

Future sea level rise constrained by observations and long-term commitment

Matthias Mengel^{a,b}, Anders Levermann^{a,b,c,1}, Katja Frieler^a, Alexander Robinson^{a,d,e}, Ben Marzeion^f, and Ricarda Winkelmann^{a,b}

^aPotsdam Institute for Climate Impact Research, 14473 Potsdam, Germany; ^bPhysics Institute, Potsdam University, 14476 Potsdam, Germany; ^cLamont-Doherty Earth Observatory, Columbia University, Palisades, NY 10964; ^dUniversidad Complutense de Madrid, 28040 Madrid, Spain; ^eInstituto de Geociencias (IGEO), Consejo Superior de Investigaciones Científicas - Universidad Complutense de Madrid, 28040 Madrid, Spain; and ^fInstitute of Geography, University of Bremen, 28359 Bremen, Germany

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Sea level has been steadily rising over the past century, predominantly due to anthropogenic climate change. The rate of sea level rise will keep increasing with continued global warming, and, even if temperatures are stabilized through the phasing out of greenhouse gas emissions, sea level is still expected to rise for centuries. This will affect coastal areas worldwide, and robust projections are needed to assess mitigation options and guide adaptation measures. Here we combine the equilibrium response of the main sea level rise contributions with their last century's observed contribution to constrain projections of future sea level rise. Our model is calibrated to a set of observations for each contribution, and the observational and climate uncertainties are combined to produce uncertainty ranges for 21st century sea level rise. We project anthropogenic sea level rise of 28–56 cm, 37–77 cm, and 57–131 cm in 2100 for the greenhouse gas concentration scenarios RCP26, RCP45, and RCP85, respectively. Our uncertainty ranges for total sea level rise overlap with the process-based estimates of the Intergovernmental Panel on Climate Change. The “constrained extrapolation” approach generalizes earlier global semiempirical models and may therefore lead to a better understanding of the discrepancies with process-based projections.

sea level rise | climate change | climate impacts

Sea level has been rising between 16 and 19 cm since 1900 (1, 2) with a rate of around 3 cm per decade since 1990 (3, 4). Thermal expansion of the oceans and retreating glaciers are the main contributors to sea level rise in the past century and the near future. On multicentennial timescales, the Greenland and Antarctic ice sheets will likely dominate global sea level rise (5). Future sea level rise will pose challenges to coastal regions around the globe, and robust projections are needed to guide adaptation investment and provide incentives for climate mitigation (6).

Projecting sea level relies on the understanding of the processes that drive sea level changes and on reliable data to verify and calibrate models. So-called process-based models now deliver projections for the main components of climate-driven sea level rise—thermal expansion, glaciers and ice caps, the Greenland ice sheet, and the Antarctic ice sheet—although solid ice discharge (SID) from the ice sheets is still difficult to constrain (3). Semiempirical models follow a different approach and use the statistical relation between global mean temperature (7, 8) or radiative forcing (9, 10) and sea level from past observations. Without aiming to capture the full physics of the sea level components, they project future sea level assuming that the past statistical relation also holds in the future. Their simpler nature makes them feasible for probabilistic assessments and makes their results easier to reproduce.

The long-term multicentennial to millennial sensitivity of the main individual sea level contributors to global temperature changes can be constrained by paleoclimatic data and is more easily computed with currently available process-based large-scale models than are decadal to centennial variations (5, 11). In addition, there is an increasing number of observations available for the historical

individual contributions to sea level rise, which capture the early response to global temperature changes.

Here we seek to combine the long-term sensitivity (or long-term commitment) and the individual observations to constrain estimates of near-future sea level rise by semiempirical relations for each sea level contributor. This expands the classical semiempirical approach that has so far been based on total sea level rise. We use a pursuit curve to estimate sea level rise in accordance with the respective long-term sensitivity. We define $S(t)$ as the time-dependent sea level contribution, $S_{eq}(T, \alpha)$ is the long-term sensitivity for the sea level component as a function of global mean temperature T and the commitment factor α (see methods), and τ is the response timescale. We can then model the short-term rate of sea level rise as a function of global mean temperature as

$$\frac{dS}{dt} = \frac{S_{eq}(T(t), \alpha) - S(t)}{\tau} \quad [1]$$

This ordinary differential equation describes a physical system in which S seeks to approach its equilibrium value (here S_{eq}) with speed linearly dependent on the deviation from the equilibrium and the inverse of τ . The approach has already been applied to project total sea level rise (10). The integrated equation yields the sea level evolution. Uncertainty in the long-term sensitivity S_{eq} is covered by variation of commitment parameters. We calibrate τ

Significance

Anthropogenic sea level rise poses challenges to coastal areas worldwide, and robust projections are needed to assess mitigation options and guide adaptation measures. Here we present an approach that combines information about the equilibrium sea level response to global warming and last century's observed contribution from the individual components to constrain projections for this century. This “constrained extrapolation” overcomes limitations of earlier global semiempirical estimates because long-term changes in the partitioning of total sea level rise are accounted for. While applying semiempirical methodology, our method yields sea level projections that overlap with the process-based estimates of the Intergovernmental Panel on Climate Change. The method can thus lead to a better understanding of the gap between process-based and global semiempirical approaches.

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Data deposition: The source code is available from <https://github.com/matthiasmengel/sealevel>.

