Comparison of hybrid schemes for the combination of Shallow Approximations in numerical simulations of the Antarctic Ice Sheet

Jorge Bernales[1,2], Irina Rogozhina[3,1], Ralf Greve[4], and Maik Thomas[1,2]

[1]GFZ German Research Centre for Geosciences, Section 1.3: Earth System Modelling, Potsdam, Germany
[2]Institute of Meteorology, Free University Berlin, Berlin, Germany
[3]MARUM Centre for Marine Environmental Sciences, University of Bremen, Bremen, Germany
[4]Institute of Low Temperature Science, Hokkaido University, Sapporo, Japan

Correspondence to: Jorge Bernales (bernales@gfz-potsdam.de)

Abstract. The Shallow Ice Approximation (SIA) is commonly used in ice-sheet models to simplify the force balance within the ice. However, the SIA alone is not able to adequately reproduce the dynamics of the fast flowing ice streams usually found at the margins of ice sheets. To overcome this limitation, recent studies have introduced heuristic, hybrid combinations of the SIA and the Shelfy Stream Approximation. Here, we implement four different hybrid schemes into a model of the Antarctic Ice Sheet in order to compare their performance under present-day conditions. For each scheme, the model is calibrated using an iterative technique to infer the spatial variability in basal sliding parameters. Model results are validated against topographic and velocity data. Our analysis shows that the calibration compensates for the differences between the schemes, producing similar ice sheet configurations through quantitatively different parameter distributions. Despite this we observe a robust agreement in the reconstructed patterns of basal sliding parameters. We exchange the retrieved parameter distributions between the schemes to demonstrate that the results of the model calibration cannot be straightforwardly transferred to models based on different approximations of ice dynamics. However, this limitation can be overcome through the implementation of easily adaptable calibration techniques, as shown in this study.

1 Introduction

Accurate projections of ice-sheet–driven sea level changes require the use of numerical models that are capable of capturing the dynamics of rapidly flowing regions and grounding-line zones (Pattyn et al., 2013). This requirement can be best accommodated using the most complete models which are currently available for modelling the ice dynamics, referred to as Full Stokes models (FS) (e.g., Gagliardini et al., 2013). However, the time scales over which an ice sheet builds up and disintegrates in response to variations in the climatic forcing typically involve many thousands of years. Numerical experiments over such time spans are necessary to separate the long-term transient component from relatively fast fluctuations in the ice volume during the observational record. These long-term, continental-scale paleo-simulations are currently infeasible using FS models due to the computational expenses triggered by the non-linearity of the model equations and the complex interdependence of the involved quantities.

To overcome the contemporary spatio-temporal limitations of FS models, a hierarchy of approximations has been developed over the last decades (e.g., Hindmarsh, 2004). The Shallow Ice Approximation (SIA) (e.g., Hutter, 1983) is a zeroth-order approximation of the momentum balance equations that keeps only the gravity-driven vertical shear stress, predicting reasonably well the behaviour of grounded ice masses which are characterised by a thickness much smaller than their horizontal length scales. Ice floating in the sea water experiences almost no friction at the base, and its behaviour is usually described by the Shallow Shelf Approximation (SSA) (Morland, 1987), which omits the vertical shear stress from the FS equations and neglects the basal drag. The transition between grounded and floating ice regions exhibits areas where ice flow is often enhanced by basal conditions favourable for sliding, generating rapid ice flow features known as ice streams. In these ice sheet sectors membrane stresses become increasingly important sharing many similarities with the floating ice shelves, and thus the SIA is no longer appropriate to describe the ice dynamics. It is also important to note that the absence of a membrane stress transfer in the SIA renders this approximation invalid for modelling the grounding-line migration, that is, the migration of an interface between grounded and floating ice sectors (Pattyn et al., 2012).

More sophisticated methods have been designed to overcome the limitations of SIA models when reproducing the dynamics of ice streams, which still aim at low computational costs. The approaches used by this new generation of continental-scale ice-sheet models include depth-integrated Blatter–Pattyn models (Blatter, 1995; Pattyn, 2003) based on the asymptotic analysis by Schoof and Hindmarsh (2010), algorithms that detect and apply the SIA where it is valid and use the FS elsewhere (Ahlkrona et al., 2015), and so-called hybrid models utilising heuristic combinations of the SIA and the Shelfy Stream Approximation (SStA, which is the SSA including basal drag) (Bueler and Brown, 2009; Winkelmann et al., 2011; Pollard and DeConto, 2012a). These hybrid models utilise a number of algorithms to identify zones of potential fast flow where ice streams may operate and heuristically weight contributions of each approximation based on predefined criteria. The use of hybrid models enables simulations over hundreds of thousands of years on continental scales, yet showing a reasonable performance compared to higher-order models in idealised scenarios and inter-comparison tests (e.g., Pattyn et al., 2013; Feldmann et al., 2014). Since the combination of the SIA and the SStA is based on heuristics, the approaches used to combine the two approximations vary from model to model, ranging from weighted averages of both velocity solutions to a simple summation over the entire domain.

Despite the above differences among existing models, all of them are subject to common limitations when applied to the present-day Antarctic Ice Sheet (AIS). These limitations include the scarcity of observational data needed to reduce the errors introduced by poorly constrained model parameters and boundary conditions, e.g., the flow enhancement factors introduced to account for the anisotropy of ice flow, geothermal heat flux, the glacial isostatic adjustment, and the distribution of water-saturated sediments at the ice sheet base. The latter has the potential to enhance basal sliding and is currently considered to be a major source of large, widespread misfits between the observed and modelled elevations of the AIS (e.g. de Boer et al., 2015). Recent studies have attempted to quantify potential distributions of these intrinsic bed properties using sophisticated inverse methods (e.g., Joughin et al., 2009; Morlighem et al., 2010; Arthern and Gudmundsson, 2010; Pralong and Gudmundsson, 2011; Arthern et al., 2015). These diagnostic methods focus mainly on the fit between the modelled and observed ice velocities. Pollard and DeConto (2012b) presented a much simpler algorithm, aiming to fit the observed surface elevations instead of velocities. The prognostic model is run forward in time, and the local elevation error is used to periodically adjust the basal

sliding parameters until the best fit between the observed and modelled elevations is attained. This procedure has the ability to drastically reduce large elevation errors during the calibration and initialisation of ice-sheet models, which is an important requirement for simulations that would otherwise be undermined by poor parameter choices.

In this paper, we evaluate the performance of four different hybrid approaches implemented as part of the same continental-scale ice-sheet model, applied to the entire AIS. To this end, a calibration procedure based on the aforementioned iterative technique of Pollard and DeConto (2012b) is applied to each hybrid scheme using the observed ice thickness (Fretwell et al., 2013). For comparison purposes, the same procedure is carried out using a SIA-only model. The results of these experiments are validated against an independent, observational data set of surface ice velocities (Rignot et al., 2011). Additionally, we provide insights into the relative contributions of the shallow approximations in different hybrid schemes. By exchanging the inferred distributions of basal sliding parameters between the applied hybrid schemes, we test the applicability of the model calibration results in different types of ice flow models. For hybrid approaches involving adjustable parameters, we also explore the sensitivity of the results to parameter variations. First, the ice-sheet model and the hybrid schemes are described in Section 2, where we also detail the adapted iterative technique for the calibration of the basal sliding parameters. The observational and model-based data sets used in our simulations are described in Section 3. The set-up of the numerical experiments can be found in Section 4. The results are presented and discussed in Section 5, followed by the summary and conclusions provided in Section 6.

2 Methods

2.1 Model overview

In this study, the simulations of the AIS are carried out using the open source, three-dimensional, thermo-mechanical ice sheet-shelf model SICOPOLIS (SImulation COde for POLythermal Ice Sheets) version 3.2-dev, revision 619 (Greve, 1997; Greve and Blatter, 2009; Sato and Greve, 2012). It uses finite differences to solve the numerically approximated SIA and SSA equations for grounded and floating ice, respectively. Relevant modifications introduced in the model specifically for this study are presented in a greater detail in Sections 2.2 and 2.3.

SICOPOLIS is able to model polythermal ice sheets, i.e., it explicitly identifies potential temperate regions in which the modelled ice temperature is at the pressure-melting point (Greve, 1997). Within these regions, ice and small amounts of liquid water can coexist, and the water content is used as an additional input for a regularised Glen's flow law (Glen, 1955) utilised in our experiments, following Greve and Blatter (2009), in the form

$$\eta(T_\mathrm{m}, \sigma_\mathrm{e}) = \frac{1}{2EA(\sigma_\mathrm{e}^{n-1} + \sigma_0^{n-1})}, \tag{1}$$

where η denotes ice shear viscosity, T_m is the temperature difference relative to the pressure-melting point, σ_e is the effective shear stress, $\sigma_0 = 10\,\mathrm{kPa}$ is a small constant used to prevent singularities when σ_e is very small, $n = 3$ is the flow law exponent, and A is a temperature- and pressure-dependent rate factor (Cuffey and Patterson, 2010). In temperate ice regions, A is modified to account for the aforementioned water content, following Lliboutry and Duval (1985). The empirical coefficient E is a flow

enhancement factor, which is introduced to account for the effects of anisotropic ice fabric. The value of E depends on the deformation regime of ice, which is different for grounded ice, where horizontal shear prevails, and floating ice, dominated by longitudinal stretching (Ma et al., 2010). In general, large-scale marine ice-sheet models use a homogeneous, constant value ranging between 1 and 8 for the grounded ice, and between 0.2 and 1 for ice shelves (e.g., de Boer et al., 2015). Within these ranges, the computed age of ice is often used to assign different values for glacial and interglacial ice. Here, we use $E = 1$ and $E = 0.5$ for grounded and floating ice, respectively. These values are smaller than those chosen in previous studies using SICOPOLIS (e.g., Sato and Greve, 2012), and are based on our initial tests and the sensitivity analysis by Pollard and DeConto (2012b).

