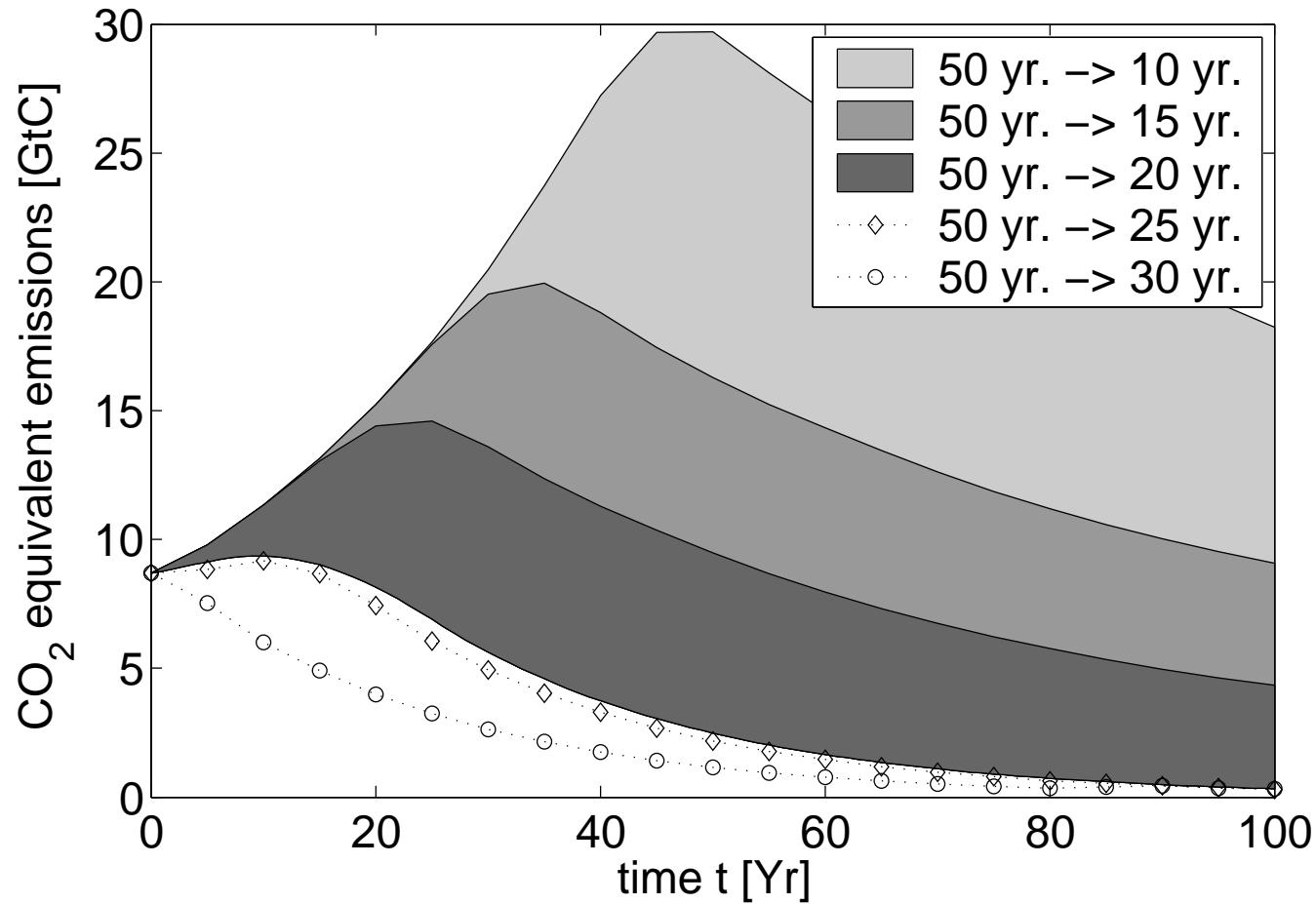


Climate Impact Response Function: $\Delta P(Q_{50yr})$



Emission corridor: flooding

ECHAM 3: Change in 50 yr. extreme: 20% affected



Summary

- Climate change will change probability of large flood events in river basins
- Changed probability can be determined using simple flooding model consisting of downscaling scheme and hydrological model
- This model can be used to determine CIRF for changes in flooding probability
- Depending on changes to monsoon climate, large proportions of population may already be affected for small climate change
- Limiting population fraction affected will be big challenge



Thank you very much for your attention.

These slides will be available at www.cru.uea.ac.uk/~thomask

If you have any questions please do not hesitate to contact me at:
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