¹To whom correspondence should be addressed. Email: levermann@pik-potsdam.de.

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by minimizing the sum of the squared residuals (“least-squares”) between observed and modeled sea level evolution for the past for each contributor and each observational dataset.

Results

Thermal Expansion. Past thermosteric sea level rise can be inferred from observations of ocean temperature that are available for several ocean depth ranges (Fig. S1). The upper ocean layer (0–700 m) is best observed (12–14). Fewer observations are available for the middepth (14) and abyssal ocean (15). To encompass the uncertainty from the different observational datasets, we create all possible combinations of the observations from different depths to yield 12 estimates for total thermosteric sea level rise (see *Supporting Information* for details). For the given range of commitment factors, our calibration method yields equilibration times τ between 82 and 1,290 y (Table S1). Driven by observed global mean temperature change (16), our model can reproduce the different time series of observed thermosteric sea level rise (see Fig. 1A for a subset and Fig. S2 for the full set). The estimates for the full time period since 1900 encompass the Coupled Model Intercomparison Project Phase 5 (CMIP5) model mean (3) (Fig. 1A, gray lines).

With the 12 calibrated tuples of α and τ , we project the sea level contribution from thermal expansion within the 21st century for the three representative concentration pathways RCP26, RCP45, and RCP85 (17). Fig. 2A shows the median and very likely (5–95%) uncertainty range for the three RCP scenarios. In 2100, the thermosteric median sea level contribution is estimated to be 15 cm, 19 cm, and 29 cm for RCP26, RCP45, and RCP85, respectively (Table 1 and Fig. 2A). The mean sea level rise 2081–2100 compared with the 1986–2005 mean (Table S2) is close to the Intergovernmental Panel on Climate Change (IPCC) projections for the three scenarios (Fig. 2A, bars at the right). In our probabilistic approach, the ocean heat uptake does not influence global mean temperature evolution, as opposed to the coupled IPCC simulations. This leads to higher uncertainty ranges for each scenario compared with the IPCC.

Mountain Glaciers. Global glacier volumes decline since the 19th century. Observation-based estimates of glacier mass changes (18–20) (see *Supporting Information* for description) have recently become more consistent despite their different reconstruction techniques (21). Although human influence dominated glacier loss in the second half of the 20th century, earlier retreat was mainly driven by natural climate variability and ongoing adjustment to past climate change. Glacier volumes decreased particularly fast in the Arctic (20) during a period of early warming (22) in the late 19th century and first half of the 20th century.

The human-induced part of total glacier loss increased over time and reached about 70% in recent years (23). We calibrate our semiempirical model to each of the anthropogenic parts of the observational datasets with each of the equilibrium sensitivities (see *Materials and Methods*). The 57 corresponding calibrated response times τ range from 98 y to 295 y. The observed anthropogenic sea level rise from glaciers is well reproduced for the second half of the 20th century, whereas the signal of early Arctic warming is not fully captured (Fig. 1B and Fig. S3). Differences remain in the early part of the time series because attribution of early Arctic warming is imperfect when the anthropogenic signal is still small.

We project a median sea level contribution of 8 cm, 9 cm, and 11 cm until 2100 for the RCP26, RCP45, and RCP85 scenarios, respectively (Fig. 2B). The glacier mass loss is less scenario-dependent than other contributions, and the 2081–2100 mean lies below the IPCC estimates (Fig. 2B, bars at the right). This is partly due to the form of its long-term contribution, which approaches a temperature-independent asymptote for strong global warming (see Fig. S4), reflecting the limited volume of the world’s glaciers. The full effect of the limited global glacier mass will become more apparent in the past-2100 contribution.