At the base of the grounded ice sectors, bedrock stress conditions and the associated potential for sliding are linked to the basal velocity, $\boldsymbol{u}_b$, used as a boundary condition for the computation of the SIA velocities, through an empirical Weertman-type sliding law (Weertman, 1964; Dunse et al., 2011), in the form

$$\boldsymbol{u}_{\mathrm{b}} = -\frac{C_{\mathrm{b}}}{N_{\mathrm{b}}^{q}} |\boldsymbol{\tau}_{\mathrm{b}}|^{p-1} \boldsymbol{\tau}_{\mathrm{b}}, \tag{2}$$

where $\boldsymbol{\tau}_{\mathrm{b}}$ is the basal shear stress, and $p = 3$ and $q = 2$ are the sliding law exponents. N_{b} is the effective basal pressure, computed as

$$N_{\mathrm{b}} = \rho_{\mathrm{ice}} g H - \rho_{\mathrm{sw}} g H_{\mathrm{sw}}, \tag{3}$$

where ρ_{ice} and ρ_{sw} are the density of ice and sea water, respectively, g is the gravitational acceleration, H is the modelled ice thickness, and H_{sw} is the difference between the mean sea level and the ice base topography. The parameter C_{b} depends on the basal temperature and pressure conditions:

$$C_{\mathrm{b}} = C_0 e^{T_{\mathrm{m}}/\gamma}, \tag{4}$$

where the exponential function controls the amount of sliding, depending on the temperature below the pressure-melting point T_{m} and the sub-melt-sliding parameter γ, ensuring a smooth transition across different basal thermal regimes and thus, prohibiting discontinuities in the velocity field (Bueler and Brown, 2009). A spatially varying factor C_0 is introduced to account for differences in the bedrock material properties affecting sliding (e.g. hard bedrock vs. soft sediments). Potential distributions of C_0 have been explored using different iterative and inverse approaches and a variety of sliding laws, aiming to find spatially varying values that minimise the discrepancy between the modelled and observed quantities such as ice thickness, ice surface velocity and elevation change (e.g., Joughin et al., 2009; Morlighem et al., 2010; Arthern and Gudmundsson, 2010; Pralong and Gudmundsson, 2011; Pollard and DeConto, 2012b; Arthern et al., 2015). A particular iterative technique implemented in our model is described in Section 2.3. Other model components include evolution equations for ice temperature and ice thickness, with the latter forced by independent modules for the computation of the surface and basal mass balances (Greve and Blatter, 2009; Sato and Greve, 2012). A summary of the model parameters used in this study is provided in Table 1.

2.2 Hybrid schemes

For this study, four hybrid approaches have been implemented into the model. Each of them offers a different way to identify the fast flowing zones and combine the horizontal SIA velocity, u, and the horizontal SStA velocity, v. For each scheme, individual velocity solutions from the shallow approximations are calculated independently. It is important to note that SStA velocities in grounded ice regions include basal drag. This implies a major difference from SSA velocities computed in the floating ice shelf sectors, for which the friction at the ice-ocean interface is negligible.

For consistency, the basal drag term that enters the SStA equations is computed using the same sliding law as described in Section 2.1 (Eq. (2)), in the form

$$\boldsymbol{\tau}_\mathrm{b} = -\beta_\mathrm{drag} \boldsymbol{v}_\mathrm{b}, \tag{5}$$

where $\boldsymbol{v}_\mathrm{b}$ is the the basal SStA velocity and the drag coefficient, β_drag, is computed as

$$\beta_\mathrm{drag} = \frac{N_\mathrm{b}^{\frac{q}{p}}}{C_\mathrm{b}^{\frac{1}{p}}} \left(\frac{1}{\sqrt{v_{\mathrm{b_x}}^2 + v_{\mathrm{b_y}}^2 + v_0^2}} \right)^{1 - \frac{1}{p}}. \tag{6}$$

Here, $v_{\mathrm{b_x}}$ and $v_{\mathrm{b_y}}$ are the horizontal components of the SStA velocity at the ice base, and $v_0 = 0.01\,\mathrm{m\,a^{-1}}$ is a small regularisation quantity introduced to prevent singularities at the locations where there is no basal sliding (Bueler and Brown, 2009).

The first hybrid scheme (henceforth HS-1) is the original implementation in SICOPOLIS v3.2-dev (revision 619) based on the slip ratio of grounded ice, computed as

$$r = \frac{|\boldsymbol{u}_\mathrm{b}|}{|\boldsymbol{u}_\mathrm{s}|}, \tag{7}$$

where $\boldsymbol{u}_\mathrm{b}$ is the Weertman sliding velocity (Eq. (2)) and $\boldsymbol{u}_\mathrm{s}$ is the surface SIA velocity. At each iteration, and for each velocity component, the local slip ratio r is compared to a prescribed threshold r_thr ranging from 0 to 1 (Section 4). If r is larger than the threshold, the grid point is flagged as streaming ice where the SStA velocities should be computed. Once SStA velocities are computed, the individual contributions of the SIA and SStA velocities at each streaming grid point are determined using the weight

$$w(r) = \frac{r - r_\mathrm{thr}}{1 - r_\mathrm{thr}}, \tag{8}$$

Then, for each streaming grid point, the hybrid horizontal velocity U is computed as

$$\boldsymbol{U} = w \cdot \boldsymbol{v} + (1 - w) \cdot \boldsymbol{u}, \tag{9}$$

recalling that u and v are the horizontal SIA and SStA velocities, respectively.

The second approach (henceforth HS-2) is based on the idea by Bueler and Brown (2009), in which SStA velocities are calculated over the entire ice sheet and used as a sliding velocity complementing a non-sliding SIA model. SIA and SStA

velocities are combined as in Eq. (9), and the weighting function is adopted from Bueler and Brown (2009, Eq. (22))

$$w(|\boldsymbol{v}|) = \frac{2}{\pi}\arctan\left(\frac{|\boldsymbol{v}|^2}{v_{\mathrm{ref}}^2}\right), \tag{10}$$

where v_{ref} is a reference ice velocity (Section 4). The velocity v_{ref} marks the point for which the SIA and SStA contributions are equally weighted, i.e., the resulting hybrid velocity is a standard mean of both solutions. The weighting function is smooth,

5   monotone, and its value converges towards 0 for small velocities and towards 1 when $\boldsymbol{v}$ is large compared to the reference velocity v_{ref}. As in the HS-1, w is used to compute respective contributions from the SIA and SStA, with the difference that Eq. (10) uses v_{ref} as the only criterion to determine the SStA contribution. The SStA velocities are calculated over the entire ice sheet, and an a priori identification of fast flow zones is not required.

As described in Section 2.1, the SIA solution in SICOPOLIS is computed using the Weertman sliding law (Eq. (2)) as a

10   boundary condition. To assess the influence of a SIA solution including the Weertman sliding, we have split this hybrid scheme into two: A sub-scheme (HS-2a) that replicates the idea of Bueler and Brown (2009) with no basal velocity prescribed in the computation of the SIA, and a sub-scheme (HS-2b) that has a Weertman sliding component and uses it to compute a slightly modified weight:

$$w(|\boldsymbol{u}_{\mathrm{b}}|) = \frac{2}{\pi}\arctan\left(\frac{|\boldsymbol{u}_{\mathrm{b}}|^2}{v_{\mathrm{ref}}^2}\right), \tag{11}$$

15   where $\boldsymbol{u}_{\mathrm{b}}$ is the basal sliding velocity as in Eq. (2). Thus, in the HS-2b the SStA solution does not serve as a replacement of a sliding law. It is rather used to determine to which extent the computed SStA velocity should replace the basal velocity used to compute the SIA solution.

The third approach (henceforth HS-3) simply adds up the non-sliding SIA and SStA solutions:

$$\boldsymbol{U} = \boldsymbol{u} + \boldsymbol{v} \tag{12}$$

20   This superposition of approximations has been employed in recent studies using SIA models in combination with a SStA solution as a sliding law (e.g., Winkelmann et al., 2011). It bypasses the need for additional free parameters, such as r_{thr} and v_{ref} in the HS-1 and HS-2, respectively. This approach is based on the assumption that on ice shelves the SIA contribution is negligible due to low surface gradients, and therefore the modelled ice flow is dominated by the SStA solution, whereas in the continental interior the modelled ice flow is dominated by the SIA solution (Winkelmann et al., 2011). Since the SIA and SStA

25   solutions are computed over the entire domain, their superposition enables a smooth transition across different flow regimes, ranging from slow ice motion in the interior to a characteristic fast flow of ice shelves, thereby allowing for stress transmission across the grounding line. As in the HS-2, an identification of fast flowing zones is purely diagnostic and not required during the computation of $\boldsymbol{U}$. Table 2 presents a summary of the hybrid schemes applied in this study.

2.3   Calibration of basal sliding coefficients

30   We have implemented an iterative method following Pollard and DeConto (2012b) in order to determine the distribution of sliding coefficients C_0 that minimises the difference between the modelled and observed ice thickness. The method starts from

a spatially uniform guess value for the distribution of C_0 and runs the model forward in time, as described in Section 4. At a given time step, Δt_{inv}, the method uses the basal temperature below the pressure-melting point, T_{m}, to identify grounded grid points where basal sliding may occur. If the absolute value of T_{m} is smaller than the parameter γ from Eq. (4), i.e., close to the pressure-melting point, the method computes the difference between the modelled and observed ice thickness, which is then

5   used to locally adjust C_0 at this grid point according to

$$C_0^* = C_0 10^{\Delta H}, \tag{13}$$

where C_0^* is an updated sliding coefficient and $\Delta H = (H - H_{\mathrm{obs}})/H_{\mathrm{inv}}$. Here, ΔH is the difference between the modelled and observed ice thickness, scaled by a factor, H_{inv}, in order to prevent overshoots. For the same reason and following the implementation by Pollard and DeConto (2012b), variations in the value of the multiplicative factor $10^{\Delta H}$ are further limited

10   by a range of $\sim .03$ to 30. In contrast to previous studies using SICOPOLIS where $\gamma = 1\,\mathrm{K}$, here we set this parameter to a value of $3\,\mathrm{K}$, allowing for a more frequent calibration of the sliding coefficients (Pollard and DeConto, 2012b).

     Studies using this iterative technique and inversion methods have shown that potential distributions of sliding coefficients C_0 are highly heterogeneous, with values spanning several orders of magnitude (e.g., Pollard and DeConto, 2012b; Arthern et al., 2015). To ensure numerical stability, we limit our inferred values to a range of 1 to $10^5\,\mathrm{m\,yr^{-1}\,Pa^{-1}}$ during the calibration

15   procedure. Additionally, we have implemented the following condition: When the computed surface ice velocity reaches an ancillary speed limit at a certain grid point, the adjustment of C_0 for that point is halted. This prevents the method from over-adjusting the sliding coefficients when the velocity limit has been reached and no noticeable changes occur in response to further adjustments of C_0. This additional constraint is applied in order to ensure numerical stability and keep the modelled ice velocities within the range of observations. For the experiments presented in this study, the lower speed limit is defined as

20    $0.1\,\mathrm{m\,a^{-1}}$, whereas the upper limit is set to $4000\,\mathrm{m\,a^{-1}}$. These values are based on the observed surface velocities of Rignot et al. (2011).

     The iterative technique involves an additional limiting condition that prevents over-adjustments of C_0. For each individual grid point, if the difference between the modelled and observed ice thickness reduced at the previous time step, the adjustment at the current time step is deactivated. This allows previous adjustments to fully develop their effects over the following time

25   steps and prevents the technique from adding unnecessary extra adjustments that often result in overshoots. The calibration is activated again as soon as the time derivative of the modelled ice thickness drops to zero (i.e., the difference between the modelled and observed ice thickness is not reduced anymore) or the misfit starts increasing (e.g. due to increased influx from surrounding areas). Our experiments have shown that this additional feature enables the use of a smaller Δt_{inv} (50 years used here compared to 500–10,000 years in Pollard and DeConto (2012b)), because further adjustments will only be applied when

30   and where strictly necessary. A further benefit is that it indirectly allows non-local adjustments of C_0 to influence the local ice dynamics: If an adjustment applied in the vicinity of a grid point reduces the misfit, further adjustments at this grid point will still be halted.