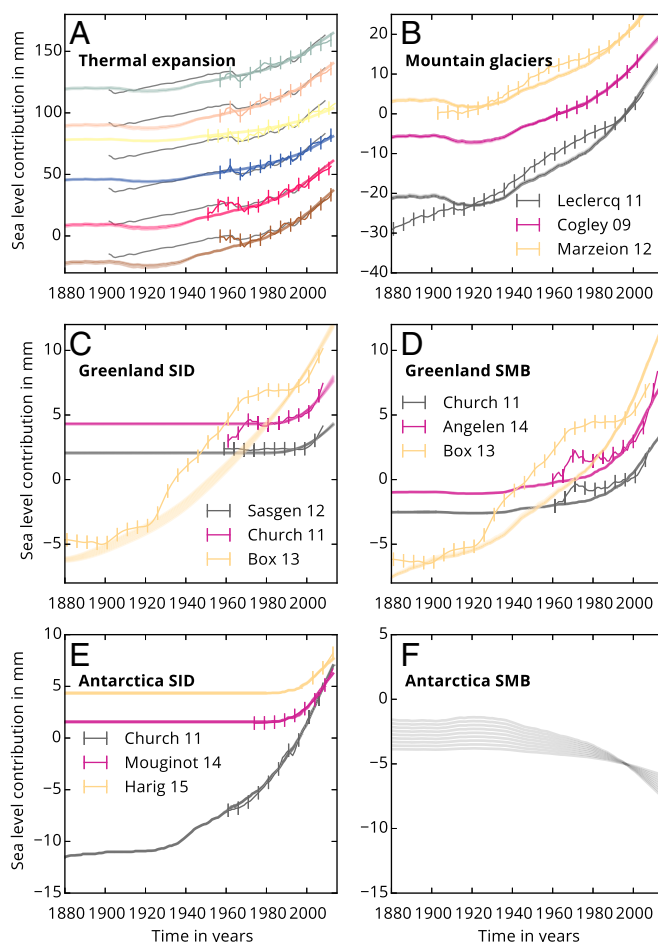


Fig. 1. Observed and calibrated sea level contributions. Observed (lines with bars) and calibrated sea level contribution (lines) during the calibration period for each sea level component. (A) Combinations of upper, middle, and deep ocean thermal expansion (12–15); only subset of all combinations shown; see Fig. S2 for full set. Gray lines show CMIP5 model mean (1). (B) Anthropogenic mountain glacier loss (18–20). (C) Greenland SID (26–28). (D) Greenland SMB (24–26). (E) Antarctic SID (26, 39, 40). The Antarctic SMB contribution (F) is not calibrated, and is provided here for comparison. All observations and calibrated contributions are relative to the 1986–2005 mean. An offset is applied between datasets for better visibility. The y axis scale varies between panels.

Greenland Surface Mass Balance. We use three different datasets for surface mass balance (SMB) reconstructions (24–26) of the Greenland ice sheet (see *Supporting Information* for details). The calibrated response time τ for the three observational datasets range from 99 y to 927 y, depending on the parameter α . The refs. 24 and 25 time series are well reproduced. For the ref. 24 time series, a preindustrial offset temperature needs to be applied (see *Supporting Information*). The recently observed high mass losses (25) are not fully captured by our global mean temperature-driven model.

The median future sea level contribution in 2100 from the Greenland ice sheet SMB is projected to be 7 cm, 12 cm, and 27 cm for the RCP26, RCP45, and RCP85 scenarios, respectively, relative to the 1986–2005 mean (Fig. 2D). Our projected 2081–2100 mean sea level is higher than ref. 3 estimates (Fig. 2D, bars at the right), with overlapping uncertainty ranges. The scenario dependency is also larger than estimated by IPCC, which is partly due to the assumed quadratic form of the millennial Greenland SMB sensitivity (see Eq. 3).

Greenland Solid Ice Discharge. We use three observational datasets of past Greenland SID (26–28) to constrain our model (see

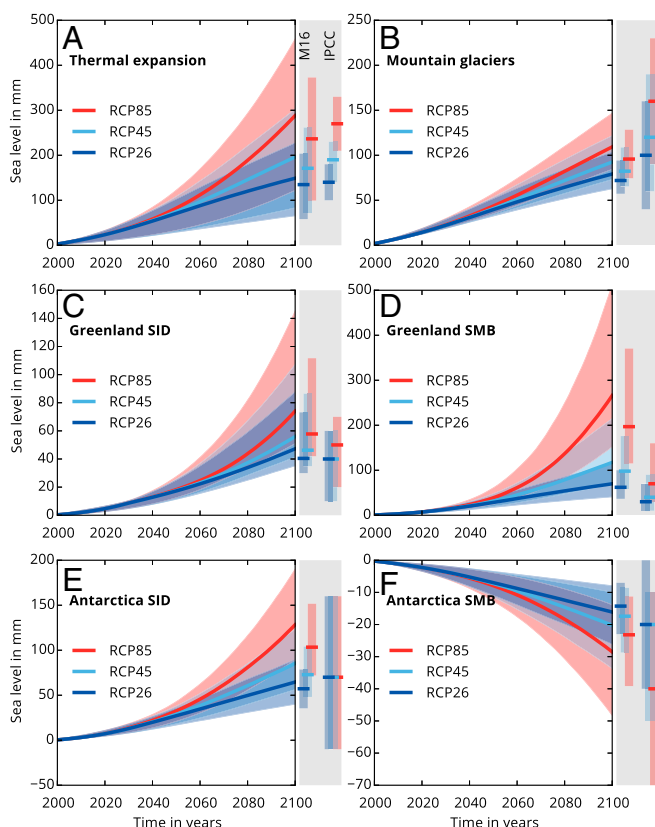


Fig. 2. Projected contributions to 21st century sea level rise for thermal expansion (A), mountain glaciers (B), Greenland solid ice discharge (C) and surface mass balance (D), and Antarctic solid ice discharge (E) and surface mass balance (F). Median (thick line) and fifth to 95th percentile uncertainty range (shading) of projected single contributions for the three RCP scenarios; based on 10,000 individual sea level curves. Bars at the right show fifth to 95th percentile range of this study (M16) and the IPCC AR5 (3) likely ranges intersected by the median for the 2081–2100 time mean. All are relative to the 1986–2005 mean. The y axis scale varies between panels.

Supporting Information for details). Because no long-term estimates are available for this contribution, we use a modified approach based on a response function driven by global mean temperature (see Eq. 4). Although North Atlantic climate variability influences SID through oceanic and atmospheric drivers (29), a link between global warming and the speedup of Greenland's glaciers is plausible (30–32) and assumed valid within our model. The response is consistent with the observed range (Fig. 1C).

The projected global warming-driven ice dynamical contribution from Greenland (Fig. 2C) is small compared with the surface melting component. We estimate a stronger scenario dependency than the IPCC Fifth Assessment Report (AR5, ref. 3), with the RCP26 median being similar to IPCC, whereas the RCP45 and RCP85 medians exceed the respective IPCC AR5 2081–2100 mean. Even for the highest emission scenario, the median estimate for 2100 does not surpass 8 cm (see Table 1).