3 Data sets

The calibration procedure takes advantage of the improved quality of the modern, continental-scale Antarctic data sets, such as climatic forcing (Van Wessem et al., 2014) and topography (Fretwell et al., 2013). The forcing data serve as time-invariant boundary conditions for our equilibrium (steady-state) model simulations. It should be noted that the modern AIS is unlikely in a steady state, and that a transient simulation of, e.g., the entire last glacial cycle would provide a more realistic scenario for the calibration procedure. However, existing reconstructions of the Antarctic paleo-climate and past ice sheet configurations still contain large uncertainties, with in-situ data being sparse in space and time. Keeping this in mind, we believe that using modern data sets and assuming equilibrium conditions is a valuable first-order approximation to a more complex model calibration. Furthermore, such equilibrium set-up can serve as an initial guess for transient deglaciation simulations, which include time-dependent processes not considered here (Fyke et al., 2014).

Initial modern conditions for surface topography, ice shelf thickness, and bedrock elevations relative to the present-day sea level are derived from the BEDMAP2 data set (Fretwell et al., 2013). BEDMAP2 is a compilation of 24.8 million ice thickness data points obtained from a variety of sources including airborne and over-snow radar surveys, satellite altimetry, seismic sounding data, and satellite gravimetry (Fretwell et al., 2013). This compilation is complemented by surface elevation data from several digital elevation models to derive previously unknown bedrock features, and allows for a detailed modelling of the AIS. Figure 1 shows the bedrock topography data from BEDMAP2, together with the locations mentioned in the text. The main sources of uncertainty in the ice thickness and bedrock elevation maps are the errors in surface digital elevation models and ice thickness measurements, as well as the applied regridding, which produce overall uncertainties ranging from $59\,\mathrm{m}$ across areas with smooth landscapes to $1000\,\mathrm{m}$ in regions where only gravimetric data are available (e.g., south of Coats Land).

At the base of the thermal bedrock, geothermal heat flux is prescribed according to the map of Shapiro and Ritzwoller (2004). This map is derived from a global seismic model of the upper mantle and the crust that assuming a relation between seismic velocities and mantle temperatures and uses observations from regions with similar structures to infer heat-flow probability distributions where such observations are scarce or non-existent. In the inferred heat-flux map West Antarctica is characterised by an average heat flow that is nearly three times higher than in East Antarctica. Although the resulting map depends on the accuracy of available observations as well as on the choice of the seismic model and similarity functional, the inferred distributions are robust to internal parameter changes, especially for continental areas such as Antarctica (Shapiro and Ritzwoller, 2004). However, the method is not able to reproduce small-scale patterns caused by local variations in crustal heat production, which may exert an important control on the dynamics of the Antarctic ice streams.

Boundary conditions at the surface include Antarctic accumulation rates and near-surface air temperature data from the regional climate model RACMO2.3 (Van Wessem et al., 2014), averaged over the period of 1979 to 2010. RACMO is forced at its boundaries by the reanalysis data from ERA-Interim (Dee et al., 2013). In the interior of the domain the Antarctic climate conditions are modelled with a horizontal resolution of $27\,\mathrm{km}$ and 40 levels in the vertical direction. RACMO contains modules that are specifically developed for glaciated regions, including a multilayer snow model. The model compares well with 3234

in situ observations of surface mass balance used for its validation, displaying a correlation of $r^2 = 0.77$ and a particularly good fit for the dry East Antarctic plateau (Van Wessem et al., 2014).

A simple lapse-rate correction of $0.008\,^\circ\mathrm{C\,m^{-1}}$ is used to account for the discrepancies between the modelled and observed surface elevations. Surface melt is computed with a positive degree-day (PDD) scheme (Reeh, 1991; Calov and Greve, 2005)

5   using the factors $\alpha_{\mathrm{ice}} = 8\,\mathrm{mm\,d^{-1}\,^\circ C^{-1}}$ and $\alpha_{\mathrm{snow}} = 3\,\mathrm{mm\,d^{-1}\,^\circ C^{-1}}$ (ice equivalent) for ice and snow, respectively (Ritz et al., 2001), and a standard deviation of $\alpha_{\mathrm{std}} = 5\,^\circ\mathrm{C}$ for the statistical fluctuations in air temperature.

All input fields are projected onto a regular, rectangular, polar stereographic grid covering the entire Antarctic continent and the surrounding Southern Ocean, with a nominal horizontal resolution of $20\,\mathrm{km}$, corresponding to 301×301 grid points. Our choice of resolution has been motivated by the large number of experiments presented in Section 5, each spanning $400{,}000$

10   years, and the fact that Pollard and DeConto (2012b) have shown that the results remain essentially unchanged when the horizontal resolution is increased to $10\,\mathrm{km}$ (in a nested simulation), even in rapidly flowing sectors. However, our tests using an increasingly higher resolution have identified areas where the modelled ice flow is more sensitive to the resolution used. These mainly occur close to the ice sheet margins where the grounding zone of a glacier is often represented by only one grid cell at lower resolution. In such regions the use of a finer grid allows for a more detailed treatment of the topographically

15   constrained glacial flow. In the context of this comparison study, however, these limitations equally affect the performance of all hybrid schemes and do not impact our conclusions. In the vertical direction, ice columns consist of 91 layers (11 equidistant grid points for temperate ice and 81 grid points for "cold" ice densifying towards the base, sharing the grid point at their interface), mapped to a $[0, 1]$ interval using a σ transformation (Greve and Blatter, 2009).

4   Experimental setup

20   The modified version of SICOPOLIS described in Section 2 is applied to the entire AIS and the fringing ice shelves. The experiments performed during the calibration of the ice sheet-shelf system aim at quantifying the differences and similarities between the hybrid approaches. Default values for the parameters of the hybrid schemes are $r_{\mathrm{thr}} = 0.5$ (HS-1), and $v_{\mathrm{ref}} = 100\,\mathrm{m\,a^{-1}}$ (HS-2a and HS-2b). As mentioned in Section 2.2, the HS-3 does not include any free parameters. Additionally, and for comparison purposes, the same experiments are performed using a SIA-only scheme for grounded ice (henceforth SoS).

25   The bulk of the calibration consists of an iterative technique used to derive the distribution of sliding coefficients C_0 (Section 2.3) that exerts a dominant control on the resulting ice distribution and its fit to observations. The calibration procedure starts from a homogeneous guess value of $C_0 = 1\,\mathrm{m\,yr^{-1}\,Pa^{-1}}$, which is the lower limit of the considered range (see Section 2.3). The time step between adjustments and the scaling factor in Eq. (13) are set to $\Delta t_{\mathrm{inv}} = 50$ years and $H_{\mathrm{inv}} = 5000$ metres, respectively. We follow the method by Pollard and DeConto (2012b) and allow for a free evolution of both the ice sheet

30   and ice shelf thickness, but their interface (the grounding line) is kept at its present-day observed position. Free evolution is needed because the calibration of the sliding coefficients requires an evolving ice thickness that will be routinely compared to observations. The reasons for a constrained grounding line are twofold: This approach 1) enables a comparison with observations in coastal regions during the calibration and 2) prevents artificial transitions between grounded and floating areas caused

by equally artificial effects of unrealistic initial thermal regimes and C_0 distributions that evolve from initial guess values. Our tests show that such artefacts can produce feedbacks that are difficult or impossible to reverse. For the same reason, glacial isostatic adjustment is not included in the simulations, and ice shelf fronts are constrained to the observed locations. As mentioned above, the ice shelf thickness is allowed to evolve, but basal melt rates are adjusted at each time step in order to keep the modelled ice shelves as close as possible to observations. This ensures a consistent computation of mass fluxes across all flow regimes, and does not overlap with the calibration of sliding coefficients, since these are not applied in floating ice sectors.

In this study, the calibration procedure is divided into four steps. In the first three steps, a relaxation scheme is applied to the evolution of the modelled ice thickness, H. Here, the difference between the current solution of the ice thickness equations, H_{new}, and the solution from the previous time step, H_{old}, is scaled at every time step by a factor h_{rlx} ranging between 0 and 1, as follows:

$$H = H_{\mathrm{old}} + h_{\mathrm{rlx}}(H_{\mathrm{new}} - H_{\mathrm{old}}). \tag{14}$$

For a time-invariant forcing, our tests have shown that different values of h_{rlx} will result in very similar equilibrium states. The relaxation simply delays the time point at which this state is reached. However, such relaxation procedure allows for bigger time steps at which the topography evolution is computed, without affecting the internal temperature evolution. In this way, an equilibrium with the boundary conditions will be reached faster by the temperature field than by the topography. This effectively minimises transient effects in the closely associated ice thickness and velocity fields, especially at the beginning of the calibration runs when model parameters follow the initial guess values. More importantly, it allows us to simultaneously apply the iterative calibration technique described in Section 2.3, in contrast to approaches in which the topography is fixed (e.g., Sato and Greve, 2012). This ensures that the modelled ice thickness distribution has the closest possible match to the initial observed distribution throughout the initial calibration stages. The value of the relaxation factor is set to $h_{\mathrm{rlx}} = 0.001$ in the first stage of the calibration driven over a simulation time of $100,000$ years ($100\,\mathrm{kyr}$), using a time step of 5 years. The following second and third stages replicate the first one, but with larger scaling factors of $h_{\mathrm{rlx}} = 0.01$ and $h_{\mathrm{rlx}} = 0.1$, respectively, for an additional simulation time of $100\,\mathrm{kyr}$ in each stage. In the final stage, the relaxation is deactivated and a smaller time step of 1 year is applied over an additional simulation time of $100\,\mathrm{kyr}$, which provides a long enough time span to attain a dynamic equilibrium.

5   Results and discussion

In this chapter we present an ensemble of simulations of the AIS that aim to comprehensively evaluate different hybrid schemes combining the SIA and SStA. Our evaluation uses the degree of agreement between the modelled and observed ice sheet geometries and surface velocities as a measure of their performance, allowing for a point-by-point quantification of the model errors.

Keeping in mind that each hybrid scheme builds upon the SIA solution, partially or entirely replacing it at variable locations with the SStA solution, we have also included the results from the SIA-only scheme in our comparison. In particular, this en-

ables a qualitative separation of relative contributions of the SIA and SStA, providing a new insight into the internal differences between the schemes and their applicability to ice sheet areas with diverse dynamical characteristics.

By applying an automated model calibration against the observed ice thickness to each of the hybrid schemes we infer spatial distributions of poorly constrained sliding coefficients as a proxy for mechanical conditions at the ice-bedrock interface and assess their sensitivity to the choice of a particular hybrid scheme. In addition, the influence of variations in the parameters controlling relative contributions of SIA and SStA velocities in each scheme is assessed for a wide range of parameter values.