Antarctic Surface Mass Balance. The recent mass changes of the Antarctic ice sheet are predominantly of dynamic origin, with SMB not showing a significant trend (33, 34). We can therefore not calibrate the Antarctic SMB component with past global mean temperatures as a driver. However, the relation between Antarctic atmospheric warming and SMB is robustly linked through the temperature dependence of the water

carrying capacity of the atmosphere (35, 36) (see *Materials and Methods* for details).

Although we currently cannot model the Antarctic SMB with the pursuit curve method, we include the projected contribution in the total projections so that we are able to approximate total future anthropogenic sea level rise. The projection yields between 1.6 cm and 2.9 cm sea level drop during the 21st century, depending on the emission scenario (Fig. 2F), which is of a lower magnitude than the estimates of ref. 37 and the IPCC AR5 (3) due to the additional discharge effect reported in ref. 38.

Antarctic Solid Ice Discharge. Because the SMB of the ice sheet has not shown a significant trend in the past (33, 34), we assume total mass changes to be a proxy for the changes in SID. We use three observational datasets for Antarctic mass loss (26, 39, 40). We find similar response times for the refs. 40 and 26 datasets and slightly shorter response times for the ref. 39 dataset. All range from 1,350 y to 2,900 y. The calibrated sea level function reproduces the observed trend well (Fig. 1E) in all three cases.

Although the 20th century contribution of Antarctic SID is limited, projections for the 21st century yield a median contribution of 6 cm, 9 cm, and 13 cm for RCP26, RCP45, and RCP85 in the year 2100 (Fig. 2E). By construction, the contribution is scenario-dependent. Our RCP26 and RCP45 median estimates are similar to the scenario-independent IPCC AR5 values. The RCP85 median exceeds the IPCC median (Fig. 2E, bars at the right) but is consistent with post-IPCC-AR5 multimodel estimates (41). Our 90% uncertainty ranges for the three scenarios are enclosed in the uncertainty range provided by the IPCC.

Total Sea Level Rise. Comparing past observed total sea level rise to the sum of our calibrated contributions is an independent test for the validity of the method. We constructed the observed anthropogenic sea level curve by subtracting the nonanthropogenic glacier part (23) from the observations of total sea level rise of refs.

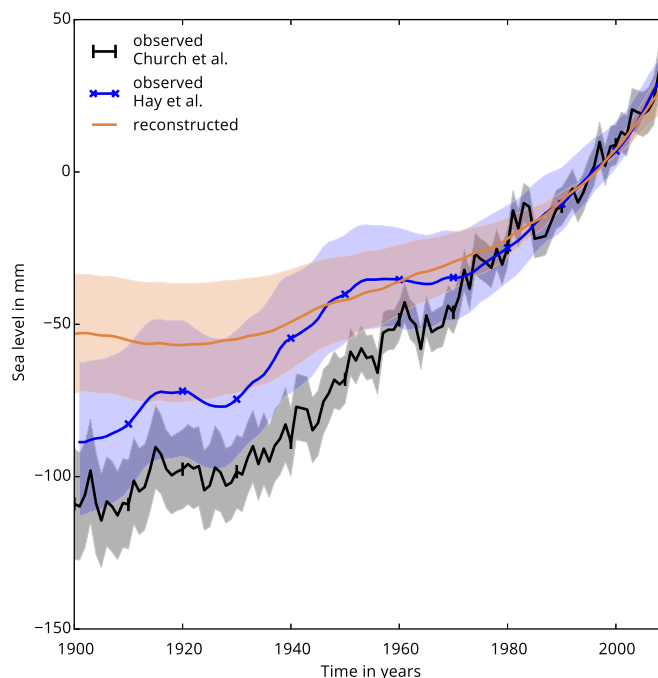


Fig. 3. Reconstructed and observed total anthropogenic sea level rise 1900 to today. Reconstructed sea level rise since 1900 as the sum of the probabilistically combined single contributions (median as brown line and 90% uncertainty range as shading) and observed sea level (blue, ref. 2; black, ref. 1; respective uncertainty ranges as shading). Only the anthropogenic part of the observed sea level is shown, produced by subtraction of the nonanthropogenic glacier contribution (23). All time series are relative to the 1986–2005 mean.

Fig. 4. Projected total anthropogenic sea level rise for the 21st century. Median (thick line) and fifth to 95th percentile uncertainty range (shading) of projected total anthropogenic sea level rise for three RCP scenarios as the sum of the single projected contributions (Fig. 2). Bars at the right show fifth to 95th percentile range of this study (M16) and the IPCC AR5 (3) likely ranges intersected by the median for 2081–2100. All are relative to the 1986–2005 mean.

25 show new records of surface melt on Greenland for recent years. These records are underestimated and therefore not fully linked to global mean temperature by our calibration (Fig. 1D), which is consistent with the suggested influence of natural variability through the North Atlantic oscillation (25, 44). Still, the inferred short response times lead to a future contribution above the range of current process-based projections (3). This highlights the importance of the attribution of recent melt records to anthropogenic forcing and raises the question of whether latest process-based estimates fully cover the mechanisms that drive 21st century Greenland surface mass loss.