5.1 Comparison of equilibrium states

As described in Section 3, our experiments use the BEDMAP2 observational data set (Fretwell et al., 2013) as an initial ice sheet configuration, running the model forward in time under a relaxation scheme (Eq. (14)) until the temperature distribution within the ice sheet reaches an equilibrium state. After this initialisation, the relaxation is deactivated and the model runs until a full thermal and dynamic equilibria are attained. The equilibrium is defined as the point in time in which the variations of the total grounded ice volume over a prolonged time ($> 10,000$ years) are smaller than 0.01%.

Starting from the initialised states, the time required to reach an equilibrium varies from scheme to scheme (Fig. 2). The HS-1, HS-2a and HS-3 attain a virtually invariable state after only $20\,\text{kyr}$, as opposed to the SoS and HS-2b that display significant oscillations in the total volume around a mean equilibrium value even after a period of $100\,\text{kyr}$, suggesting unstable equilibrium states. Compared to the SoS, the computation of SStA velocities in the grounded ice sectors requires an extra computational effort for each iteration in the numerical solvers, with the computing time increasing by a factor of ~ 4 for the applied hybrid schemes. The characteristic computing time of the HS-1 is somewhat shorter than that of the other hybrid schemes (but still longer than for the SoS) due to the prognostic identification of ice streams that obviates the need for the computation of SStA velocities over the entire ice sheet. However, we have observed that the iterative solvers in the model require a substantially smaller number of iterations when the hybrid schemes are used, making them numerically more stable compared to the SoS.

The calibration procedure applied to all schemes yields total grounded ice volumes which are in a close agreement with the reference value of $\sim 2.56 \times 10^7\,\text{km}^3$ from the observational data (Fretwell et al., 2013), with the maximum deviation being below 1.5%. Individual values of the modelled ice volumes and their respective deviations for most of our experiments are summarised in Table 3. The best fit to observations is obtained using the HS-2b, which corresponds to an overestimation of the total grounded ice volume by 0.1%. The other schemes produce relatively larger misfits, with the HS-2a simulation yielding the greatest deviation of 1.46% arising from an overestimation of the total ice volume. The smallest ice sheet is produced by the SoS that underestimates the total grounded ice volume by 0.72%.

The inset of Figure 2 shows the total ice thickness errors for all schemes. The errors are computed as an average of the absolute misfit values in all grounded grid points and remain below 50 metres for all hybrid schemes (Table 3). Among these, the largest error of $46.6\,\text{m}$ is produced by the HS-2a, while the smallest error of $40.0\,\text{m}$ is obtained using the HS-2b. For the former, an 83% of the misfit is due to overestimations of ice thickness, while the latter shows nearly even contributions from under- and overestimations. This is in accordance with the results shown in Fig. 2, where the misfits obtained from the two schemes using the SStA as a sliding law are dominated by an excessive ice thickness. The hybrid schemes that include a

11

Weertman basal sliding component during the computation of SIA velocities are not necessarily dominated by underestimations of the ice thickness. For example, less than a quarter of the misfit produced by the HS-1 can be attributed to areas with an ice thickness deficit. For comparison, the total error produced by the SoS is $52.4\,\mathrm{m}$, with $65\,\%$ of the misfit coming from an underestimation of the ice thickness.

5   The results depicted in Fig. 2 reflect only a generalised, time-dependent information that can be inferred from each run. In order to have a more detailed overview of the results, Fig. 3 (left column) shows the corresponding spatial maps of ice thickness errors. All schemes provide a reasonably good fit to the observational data in the continental interior, with larger discrepancies mainly occurring at the ice sheet margin. It can be readily observed that the modelled ice sheet is too thick over mountainous regions and across the region between the Shakleton Range and the Pensacola Mountains (south of Coats Land), which has

10 some of the largest uncertainties in the topographic data (Fretwell et al., 2013). These are common features for all schemes, regardless of the approach chosen for the sliding component, and could originate from insufficient geothermal flux, too high precipitation rates, and/or an unrealistic smoothness of the bedrock in this area that prevents the formation of topographically driven ice streams. The distribution of zones where the modelled basal temperatures are far below the pressure-melting point are depicted as white-coloured areas in Fig. 3 (right column). In general, these areas coincide with the locations where the

15 largest ice thickness errors occur. Areas where the ice thickness is underestimated are mainly located at and around the ice margins. In many of these areas, however, sliding is identified and the calibration of C_0 reaches its lower limit, implying that the surface mass input is insufficient to minimise the misfit. The SoS produces too thin ice along most of the ice sheet margin, which explains the high percentage of the ice thickness errors arising from underestimations (inset of Fig. 2). In contrast, the hybrid schemes produce error patterns that differ only slightly between each other. The only exception is the HS-2b, which is

20 closest to the SoS, albeit exhibiting smaller underestimations at the ice sheet margins.

Furthermore, we evaluate the performance of the calibration procedure over the areas where it is applied directly. For this purpose we have calculated averaged absolute errors in the ice thickness across regions where basal sliding operates (inset of Fig. 2). It should be kept in mind that the calibration procedure also affects the ice masses located in the immediate proximity to sliding areas through surface elevation gradients and/or stress transmission. Mean ice thickness errors over the sliding areas

25 are smaller than those estimated for the whole ice sheet, although the degree of relative improvement varies from scheme to scheme (Table 3). For example, the misfit produced by the SoS decreases by only $\sim 20\,\%$ if calculated over the sliding ice sheet sectors, while more than $60\,\%$ of the errors resulting from the HS-2a occur over the areas where no sliding is recovered and C_0 is not calibrated. For all hybrid schemes, the percentage of the errors associated with an underestimation of the ice thickness increases substantially when their performance is assessed across areas of the retrieved basal sliding, supporting the

30 observation that the modelled ice is excessively thick in regions where C_0 is not calibrated.

The inferred distributions of C_0 are shown in Fig. 3 (right column). In general, the areas where the calibration is performed are similar for all schemes (Table 3), although there is a significant spread in the retrieved values. The HS-1 scheme predicts a minimum corresponding to $45\,\%$ of the total grounded ice area, contrasted by higher percentages for the other schemes ranging between $54\,\%$ and $62\,\%$. The upper limit of $10^5\,\mathrm{m\,yr^{-1}\,Pa^{-1}}$ for the sliding coefficient is reached by all schemes across up

35 to $18\,\%$ of the ice-covered land, with their highest concentration occurring in the Siple Coast region, where ice streams flow

rapidly over a smooth and deformable bed provided by strong lubrication from water saturated subglacial sediments (e.g., Blankenship et al., 1986; Alley et al., 1987; Kamb, 2001). The upper limit is also reached in Coats, MacRobertson, and Ellsworth Lands (Fig. 1). The C_0 values placed at the lower limit of $1\,\mathrm{m\,yr^{-1}\,Pa^{-1}}$ are also present in the estimates from all runs, with a relatively higher coverage ranging between $\sim 10\,\%$ and $\sim 32\,\%$ of the ice sheet area. The SoS and HS-2b infer this value throughout most of the West AIS and over vast parts of the East AIS, particularly at the ice margins. Its incidence is relatively lower for the other hybrid schemes, especially for the HS-1 and HS-2a. Quantitatively, the good agreement between the inferred coefficient distributions is mainly restricted to the areas where C_0 reaches an upper or lower limit. In other areas the values generally vary across orders of magnitude.

Direct comparison of our results to those from Pollard and DeConto (2012b) is hindered by the differences in the sliding laws and hybrid schemes. Nevertheless, we have found a good qualitative agreement in the inferred distributions of C_0, with similar patterns of high vs. low values. For instance, both studies identify the Siple Coast and Coats Land regions as areas where the ice streaming flow is driven by basal conditions favourable for sliding. A similar agreement is found for the Thwaites and Pine Island Glacier areas, as well as in the MacRobertson Land (Fig. 1). Low values of C_0 are predicted in the continental interior of West Antarctica and over most of the East AIS.

5.2 Analysis of the SIA and SStA contributions

As described in Section 2.2, different hybrid approaches are not expected to produce exactly the same equilibrated ice velocity fields. We demonstrate this in Fig. 4, where the modelled steady-state surface velocities derived from the SoS and the hybrid schemes are compared to a continental-scale observational data set (Rignot et al., 2011). This high-resolution ($900\,\mathrm{m}$) data set contains many small-scale features that are unresolved by the model due to its lower resolution. Nevertheless, all schemes are able to reproduce the observed range of ice flow regimes, distinguishing between the ice sheet areas with low velocities near the ice divides and fast flowing ice streams reaching the ice sheet margins. In the transition zones between the continental interior and the ice sheet margins, all schemes reproduce to some extent the fast flowing ice streams identified by observations. However, in contrast with the results from the hybrid schemes, surface ice velocities derived from the SoS are characterised with noise-like patterns, which are especially visible in the areas of rapid ice flow. We attribute these artefacts to a combination of lacking stress transmission in the SIA, which allows for steep gradients in the modelled velocities, and the calibration procedure, which can potentially amplify these gradients through local adjustments of C_0.

Although the overall character of the observed surface ice velocities is qualitatively well reproduced by all the hybrid schemes, modelled ice flow is clearly too fast at and around several ice stream locations, such as for example at the Siple Coast. Furthermore, modelled surface velocities are generally overestimated close to the grounding lines of most outlet glaciers, in particular due to the resolution-related limitations discussed in Section 3. These overestimations are particularly large in the SoS simulation, even though the discrepancies between the modelled and observed ice thickness are small (Fig. 3, left column). Overestimations of the ice flow velocity by the hybrid schemes at the grounding zone are smaller than those derived from the SoS simulation, but they cover a wider area and reach further upstream. Similar observations have been made by Pollard and DeConto (2012b) using a different hybrid ice sheet-shelf model and a different set of topographic ice observations and external

13

forcing. They proposed that the overestimation of surface velocities in these areas may be caused by a coarse model resolution, exaggerated snowfall rates or an excessive internal deformation compared to sliding near the ice margins.

In order to quantitatively evaluate the model fit to observations, we have calculated point-by-point ratios between the modelled and observed surface velocities (Fig. 4, right column). As mentioned above, the modelled velocities are in a good agreement with observations in the continental interior characterised by slow ice motion. In contrast, at the margins the ice velocity simulated by the hybrid schemes sometimes reaches values that are several hundred times higher than in the observational data set. Yet, this mostly happens across areas where our model generates a non-existent fast flow. One of the best examples of such model artefacts is the former Ice Stream C in the Siple Coast, which is predicted by the model but has been stagnant for ~ 150 years in reality (Hulbe and Fahnestock, 2007; Engelhardt and Kamb, 2013). In some cases, the locations of the modelled ice streams are shifted relative to the observed ones, thereby generating adjacent zones of under- and overestimations of the surface velocity. In other cases, the modelled rapid ice flow follows a different route compared to observations, sometimes merging with neighbouring ice streams. These shifts may originate from local errors in the bedrock topography data accentuated by its projection onto the coarse horizontal grid we use here.