As with other semiempirical approaches, our method cannot cover processes that are (or will become) independent of the forcing. Examples are the collapse of the Amundsen sector of the West Antarctic ice sheet, which is hypothesized to be already underway (45, 46), or a destabilization of the Wilkes basin in East Antarctica (47). The method can, however, account for processes that are not yet initiated but are reflected in the long-term sensitivity, which is an advantage over other semiempirical approaches. Contributions like groundwater depletion that are not linked to global warming (48) are not included in our calibration and do not bias our results. The model can be updated per contribution upon new physical insight, as, for example, for the dynamic discharge of the ice sheets. The method is limited to sea level contributors with monotonic long-term sensitivities. The Antarctic SMB may violate this condition for warming that is strong enough to initiate large-scale surface melting. Such melting is estimated to be small within this century (37) but may significantly reduce the ice body under strong greenhouse gas forcing in the long term (49).

The presented approach complements but cannot replace process-based modeling. It bridges the gap between classical semiempirical models and process-based models, because the parameters are chosen so that the model behavior is consistent with both past observations, which is a feature of semiempirical models, and long-term sensitivities as derived from process-based simulations. As opposed to complex process-based models, our method has low computational cost and can be used probabilistically. This allows the method to be incorporated in probabilistic impact studies that assess the causal chain of global warming from anthropogenic greenhouse gas emissions to the impacts of climate change.

Materials and Methods

Sea level rise in the 21st century is the combined response of highly inert systems to a common forcing. Therefore, it is reasonable to assume that the near-future response can be extrapolated from the past contributions, assuming the historical relationship between global mean warming and individual contributions remains the same. We use a pursuit curve to estimate the near-future sea level rise for each component as shown in Eq. 1. The applied long-term sensitivities S_{eq} are detailed below for each contribution.

Thermal expansion long-term sensitivity $S_{eq,te}$ can be inferred from long-term integrations of Earth system models of intermediate complexity and be approximated as

$$S_{eq,te} = \alpha_{te} \cdot \Delta T \quad [2]$$

with the commitment factor α_{te} and the deviation from preindustrial global mean temperature ΔT (5). Our estimates of α_{te} are based on six of such models and range from 0.2 m to 0.63 m per degree of warming (see [Supporting Information](#) for details).

For mountain glaciers, we apply a set of distinct functions $S_{eq,gic,i}$. Two different models (20, 50) have been used to estimate the glacier equilibrium sea level sensitivity globally (5). Forced by atmospheric data from 4 and 15 different climate models, respectively, they provide 19 different sensitivity curves for six levels of global warming, as shown in Fig. S4. As we are only interested in the ice loss that can be attributed to anthropogenic climate change, we remove the fraction caused by natural variability from the observational datasets and from the equilibrium sensitivities based on the data of ref. 23 (see [Supporting Information](#) for details).

The Greenland ice sheet is subject to an SMB feedback that leads to thresholds in the equilibrium response of total ice volume with respect to the surface air temperatures (51, 52). For sea level projections on centennial timescales, we rate the millennial (but not equilibrium) sensitivity to be a better approximation, as derived from refs. 5 and 52 and roughly of the form

$$S_{eq,gis,smb} = \alpha_{gis,smb} \cdot \Delta T^2 \quad [3]$$

where $\alpha_{gis,smb}$ ranges from 0.05 to 0.21 m °C⁻² and ΔT denotes the global mean temperature anomaly above preindustrial.

An estimate of the long-term sensitivity of Greenland's SID to global warming is not available. We thus modify the approach for this contribution following ref. 53. In response to ocean warming, the mechanical frontal stress at the marine termini of outlet glaciers is reduced, leading to enhanced ice discharge from Greenland (53). Increased melt water through a warmer atmosphere can lead to increased lubrication that speeds up glaciers and increases discharge (30–32). We here assume that frontal stress release (54) and runoff lubrication (27) can be approximated as linearly depending the global mean temperature anomaly ΔT . Ref. 55 has shown that the resulting sea level rise from Greenland's dynamic discharge $S_{gis,dyn}$ can be described via the response function

$$\frac{dS_{gis,dyn}}{dt}(t) = \Gamma \int_0^t \left(\frac{t-t'}{t_0} \right)^\beta \Delta T dt' \quad [4]$$

with β equal to -0.7 and temperature anomaly ΔT . We estimate the prefactor Γ in the interval $1.6\text{--}11 \times 10^{-5} \text{ m} \cdot \text{y}^{-2} \cdot \text{K}^{-1}$. This factor implies that a linear scaling between the global mean temperature and the local temperatures is applicable. To cover the uncertainty in the response, we here vary β between -0.9 to -0.5 ; t_0 is only used to nondimensionalize the time dependence and is chosen as 1 y.

In Antarctica, the observed relation between temperature and snowfall increase has been shown to be almost linear on centennial timescales. Ref. 37 estimated the sea level sensitivity to Antarctic warming to be $2.7 \text{ mm} \cdot \text{y}^{-1}$ per degree Celsius. Other studies found different scaling factors (see ref. 36 and table 4 in ref. 37), and, to reflect these, we vary the factor between $2 \text{ mm} \cdot \text{y}^{-1}$ per degree Celsius and $5 \text{ mm} \cdot \text{y}^{-1}$ per degree Celsius.

We do not apply a scaling factor between global and Antarctic atmospheric temperature change because the polar amplification is negligible for Antarctica (56). Ref. 38 showed that the increase in snowfall and the consequent steepening of the surface gradient at the grounding line leads to enhanced dynamic discharge along the coastline of Antarctica, which compensates between 15% and 35% of the mass gain through snowfall on a centennial timescale.