The mean errors in the absolute surface ice velocity fall within the range of ~ 16 to $\sim 55 \, \mathrm{m \, yr^{-1}}$ (Table 3), with the HS-2b and SoS producing the minimum and maximum misfits, respectively, analogously to the results for the mean ice thickness errors discussed in Section 5.1. In the SoS simulation a general underestimation of the ice thickness near the ice sheet margins coincides with areas where sliding coefficients tend to reach the lower limit prescribed for the calibration. In turn, the use of such low values triggers a slowdown of the ice flow in the transition zone.

Although the results of the HS-2b simulation presented in Section 5.1 are in many aspects similar to those from the SoS, their abilities to reproduce the observed ice flow patterns are very different. The HS-2b uses the basal velocity from Equation (2) (utilised by the SoS) to compute the weights of relative contributions of the shallow approximations, thereby adding the SStA contribution where sliding velocities from Equation (2) are high. In these rapidly flowing sectors, we attribute a better performance of the HS-2b compared to the SoS to the inclusion of the stress transmission by the SStA. It also fosters a SStA-dominated modelled ice flow in the surroundings of ice streams, particularly upstream, thereby improving the overall fit to observations. Compared to the HS-2b, the HS-2a has a similar performance with the second smallest mean velocity error of $17.6 \, \mathrm{m \, yr^{-1}}$, followed by higher mean errors of 23.2 and $29.7 \, \mathrm{m \, yr^{-1}}$ produced by the HS-3 and HS-1, respectively. Thus, the use of hybrid schemes allows for a two- to threefold reduction in the surface velocity misfit.

As described in Section 2.2, the hybrid schemes used in this study mainly differ in how the SStA weight, w, is computed. This does not only control relative contributions from the shallow approximations, but also the locations where such combinations are implemented. In order to provide a deeper insight into how each hybrid scheme combines the SIA and SStA velocities, Fig. 5 illustrates the equilibrated distributions of w. It can be immediately observed that the distributions of w produced by the hybrid schemes are very different, both in spatial coverage and inferred values. For example, the percentage of the grounded ice area where the modelled ice flow is dominated by the SStA ($w > .75$) is $\sim 30\%$ for the HS-1, but only 2.4% and 8.5% for the HS-2a and HS-2b, respectively (Table 3). Furthermore, the transition between the SIA- and SStA-dominated ice sheet sectors appears patchy in the HS-1, whereas it is smooth and collocated with the present-day ice streams in the HS-2a, thereby

14

resembling the observed surface velocities. Such transitions are sharp in the HS-2b, implying a simple differentiation between fast and slow ice flow areas. This particular scheme exhibits a SIA-dominated ice flow regime ($w < .25$) over 82% of the grounded ice area that may explain a high degree of similarity with the SoS, especially keeping in mind that it also includes basal sliding in the SIA. In contrast, the HS-2a sets basal ice velocities to zero in the computation of the SIA velocities, still
5   producing a similar percentage of SIA-dominated area. In these sectors, differences in the ice thickness derived from the HS-2a and HS-2b (Fig. 3, left column) can be attributed to a presence or absence of basal sliding in the computation of the SIA solution that may prevent overestimations, e.g., in interior East Antarctica and cause underestimations in other sectors, such as, for example, the surroundings of Dronning Maud and MacRobertson Lands.

It is clear that the degree of agreement between the modelled and observed ice flow is sensitive to the choice of a particular
10   hybrid scheme and the way it measures relative SIA and SStA contributions. This can be best visualised using scatter plots of the modelled vs. observed surface velocities for each scheme, color-coded for the values of the corresponding w distributions (Fig. 5). In general underestimations of the modelled velocities occur across slowly flowing ($< 10\,\mathrm{m\,yr}^{-1}$) areas, which are dominated by the SIA, while the areas of fast flow dominated by the SStA are responsible for most of the overestimation. At a first glance it may seem that overestimations are caused by an excessive contribution of the SStA, but a comparison with
15   the SoS scatter plot shows that this is not necessarily the case. The largest overestimations occur in the SoS simulation, with surface velocities reaching the upper permitted limit clustered in the upper part of the scatter plot (Fig. 5). As mentioned above, the SoS and HS-2b share many similarities, and a comparison between their respective scatter plots shows that the use of the hybrid scheme reduces both underestimations in the lower velocity range and overestimations in the fast flowing areas.

5.3 Intercomparison of the inferred basal sliding coefficients

20   Although the performance of the hybrid schemes is quantitatively similar when evaluated against observations, the inferred values of basal sliding coefficients (C_0) vary by orders of magnitude in many regions of Antarctica (Fig. 3, right column). It is important to keep in mind that these large differences between the inferred distributions of C_0 arise solely from the differences in the hybrid schemes, namely in their ways to combine the shallow approximations, since all other model components are exactly the same for all experiments.

25   To demonstrate the effects of prescribing a distribution of C_0 derived from one ice model (using a specific hybrid scheme) into a different ice model (using a different hybrid scheme or other level of approximation), we performed an additional set of experiments, which utilise the equilibrium states described in Section 5.1 as initial conditions. In these experiments, we exchange the distributions of C_0 inferred from the HS-2a and HS-2b, and then run the model over a period of $100\,\mathrm{kyr}$. As described in Section 2.2, these schemes slightly differ in how they compute the SStA contribution, mainly due to different
30   techniques used to account for basal sliding. Although both schemes identify similar locations of rapidly flowing ice (Section 5.2), their inferred distributions of basal sliding coefficients contain the highest variability among all hybrid schemes implemented in this study. At the end of these additional 100kyr runs, the mean misfit between the modelled and observed ice thickness is above $200\,\mathrm{m}$ for the HS-2a, and almost $350\,\mathrm{m}$ for the HS-2b (not shown). This represents increments of $\sim 330\%$ and $\sim 775\%$ in the deviations from observations, respectively.

To further exemplify the significance of the associated uncertainties in the retrieved basal sliding parameters, we prescribe the median of the inferred distributions of C_0, computed from all hybrid schemes, at the base of the modelled ice sheet. Our choice of a median over an average is motivated by our initial tests in which generally larger values of C_0 inferred from the HS-2a tend to produce an average biased towards this particular distribution. Figure 6 shows differences between the modelled and observed ice thickness at the end of these additional runs. Comparison with Fig. 3 reveals a general degradation of the fit between the model and observations. The HS-2a and HS-2b exhibit the largest sensitivity to the change in the basal sliding parameters, with a significant amplification of over- and underestimations of the ice thickness occurring across most of the ice sheet, respectively. For the HS-2a, an absolute ice thickness misfit increases by $\sim 180\,\%$ to a mean value of $131\,\mathrm{m}$ (Table 4), of which $95\,\%$ is due to overestimations. The misfit increases by $\sim 355\,\%$ to a mean value of $182\,\mathrm{m}$ in the HS-2b simulation, with underestimations of the ice thickness accounting for $88\,\%$ of the total difference. This degree of degradation is less pronounced when the HS-1 and HS-3 are used. The misfit for the HS-1 increases by only $\sim 50\,\%$, displaying a mixture of areas where the modelled ice thickness is either too large or too small relative to observations. The HS-3 experiment shows an intermediate degree of sensitivity to a change in basal parameters and exhibits similar misfit patterns as the HS-2b, albeit the magnitude of the underestimation is smaller and the mean absolute error in the ice thickness remains around $100\,\mathrm{m}$. These experiments show that care is needed when using quantifications of the basal conditions obtained from external sources (e.g., Joughin et al., 2009; Morlighem et al., 2010; Arthern and Gudmundsson, 2010; Pralong and Gudmundsson, 2011; Pollard and DeConto, 2012b; Arthern et al., 2015) as input in ice flow models. This concerns not only differences between the basal sliding approaches implemented in each model, but also models using the same sliding law as part of different hybrid schemes, as shown in this study.

5.4 Exploration of the hybrid parameter space

The disparity in the results from the hybrid schemes presented in the previous sections showcases the impacts of slight changes in the model representation of fast flowing zones of ice sheets, even though it corresponds to a small fraction of the parameter space in ice-sheet models. In order to explore the sensitivity of the results to parameter variations within this parameter space, we perform an additional series of experiments where we vary the somewhat arbitrary threshold and reference quantities used by some of the hybrid schemes.

As described in Section 2.2, the hybrid schemes HS-1, HS-2a, and HS-2b use the weight w to calculate relative contributions of the SIA and the SStA. For the HS-1, the computation of w involves a prescribed threshold for the slip ratio of grounded ice (Eq. (8)), that determines the locations where the hybrid velocity is calculated. The experiments discussed in the previous sections use a default value of $r_{\mathrm{thr}} = 0.5$, meaning that the SIA and SStA solutions are combined only in areas where the basal velocity is at least half of the surface velocity. In contrast, the HS-2a and HS-2b combine both shallow approximations everywhere, and w is computed using a prescribed reference velocity (Eqs. (10) and (11)), using a default value of $v_{\mathrm{ref}} = 100\,\mathrm{m\,yr}^{-1}$. If the SStA velocity in the HS-2a or the basal velocity in the HS-2b reaches this reference value, the schemes assign equal weights to the SIA and the SStA. In general, higher values of r_{thr} or v_{ref} imply less contribution of the SStA solution, and vice versa.

The effects of variations in these parameters on the evolution of the total grounded ice volume during the calibration procedure are demonstrated in Fig. 7. Here we test parameter values within a range that contains almost every possible scenario, either because values outside the range are non-physical (HS-1) or they exhibit no noticeable differences compared to the range limits (HS-2a and HS-2b).

5   For the HS-1, the upper limit for the slip ratio threshold, $r_{\text{thr}} = 1$, implies the use of a SIA-only solution, whereas the lower limit, $r_{\text{thr}} = 0$, implies that a combination of SIA and SStA solutions is applied everywhere, determined by the weight w computed using the slip ratio. Within the range of tested values, maximum deviations from the observed grounded ice volume are below 1.2%. Only the use of $r_{\text{thr}} = 1$ produces an ice sheet that is smaller than observed. An analysis of the mean absolute differences between the modelled and observed ice thickness (Fig. 8) reveals that the use of larger values of r_{thr} leads to a

10   larger misfit with observations, with the maximum error corresponding to $r_{\text{thr}} = 1$.

     In the HS-2a, all tested values produce an ice sheet that is larger than observed. This is caused by a general overestimation of the ice thickness in the continental interior. Here, the modelled ice flow is slow and dominated by the SIA. Thus, the weights w are small, implying a negligible contribution of the SStA, independently of the value of v_{ref}. Values smaller than $v_{\text{ref}} = 100 \, \text{m} \, \text{yr}^{-1}$ are found to produce total grounded ice volumes and mean ice thickness misfits that are close to the results

15   of the reference run ($v_{\text{ref}} = 100 \, \text{m} \, \text{yr}^{-1}$), although with a slight improvement in the model fit to observations. Values of $v_{\text{ref}} > 100 \, \text{m} \, \text{yr}^{-1}$ lead to a larger misfit, with the highest tested value of $v_{\text{ref}} = 1000 \, \text{m} \, \text{yr}^{-1}$ producing a grounded ice volume that displays a deviation of 2.3% from observations. The use of even larger values of v_{ref} asymptotically decreases the contribution of the SStA. It is important to note, however, that since the HS-2a does not include sliding in the SIA component, its solutions will never approach that of the SoS.

20   In contrast, the HS-2b activates the sliding law even in the regions where ice streams have not been identified by the scheme, as long as the conditions for sliding described in Section 2.1 are fulfilled. This may explain the similarity between the grounded ice volumes produced by the SoS and the HS-2b solutions, using a reference velocity set to the upper limit of $v_{\text{ref}} = 1000 \, \text{m} \, \text{yr}^{-1}$. This parameter value produces a deviation of -0.9% from the observed ice volume, which is similar (but with opposite signs) to the misfit obtained from the simulation using the lower limit of $v_{\text{ref}} = 5 \, \text{m} \, \text{yr}^{-1}$ tested in this study. For

25   values higher than the default value of $v_{\text{ref}} = 100 \, \text{m} \, \text{yr}^{-1}$, the simulations produce an ice sheet that is smaller than observed. The use of $v_{\text{ref}} = 60 \, \text{m} \, \text{yr}^{-1}$ leads to the best fit between the modelled and observed ice thickness among all schemes and parameter values, reaching $\sim 39 \, \text{m}$, with a small deviation from the observed grounded ice volume of 0.27%.

     However, it is important to keep in mind that the calibration procedure has a limited power to reduce the misfits between the model and observations in areas where no sliding occurs. This is why some deviations are expected from our numerical

30   experiments, in particular due to overestimations of the ice volume over mountainous regions where the calibration procedure does not operate. Therefore, a perfect fit to the observed ice volume obtained by any of the schemes would likely involve underestimations of ice thickness in other regions.

6 Summary and conclusions

We implemented and compared the performance of four hybrid schemes for the combination of the Shallow Ice and Shelfy Stream Approximations as part of the ice-sheet model SICOPOLIS. The use of shallow approximations enables continental-scale, long-term paleo-simulations of ice sheets, preventing the restrictive computational expenses of more complex models. Moreover, hybrid schemes overcome the limitations of simpler SIA-only models in regions characterised by rapid ice flow driven by basal sliding. The hybrid schemes in this study differ in the ways 1) relative contributions of the shallow approximations are computed, 2) areas where a combination of shallow approximations is applied are identified, and 3) the basal sliding is accounted for.

By adapting a simple iterative technique to infer the distribution of basal sliding parameters from observational topographic data sets (Pollard and DeConto, 2012b), we show that all the applied hybrid schemes produce dynamic equilibrium states that are in a good agreement with observations. For all the schemes, mean ice thickness misfits range from 40 to $52.4\,\mathrm{m}$, and the total grounded ice volume deviations do not exceed $2.5\,\%$ for a wide range of the associated parameter uncertainties. For optimal parameter choices, mean errors in the ice thickness remain below $40\,\mathrm{m}$. For comparison, present-day simulations in continental-scale ice-sheet models typically produce widespread errors in the ice thickness reaching hundreds of metres, with deviations in total grounded ice volume that can exceed $10\,\%$ (e.g., de Boer et al., 2015). A clear limitation of the calibration method presented here is the possibility of an error cancellation during the adjustment of C_0. At present there is no easy way to differentiate between the effects of an artificial compensation of, for example, the lacking model physics or errors in the input data sets through the model calibration and the model parameters that are representative of the actual conditions at the base of the ice sheet. This shortcoming will hopefully be overcome in the future once the necessary observational data sets become available and the use of more sophisticated ice models more feasible. The degree of applicability of the calibrated C_0 distributions for long-term paleoclimate simulations is currently unclear, and its quantification requires further modelling studies of the past ice sheet geometries. However, we believe that the retrieved variability in basal sliding coefficients is a good first-order approximation, and is certainly a better guess than commonly used spatially uniform values of C_0 over the entire domain. We believe that the method used to derive C_0 could be potentially extended to the use for other time periods or climate regimes (or even adapted for the transitions between largely dissimilar regimes), providing that the necessary topographic and climate data are available. Considering the recent efforts to reconstruct the past geometries of the AIS (Mackintosh et al., 2014; Bentley et al., 2014), this may well become possible in the upcoming decades.

We also computed a non-hybrid, SIA-only solution to allow for a qualitative separation of relative contributions of the SIA and the SStA. Such direct comparison is possible because all the hybrid schemes analysed in this study share identical model components. Although the calibration procedure applied to the SIA-only scheme also produces a reasonable fit to observations, the misfits are overall larger than those from the hybrid schemes. Moreover, the modelled surface velocities exhibit noise-like patterns at and around locations where rapid ice flow exists. Since these patterns are not present in the hybrid solutions, we attribute their occurrence to the lack of stress transmission in the SIA flow.

We find that individual weights assigned to the SIA and SStA solutions vary significantly from scheme to scheme. For the schemes in which the SIA/SStA weights are computed, the modelled ice flow is dominated by the SIA over $\sim 50\%$ to $\sim 80\%$ of the AIS, with the predominant contribution of the SStA generally limited to the locations and surroundings of the observed ice streams.

5   Comparison of our results with an independent, observational data set of surface ice velocities reveals a reasonable agreement in the continental interior of Antarctica. However, the modelled ice flow appears to be too fast at the ice sheet margins in all hybrid schemes, especially close to the observed ice stream locations. These misfits between the modelled and observed surface velocities are contrasted by relatively small differences between the modelled and observed ice thickness in these areas. Such misfits may originate from other factors such as, e.g., uncertainties in the observational ice thickness and surface velocity data

10   sets (Fretwell et al., 2013; Rignot et al., 2011, see Section 3), the limited model resolution, the assumption of an ice sheet equilibrium with present-day climate conditions, and/or errors in the model-based geothermal heat flux and/or climatic forcing (Pollard and DeConto, 2012b).

     Although all hybrid schemes produce a comparably good fit to the observational topographic data set, the inferred values of basal sliding coefficients largely differ between different schemes. Therefore, the discrepancies in the representation of the ice

15   dynamics by different hybrid schemes are compensated through adjustments of this poorly constrained parameter. Nevertheless, the schemes mostly agree on the areas where the ice base is at or close to the local pressure-melting point. Furthermore, there is a qualitative agreement in the patterns of low vs. high values of C_0 obtained from each calibration run. Although any attempt to quantify a local potential for sliding would require new and improved observational data sets describing basal conditions, the robustness of the inferred patterns provides an initial guess for their real distribution.

20   Furthermore, we have evaluated the performance of each hybrid scheme based on both its fit to observations and its numerical stability for a range of model set-ups and have found that at higher grid resolutions the HS-1 and HS-2b become numerically unstable, owing to large gradients in basal sliding coefficients arising from the use of basal velocities as boundary conditions for the SIA solutions in conjunction with the calibration of sliding coefficients. In addition, variations in the free parameter r_{thr} of the HS-1 generate artefacts in the resulting basal temperature field. The HS-2a and HS-3, which utilise the SStA as

25   a sliding law are numerically more stable to variations in model parameters and changes in grid resolution. The drawback of the HS-2a is its limited ability to influence the fit between the modelled and observed ice thickness in ice sheet sectors where the SStA velocities are low (<100 m/yr), which is the case for large parts of the Antarctic interior. This limitation is independent of the choice of the parameter v_{ref} (Figure 7). The HS-3 overcomes this limitation by accounting for the SStA contribution everywhere. However, the HS-3 also utilises the SIA velocities over the entire ice sheet, which in certain areas are

30   excessively high, such as the steep ice sheet margins (Figure 4), leading to an increased root-mean-square error compared to HS-2a (Figure 5). To improve the performance of both schemes, our future work will reconcile the drawbacks of HS-2a with the advantages of HS-3, providing a very stable and flexible hybrid scheme.

     Finally, we assessed the effects of differences between the calibrated parameter distributions derived from the four hybrid schemes by performing additional experiments in which the inferred distributions of C_0 were either averaged or exchanged

35   between the schemes. Our results show that the parameter distributions are not exchangeable between the schemes, since

this leads to a strong degradation of the model fit to observations. This suggests that results of a model calibration and/or initialisation cannot be straightforwardly transferred to a model that uses a different level of approximation of the Stokes equations. Given an increasing number of studies attempting to quantify the basal conditions under ice sheets through a variety of methods including ice flow models, our experiments show that one need to be careful when using these results as input data

5 sets in glaciological models.

References

Ahlkrona, J., Loetstedt, P., Kirchner, N. and Zwinger, T., 2016. Dynamically coupling the non-linear Stokes equations with the Shallow Ice Approximation in glaciology: Description and first applications of the ISCAL method. Journal of Computational Physics, 308, pp.1–19.

Alley, R.B., Blankenship, D.D., Bentley, C.R. and Rooney, S., 1987. Till beneath ice stream B: 3. Till deformation: evidence and implications. Journal of Geophysical Research: Solid Earth, 92(B9), pp.8921–8929.

Arthern, R.J. and Gudmundsson, G.H., 2010. Initialization of ice-sheet forecasts viewed as an inverse Robin problem. Journal of Glaciology, 56(197), pp.527–533.

Arthern, R.J., Hindmarsh, R.C. and Williams, C.R., 2015. Flow speed within the Antarctic ice sheet and its controls inferred from satellite observations. Journal of Geophysical Research: Earth Surface, 120(7), pp.1171–1188.

Bentley, M. J., Cofaigh, C. O., Anderson, J. B., Conway, H., Davies, B., Graham, A. G., ... and Mackintosh, A. (2014). A community-based geological reconstruction of Antarctic Ice Sheet deglaciation since the Last Glacial Maximum. Quaternary Science Reviews, 100, 1-9.

Blankenship, D.D., Bentley, C.R., Rooney, S.T. and Alley, R.B., 1986. Seismic measurements reveal a saturated porous layer beneath an active Antarctic ice stream.

Blatter, H. (1995). Velocity and stress fields in grounded glaciers: a simple algorithm for including deviatoric stress gradients. Journal of Glaciology, 41(138), 333-344.

de Boer, B., Dolan, A.M., Bernales, J., Gasson, E., Golledge, N.R., Sutter, J., Huybrechts, P., Lohmann, G., Rogozhina, I., Abe-Ouchi, A. and Saito, F., 2015. Simulating the Antarctic ice sheet in the late-Pliocene warm period: PLISMIP-ANT, an ice-sheet model intercomparison project. The Cryosphere, 9, pp.881–903.