The SMB change from Antarctica is therefore estimated via

$$\frac{dS_{ais,smb}}{dt}(t) = -(1-\kappa) \cdot \gamma \cdot \Delta T \quad [5]$$

where ΔT denotes the global mean temperature change, γ is the snowfall sensitivity, and κ is the fraction lost due to the increase in dynamic discharge. We use a constant κ of 0.25 within this study.

Quasi-equilibrium estimates for the Antarctic ice sheet dynamic discharge contribution to sea level rise have been derived in ref. 5 from a 5-million-year simulation of the Antarctic ice sheet (57). A relatively constant commitment of $\alpha_{ais,dyn} 1.2 \text{ m}$ sea level per degree of warming is deduced from correlating global mean temperature with ice volume (figure 1D in ref. 5),

$$S_{eq,ais,dyn} = \alpha_{ais,dyn} \cdot \Delta T \quad [6]$$

We account for the uncertainty that originates from forcing data, ice physics, and memory of the ice sheet by sampling α from the interval $[1.0\text{--}1.5] \text{ m}$ per degree Celsius, which reflects the first standard deviation of the model simulations on which the relation is based.

For all contributions, we apply Monte Carlo sampling to project future sea level rise including observational and climate system uncertainties. Sea level uncertainty is covered by sampling from the calibrated set of semiempirical functions, which incorporates the various observations and long-term responses. The uncertainty in the climate system response to future greenhouse gas emissions is accounted for by sampling from an ensemble of 600 global mean temperature pathways for the three representative concentration pathways RCP26, RCP45, and RCP85 (17). The pathways, produced by the MAGICC 6.0 simple climate model (58), are consistent with past climate change and the results from climate models of higher complexity (59). To capture sea level and climatic uncertainty, we repeat the procedure 10,000 times. Uncertainty intervals are calculated on the basis of the 10,000 sea level curves.

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Supporting Information

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Description of Datasets

We here describe the observational datasets as used for the calibration. We also provide extended information on the equilibrium sea level responses. We discuss deviations from the standard calibration if applied.

Thermal Expansion.

Observations of thermal expansion. Past thermosteric sea level rise is estimated from past changes of ocean temperature. Observations of ocean temperature are available for several ocean depth ranges. All datasets applied here are shown in Fig. S1. The upper ocean layer (0–700 m) is best sampled, and three datasets are available. Ref. 12 provides an updated dataset for the period 1950–2012. Ref. 13 provides updated data from 1945 to 2014. Ref. 14 provides estimates for the 0- to 700-m layer and additionally for the 0- to 2,000-m layer from 1957 to 2012. All datasets apply expendable bathy-thermograph (XBT) bias corrections; see table 1 in ref. 60 for details.

Below 2,000 m, samples are sparse and observed temperature trends are subject to high uncertainty. Ref. 15 provides a trend for the abyssal ocean below 2,000 m for the period 1990–2010. We assume this trend spanning also into the past. Alternatively, we include the assumption that the below-2,000-m ocean did not contribute to sea level rise. This is consistent with observations during the ARGO deployment period 2005–2013 (61). Ref. 26 provides trend estimates for 700–3,000 m and below 3,000 m. We use the ref. 15 estimate below 2,000 m to construct a second time series for 700–2,000 m to complement the ref. 14 estimate.

To encompass the uncertainty from the different observational datasets, we create all possible combinations of the observations from different depths to yield 12 estimates for total thermosteric sea level rise (see Fig. S2). We calibrate our semiempirical model over the full period of observations. The calibrations with ref. 12 data are restricted to 1970 to today, as the early dip cannot be covered well by our model and therefore would result in a bad fit for the whole period.

We start the integration of Eq. 1 with the beginning of global mean temperature observations (16) in the year 1880, as atmospheric warming influenced ocean temperatures already before the start of the observations.

Estimates of equilibrium contribution for thermal expansion. We use the results of 10,000-y simulations from six climate models of intermediate complexity (table 8.3 in ref. 62). The results were also used for the IPCC AR4 report (figure 10.34 in ref. 62). Assuming a long-term linear increase with global mean temperature (5), the models yield commitment parameters between 0.2 m and 0.63 m per degree of global warming. Specifically, for each model: Bern2D-CC, 0.488; CLIMBER-2, 0.458; CLIMBER-3a, 0.200; MIT, 0.214; MoBidiC, 0.626, and UCL, 0.386.

Glaciers and Ice Caps.

Datasets of glacier observations. We here use three different reconstructions of global glacier mass changes (18–20). Updates to these are described in detail in ref. 21. We here repeat them briefly. The respective contribution to sea level rise for each dataset is shown for each dataset in Fig. S3, *Top*.

The global mass balance compilation of ref. 18 includes geodetic and direct measurements. The compilation has been updated several times; we here use the latest release 1301. The dataset of ref. 19 is extended to the late 19th century by using observations of glacier length changes from 13 different regions. Upscaling is applied to yield global glacier volume change from the

limited number of glacier lengths. Ref. 20 follows a different approach by modeling the response of each glacier to the evolving climate, based on gridded climate observation data [CRU CL 2.0 and CRU TS 3.0 (63, 64)]. All glaciers from version 1.0 of the Randolph Glacier Inventory (65) are being modeled.