Bueler, E. and Brown, J., 2009. Shallow shelf approximation as a "sliding law" in a thermomechanically coupled ice sheet model. Journal of Geophysical Research: Earth Surface, 114(F3).

Calov, R. and Greve, R., 2005. A semi-analytical solution for the positive degree-day model with stochastic temperature variations. Journal of Glaciology, 51(172), pp.173–175.

Cuffey, K.M. and Paterson, W.S.B., 2010. The physics of glaciers. Academic Press.

Dee, D. P., Uppala, S. M., Simmons, A. J., Berrisford, P., Poli, P., Kobayashi, S., ... and Bechtold, P. (2011). The ERA-Interim reanalysis: Configuration and performance of the data assimilation system. Quarterly Journal of the royal meteorological society, 137(656), 553-597.

Dunse, T., Greve, R., Schuler, T.V. and Hagen, J.O., 2011. Permanent fast flow versus cyclic surge behavior: numerical simulations of the Austfonna ice cap, Svalbard. Journal of Glaciology, 57(202), pp.247–259.

Engelhardt, H. and Kamb, B., 2013. Kamb Ice Stream flow history and surge potential. Annals of Glaciology, 54(63), pp.287–298.

Gagliardini, O., Zwinger, T., Gillet-Chaulet, F., Durand, G., Favier, L., De Fleurian, B., Greve, R., Malinen, M., Martin, C., Raback, P. and Ruokolainen, J., 2013. Capabilities and performance of Elmer/Ice, a new-generation ice sheet model. Geoscientific Model Development, 6(4), pp.1299-1318.

Feldmann, J., Albrecht, T., Khroulev, C., Pattyn, F. and Levermann, A., 2014. Resolution-dependent performance of grounding line motion in a shallow model compared with a full-Stokes model according to the MISMIP3d intercomparison. Journal of Glaciology, 60(220), pp.353–360.

Fretwell, P., Pritchard, H.D., Vaughan, D.G., Bamber, J.L., Barrand, N.E., Bell, R., Bianchi, C., Bingham, R.G., Blankenship, D.D., Casassa, G. and Catania, G., 2013. Bedmap2: improved ice bed, surface and thickness datasets for Antarctica. The Cryosphere, 7(1).

Fyke, J. G., Sacks, W. J., and Lipscomb, W. H. (2014). A technique for generating consistent ice sheet initial conditions for coupled ice sheet/climate models. Geoscientific Model Development, 7(3), 1183-1195.

Glen, J.W., 1955, March. The creep of polycrystalline ice. In Proceedings of the Royal Society of London A: Mathematical, Physical and Engineering Sciences (Vol. 228, No. 1175, pp. 519–538). The Royal Society.

Greve, R., 1997. A continuum-mechanical formulation for shallow polythermal ice sheets. Philosophical Transactions of the Royal Society of London A: Mathematical, Physical and Engineering Sciences, 355(1726), pp.921–974.

Greve, R. and Blatter, H., 2009. Dynamics of ice sheets and glaciers. Springer Science and Business Media.

Hindmarsh, R.C.A., 2004. A numerical comparison of approximations to the Stokes equations used in ice sheet and glacier modeling. Journal of Geophysical Research: Earth Surface, 109(F1).

Hulbe, C. and Fahnestock, M., 2007. Century-scale discharge stagnation and reactivation of the Ross ice streams, West Antarctica. Journal of Geophysical Research: Earth Surface, 112(F3).

Hutter, K., 1983. Theoretical glaciology; material science of ice and the mechanics of glaciers and ice sheets, D. D. Reidel Publishing Co./Tokyo, Terra Scientific Publishing Co.

Joughin, I., Tulaczyk, S., Bamber, J.L., Blankenship, D., Holt, J.W., Scambos, T. and Vaughan, D.G., 2009. Basal conditions for Pine Island and Thwaites Glaciers, West Antarctica, determined using satellite and airborne data. Journal of Glaciology, 55(190), pp.245–257.

Kamb, B., 2001. Basal zone of the West Antarctic ice streams and its role in lubrication of their rapid motion. The West Antarctic ice sheet: behavior and environment, pp.157–199.

Lenaerts, J.T.M., Den Broeke, M.R., Berg, W.J., Meijgaard, E.V. and Kuipers Munneke, P., 2012. A new, high-resolution surface mass balance map of Antarctica (1979–2010) based on regional atmospheric climate modeling. Geophysical Research Letters, 39(4).

Lliboutry, L. and Duval, P., 1985, December. Various isotropic and anisotropic ices found in glaciers and polar ice caps and their corresponding rheologies: Ann Geophys V3, N2, March-April 1985, P207–224. In International Journal of Rock Mechanics and Mining Sciences and Geomechanics Abstracts (Vol. 22, No. 6, p. 198). Pergamon.

Ma, Y., Gagliardini, O., Ritz, C., Gillet-Chaulet, F., Durand, G. and Montagnat, M., 2010. Enhancement factors for grounded ice and ice shelves inferred from an anisotropic ice-flow model. Journal of Glaciology, 56(199), pp.805–812.

Mackintosh, A. N., Verleyen, E., O'Brien, P. E., White, D. A., Jones, R. S., McKay, R., ... and Miura, H. (2014). Retreat history of the East Antarctic Ice Sheet since the last glacial maximum. Quaternary Science Reviews, 100, 10-30.

Morland, L.W., 1987. Unconfined ice-shelf flow. In Dynamics of the West Antarctic Ice Sheet (pp. 99–116). Springer Netherlands.

Morlighem, M., Rignot, E., Seroussi, H., Larour, E., Ben Dhia, H. and Aubry, D., 2010. Spatial patterns of basal drag inferred using control methods from a full-Stokes and simpler models for Pine Island Glacier, West Antarctica. Geophysical Research Letters, 37(14).

Pattyn, F. (2003). A new three-dimensional higher-order thermomechanical ice sheet model: Basic sensitivity, ice stream development, and ice flow across subglacial lakes. Journal of Geophysical Research: Solid Earth, 108(B8).

Pattyn, F., Schoof, C., Perichon, L., Hindmarsh, R.C.A., Bueler, E., Fleurian, B.D., Durand, G., Gagliardini, O., Gladstone, R., Goldberg, D. and Gudmundsson, G.H., 2012. Results of the marine ice sheet model intercomparison project, MISMIP. The Cryosphere, 6(3), pp.573–588.

Pattyn, F., Perichon, L., Durand, G., Favier, L., Gagliardini, O., Hindmarsh, R.C., Zwinger, T., Albrecht, T., Cornford, S., Docquier, D. and Fuerst, J.J., 2013. Grounding-line migration in plan-view marine ice-sheet models: results of the ice2sea MISMIP3d intercomparison. Journal of Glaciology, 59(215), pp.410–422.

Pollard, D. and DeConto, R.M., 2012a. Description of a hybrid ice sheet-shelf model, and application to Antarctica. Geoscientific Model Development, 5(5), pp.1273–1295.

Pollard, D. and DeConto, R.M., 2012b. A simple inverse method for the distribution of basal sliding coefficients under ice sheets, applied to Antarctica. The Cryosphere, 6(5), pp.953–971.

5   Pralong, M.R. and Gudmundsson, G.H., 2011. Bayesian estimation of basal conditions on Rutford Ice Stream, West Antarctica, from surface data. Journal of Glaciology, 57(202), pp.3

Reeh, N., 1991. Parameterization of Melt Rate and Surface Temperature in the Greenland Ice Sheet. Polarforschung, 59(3), pp.113–128.

Rignot, E., Mouginot, J. and Scheuchl, B., 2011. Ice flow of the Antarctic ice sheet. Science, 333(6048), pp.1427–1430.

Ritz, C., Rommelaere, V. and Dumas, C., 2001. Modeling the evolution of Antarctic ice sheet over the last 420,000 years: Implications for altitude changes in the Vostok region. Journal of Geophysical Research: Atmospheres, 106(D23), pp.31943–31964.

Sato, T. and Greve, R., 2012. Sensitivity experiments for the Antarctic ice sheet with varied sub-ice-shelf melting rates. Annals of Glaciology, 53(60), pp.221–228.

Schoof, C., 2007. Ice sheet grounding line dynamics: Steady states, stability, and hysteresis. Journal of Geophysical Research: Earth Surface, 112(F3).

15   Schoof, C. and Hindmarsh, R.C., 2010. Thin-film flows with wall slip: an asymptotic analysis of higher order glacier flow models. The Quarterly Journal of Mechanics and Applied Mathematics, p.hbp025.

Shapiro, N.M. and Ritzwoller, M.H., 2004. Inferring surface heat flux distributions guided by a global seismic model: particular application to Antarctica. Earth and Planetary Science Letters, 223(1), pp.213–224.

Van Wessem, J.M., Reijmer, C.H., Morlighem, M., Mouginot, J., Rignot, E., Medley, B., Joughin, I., Wouters, B., Depoorter, M.A., Bamber, J.L. and Lenaerts, J.T.M., 2014. Improved representation of East Antarctic surface mass balance in a regional atmospheric climate model. Journal of Glaciology, 60(222), pp.761–770.

Weertman, J., 1964. Glacier sliding (No. CRREL-RR-162). Cold regions research and engineering lab Hanover NH.

Winkelmann, R., Martin, M.A., Haseloff, M., Albrecht, T., Bueler, E., Khroulev, C. and Levermann, A., 2011. The Potsdam parallel ice sheet model (PISM-PIK)-Part 1: Model description. The Cryosphere, 5(3), pp.715–726.

Table 1. Symbols and values for the model parameters used in this study.