Estimates of equilibrium contribution for glaciers and ice caps. We use two glacier models to compute the sensitivity of global glacier volumes to different levels of global mean temperature. The model of ref. 50 applies monthly temperature and precipitation time series from four global climate models from the Climate Model Intercomparison Project 3 (CMIP3) (models U.K.MO-HadCM3, ECHAM5/MPI-OM, GFDL-CM2.0, and CSIRO-Mk3.0). The model of ref. 20 applies forcing from 15 global climate models from CMIP5 (bcc-csm1-1, CanESM2, CCSM4, CNRM-CM5, CSIRO-Mk3, GFDL-CM3, GISS-E2-R, HadGEM2, Inmcm4, IPSL-CM5A-LR, MIROC5, MIROC-ESM, MPI-ESM-LR, MRI-CGCM3, and NorESM1-M). The SMB as forced by the climate model input is coupled to a simple parameterization of ice dynamics in both models. To yield equilibrium values, the models apply time slices from climate scenarios that fit to temperature levels between 0 °C and 6 °C above preindustrial (50) and 0 °C and 10 °C above the 1961–1990 mean temperature (20). These time slices are repeated until equilibrium is reached. Climate model data are taken from the A1B scenario for the ref. 50 model and from RCP85 scenario in ref. 20. The 19 glacier model curves are shown in Fig. S4.

The equilibrium data are only available for discrete steps of 1 °C, and not all models cover the full range of temperatures. Further, the ref. 20 estimates do cover the temperature range before 1961–1990. We therefore parametrize the equilibrium response E_0 with

$$E_0 = a(1 - e^{bT}).$$

The parametrization passes zero for $T = 0$, and therefore ensures an anthropogenic glacier contribution of zero for zero temperature deviation above preindustrial. It also ensures saturation for high temperatures. We determine the parameters a and b for each of the 19 glacier model results. The resulting curves are shown in Fig. S4.

Glacier calibration. We restrict the calibration period for the refs. 19 and 20 datasets to the years post-1930. The earlier period cannot be covered well with our global mean temperature-driven model, as deviations from preindustrial temperature are small. Therefore, longer calibration periods result in a misfit of the later 20th century part. The ref. 18 calibration period starts with data availability in 1961. Although the restriction of the calibration period to post-1961 for refs. 19 and 20 would lead to an improved match of observations and model for the second half of the 20th century, we stick to the longer calibration period to better reflect the uncertainty in observations, resulting in a higher parameter spread.

Greenland Ice Sheet SMB.

Datasets on Greenland's observed mass balance changes. It is not fully clear if the Greenland ice sheet SMB was in equilibrium before systematic measurements were taken. For example, ref. 66 derives a reference period 1971–1988 by iteration, with zero total Greenland mass change as the criterium. Ref. 67 states a net positive SMB before 1960 using the reference period 1960–1990, implying a growing ice sheet through SMB before 1960. In contrast, refs. 24 and 27 reconstruct SMB change since 1840 on

the basis of regional climate modeling and ice core data and identify several periods of significant ice sheet mass loss (ref. 27, figure 7) before 1960, with total ice mass loss calibrated with independent gravimetry estimates from Gravity Recovery and Climate Experiment (GRACE) (68) for the period 2003–2010. As we cannot rule out one of the cases, we incorporate both in our calibration. The missing knowledge on past changes translates therewith to wider parameter ranges.

We calibrate our model to changes beginning in 1960 with the SMB time series of ref. 25, which applies an updated version of RACMO2. The influence of global mean temperature on SMB is assumed zero before that year. To estimate the sea level contribution from SMB, we choose the reference SMB period 1870–1900 from ref. 24 as a proxy for preindustrial conditions.

As a second dataset, and representative for longer-term change, we use the ref. 24 SMB time series. The periodically negative SMB implies a Greenland SMB contribution to sea level rise since the 1920s. To fit our model to this time series, we need to apply an offset temperature of 0.5 °C. Such an offset has been applied also in other semiempirical studies (e.g., ref. 7) and may be interpreted as a remainder of small ice age influence.

Third, we use the total mass balance estimate from ref. 26 to estimate SMB by assuming a fixed partitioning between SMB and SID. There are two reasons for a reverse calculation of SMB from their total mass balance estimate. The split of total mass balance into SMB and SID ensures consistency of the sum with total mass balance, and the ref. 26 dataset is consistent with Earth's energy budget. We assume a fixed 50% split between SMB and SID. Although this may not be true on the very short term and does not reflect the very latest years of observations [68% in the period 2009–2012 (69)], it is more probable on the longer term. Ref. 27 finds that SID and SMB are well correlated on the longer term on a 1:1 basis. Ref. 70 finds a 52% contribution over the period 2003–2009, and ref. 71 finds an equal split for the period 2000–2008.

SMB equilibrium response of the Greenland ice sheet. We determine the sensitivity of the Greenland ice sheet to SMB changes driven by global warming from simulations with an energy–moisture balance model coupled to a dynamic ice sheet model (52). We do not use the equilibrium estimates that include a threshold in the ice sheet volume but the estimates of the millennial sensitivity to global warming as discussed in ref. 5.

Greenland SMB calibration. The full period of observation is used for calibration for each dataset. The calibrated model slightly underestimates the latest strong increase in negative SMB. This may reflect that the recent exceptional mass losses due to increased surface melting are rather part of natural variability than driven by global temperature increase.

Greenland Ice Sheet SID.

Greenland SID observations. We use three datasets for past Greenland SID. The ref. 27 long-term estimate for the discharge is based on a runoff-dependent parametrization. The sum of discharge and SMB

is consistent with recent GRACE estimates of total mass change. As ref. 25 does not provide an estimate for SID, we use the time series of ref. 28. The SMB time series of refs. 25 and 28 differ only marginally, so that the combination of ref. 28 SID with ref. 25 SMB can be justified. As a third estimate, we use the 50% part of the ref. 26 total mass balance for reasons as discussed in *Datasets on Greenland's observed mass balance changes*.

Greenland SID calibration. We use a custom model to project future Greenland SID (see *Materials and Methods*). Equilibrium sensitivities are not available for Greenland SID.

Antarctic Ice Sheet.

Datasets on Antarctica's mass balance changes. The SMB of the ice sheet has not shown a significant trend in the past (33, 34); we therefore assume total mass changes to be induced by changes in SID alone. We use three datasets that estimate the changes of the Antarctic ice sheet mass balance. Ref. 26 bases their time series on refs. 72 and 73, which estimate the mass balance from 1980 based on input–output analysis (SMB versus ice discharge). The ref. 26 time series includes a nonzero trend before 1980 based on ref. 62, chapter 4. It therefore represents a long-term contributing Antarctic ice sheet. Ref. 39 provides a time series for mass changes of the West Antarctic ice sheet from 1974 to 2013. Assuming a zero trend in East Antarctica and the West Antarctic glaciers still being in balance in the 1970s (73), we use the ref. 39 dataset as representative for the assumption that Antarctica's imbalance is driven by glacier speedup in West Antarctica since the 1990s. Third, we use the gravimetry-derived time series of ref. 40 for the period 2003–2014. The gravimetry method does not suffer uncertainty from the choice of reference period. The indication for West Antarctic glaciers being still in balance in the 1970s (73) makes a zero forcing before 1980 plausible, which we apply for both datasets in refs. 39 and 40. To cover the assumption of a longer-term influence, we allow for the time series in ref. 26 that global temperatures drive ice loss from 1880.

Equilibrium estimates for Antarctic SID. Quasi-equilibrium estimates for the Antarctic ice sheet contribution to sea level rise have been derived from a 5-million-year simulation of the Antarctic ice sheet (5). The applied model (57) is able to simulate the larger-scale grounding line retreat and advance on multimillennial timescales, as comparison with reconstructed past grounding lines showed. The sensitivity of the ice sheet to global mean temperature changes has been deduced from correlating 1,000-y averages of global mean temperature with the respective average sea level volume of the same time frame.

Antarctica SID calibration. For incorporating the unknown timing when the Antarctic ice sheet entered a state of imbalance, we allow global mean temperature to affect SID for the whole period of global warming to calibrate the ref. 26 time series, consistent with the trend already in the beginning of the time series. For the other two datasets (39, 40), we assume global warming only driving the ice loss after the 1970s (73).

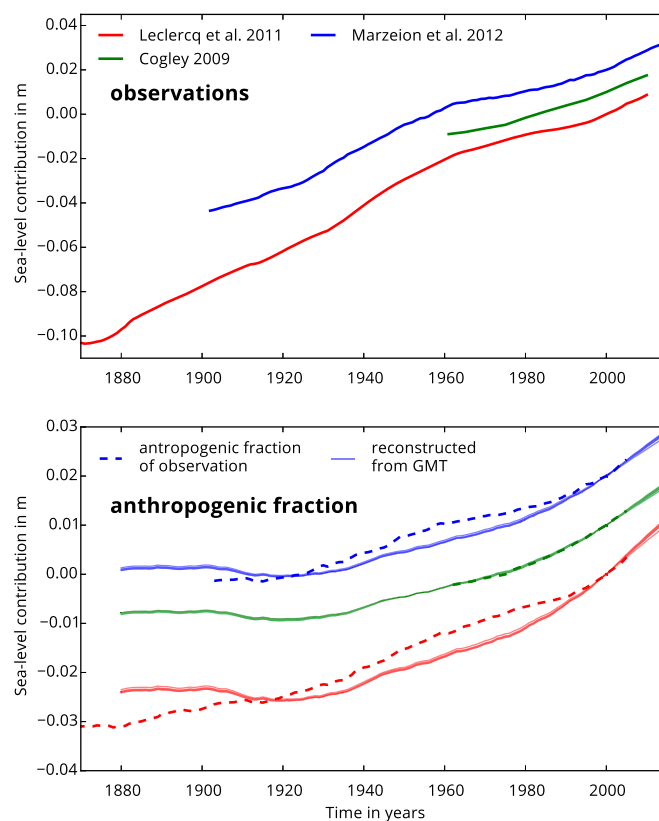


Fig. S3. Glacier observations and fitted model time series. Glacier observational datasets as used for the calibration (18–20). (*Top*) Total sea level contribution for each dataset. (*Bottom*) Anthropogenic fraction (23) of the observations (dashed lines). The thin full lines show the anthropogenic sea level contribution for each dataset as in our fitted semiempirical model forced by global mean temperature. All time series are relative to year 2000. In both panels, an arbitrary offset is applied between the different datasets for better visibility.