Symbol	Description	Units	Value
g	gravitational acceleration	$\mathrm{m\,s^{-2}}$	9.81
ρ	density of ice	$\mathrm{kg\,m^{-3}}$	910
T	absolute temperature of ice	K	
T_m	temperature below pressure-melting point	K	
κ	heat conductivity of ice	$\mathrm{W\,m^{-1}\,K^{-1}}$	$9.828\mathrm{e}^{-0.0057T}$
c	specific heat of ice	$\mathrm{J\,kg^{-1}\,K^{-1}}$	$146.3 + 7.253T$
L	latent heat of ice	$\mathrm{kJ\,kg^{-1}}$	335
β	Clausius–Clapeyron gradient	$\mathrm{K\,m^{-1}}$	8.7×10^{-4}
R	universal gas constant	$\mathrm{J\,mol^{-1}\,K^{-1}}$	8.314
κ_r	heat conductivity of the lithosphere	$\mathrm{W\,m^{-1}\,K^{-1}}$	3
$\alpha_\mathrm{ice}, \alpha_\mathrm{snow}$	PDD factors for ice and snow	$\mathrm{mm\,d^{-1}\,{}^\circ C^{-1}}$	8, 3
α_std	PDD standard deviation	$^\circ$C	5
σ_e	effective shear stress	Pa	
σ_0	residual stress	kPa	10
$E_\mathrm{SIA}, E_\mathrm{SSA}$	enhancement factor for the SIA and SSA		1, 0.5
A	ice rate factor	$\mathrm{s^{-1}\,Pa^{-3}}$	
n	Glen flow law exponent		3
$\boldsymbol{u}$	horizontal SIA velocity	$\mathrm{m\,s^{-1}}$	
$\boldsymbol{v}$	horizontal SStA velocity	$\mathrm{m\,s^{-1}}$	
$\boldsymbol{U}$	horizontal hybrid velocity	$\mathrm{m\,s^{-1}}$	
v_0	regularisation speed in SStA equations	$\mathrm{m\,yr^{-1}}$	0.01
$\boldsymbol{\tau}_\mathrm{b}$	basal shear stress	Pa	
N_b	effective basal pressure	Pa	
p, q	sliding law exponents		3, 2
γ	sub-melt-sliding parameter	K	3
C_0	calibrated basal sliding parameter	$\mathrm{m\,yr^{-1}\,Pa^{-1}}$	
Δt_inv	time step for inversion of C_0	yr	50
H_inv	scaling factor for calibration of C_0	m	5000
r	slip ratio of grounded ice		
r_thr	default threshold value of r		0.5
v_ref	default reference value of $|\boldsymbol{v}|$	$\mathrm{m\,yr^{-1}}$	100
w	weighting function in hybrid schemes		
h_rlx	scaling factor for relaxation procedure		0.001

Table 2. Summary of the hybrid schemes implemented in this study.

Scheme	Basal sliding	Reference in text	References
HS-1	Weertman-type sliding	Equation (8)	This study
HS-2a	SStA velocities	Equation (10)	Bueler and Brown (2009)
HS-2b	Weertman-type sliding	Equation (11)	This study
HS-3	SStA velocities	Equation (12)	Winkelmann et al. (2011)

Table 3. Summary of the results at the end of the calibration procedure for each of the applied hybrid schemes (HS, see Section 2.2), including: The total grounded ice volume, V_{grd} (km^3); a deviation from the total grounded ice volume, ΔV_{grd} (%); a mean absolute error in the ice thickness, $\bar{\Delta} H$ (m); a fraction of the total area where basal sliding occurs, A_{sld} (%); a mean ice thickness error only where basal sliding operates, $\bar{\Delta}_{\mathrm{sld}} H$ (m); a mean surface velocity error, $\bar{\Delta} v_{\mathrm{s}}$ (m yr^{-1}); a fraction of the grounded ice area dominated by the SIA (only where $w < 0.25$), $A_{\mathrm{SIA}.25}$ (%); and a fraction of the grounded ice area dominated by the SStA (only where $w > 0.75$), $A_{\mathrm{SStA}.75}$ (%).

Scheme	V_{grd} [km^3]	ΔV_{grd}	$\bar{\Delta} H$ [m]	A_{sld}	$\bar{\Delta}_{\mathrm{sld}} H$ [m]	$\bar{\Delta} v_{\mathrm{s}}$ [m yr^{-1}]	$A_{\mathrm{SIA}.25}$	$A_{\mathrm{SStA}.75}$
HS-1	2.58×10^7	+1.20 %	44.6	45.2 %	20.8	29.7	47.9 %	29.4 %
HS-2a	2.59×10^7	+1.46 %	46.6	54.2 %	17.6	17.6	82.0 %	2.4 %
HS-2b	2.56×10^7	+0.10 %	40.0	62.0 %	25.4	16.0	82.3 %	8.5 %
HS-3	2.57×10^7	+0.77 %	42.6	58.7 %	22.3	23.2	-	-
SoS	2.53×10^7	−0.72 %	52.4	58.9 %	41.3	55.3	100 %	0 %

Table 4. Summary of the results at the end of additional 100kyr runs with a prescribed median of the inferred basal sliding coefficients, including: The total grounded ice volume, V_{grd} (km^3); a deviation from total grounded ice volume, ΔV_{grd} (%); a mean absolute error in the ice thickness, $\bar{\Delta} H$ (m); and a mean ice thickness error only where basal sliding operates, $\bar{\Delta}_{\mathrm{sld}} H$ (m).

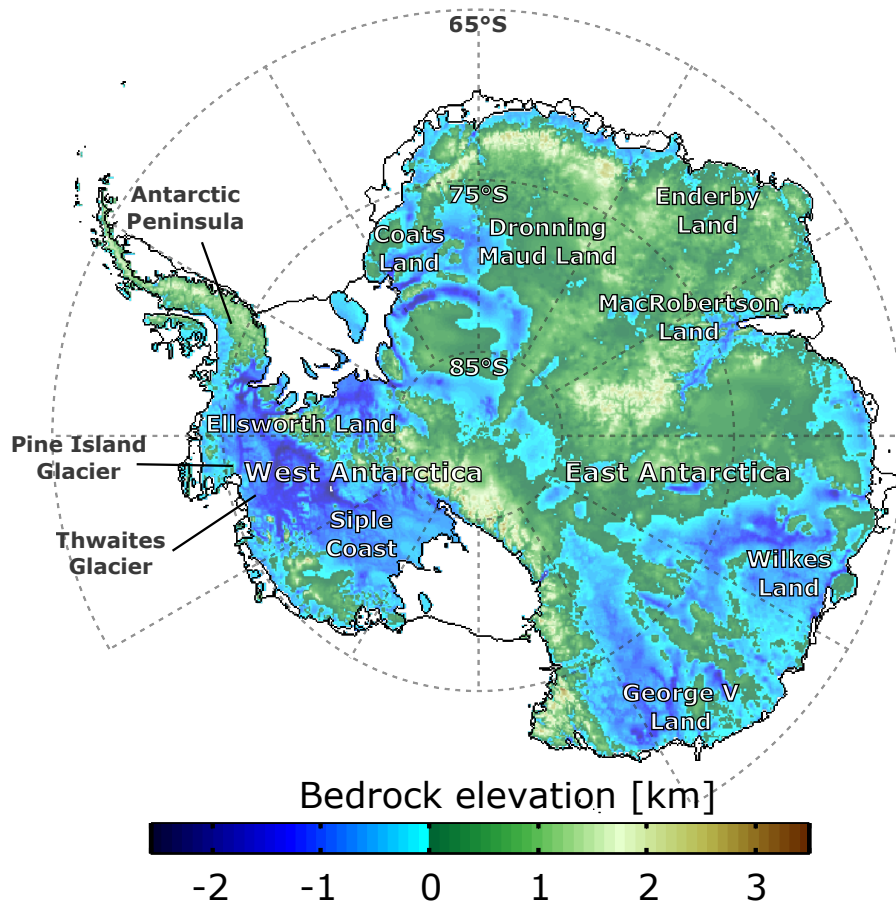| Scheme	V_{grd} [km^3]	ΔV_{grd}	$\bar{\Delta} H$ [m]	$\bar{\Delta}_{\mathrm{sld}} H$ [m]
HS-1	2.60×10^7	+1.95 %	67.0	45.2
HS-2a	2.70×10^7	+5.58 %	131.2	110.1
HS-2b	2.37×10^7	−7.41 %	182.1	167.6
HS-3	2.48×10^7	−2.97 %	105.0	104.8

Figure 1. A bedrock elevation map of Antarctica obtained from the BEDMAP2 data set (Fretwell et al., 2013), including the location of the sites mentioned in the text.
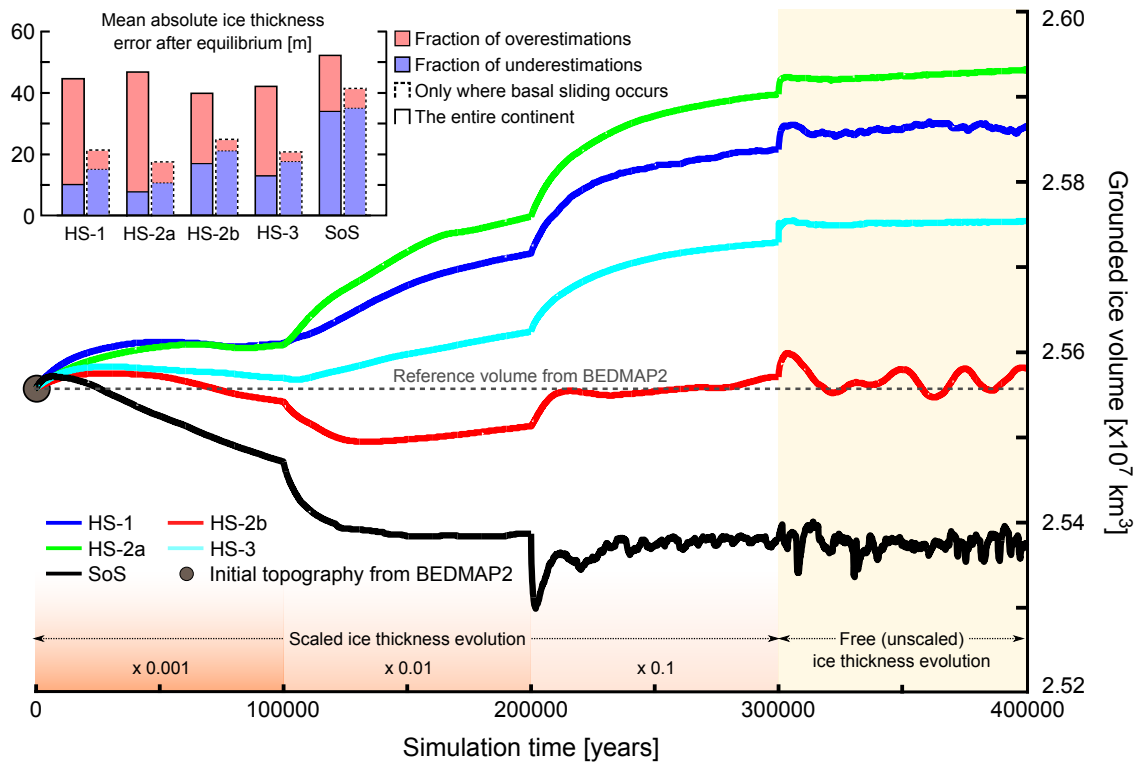
Figure 2. Overview of the calibration procedure. **Main:** The evolution of the total grounded ice volume during the calibration procedure, in $\times 10^7 \, \mathrm{km}^3$. **Inset:** Mean differences between the modelled and observed ice thickness at the end of the simulations, in m. For each scheme, a mean error is calculated for the entire ice sheet (left bars) and separately over the areas where basal sliding is identified (right bars). Fractions of the mean error arising from under- and overestimations are shown in blue and pink, respectively.

Ice thickness error   Basal sliding coefficient

HS-1

HS-2a

HS-2b

HS-3

SoS

[m]

-1000   0   1000

$[\mathrm{m\,yr^{-1}Pa^{-1}}]$

1   10^1 10^2 10^3 10^4 10^5
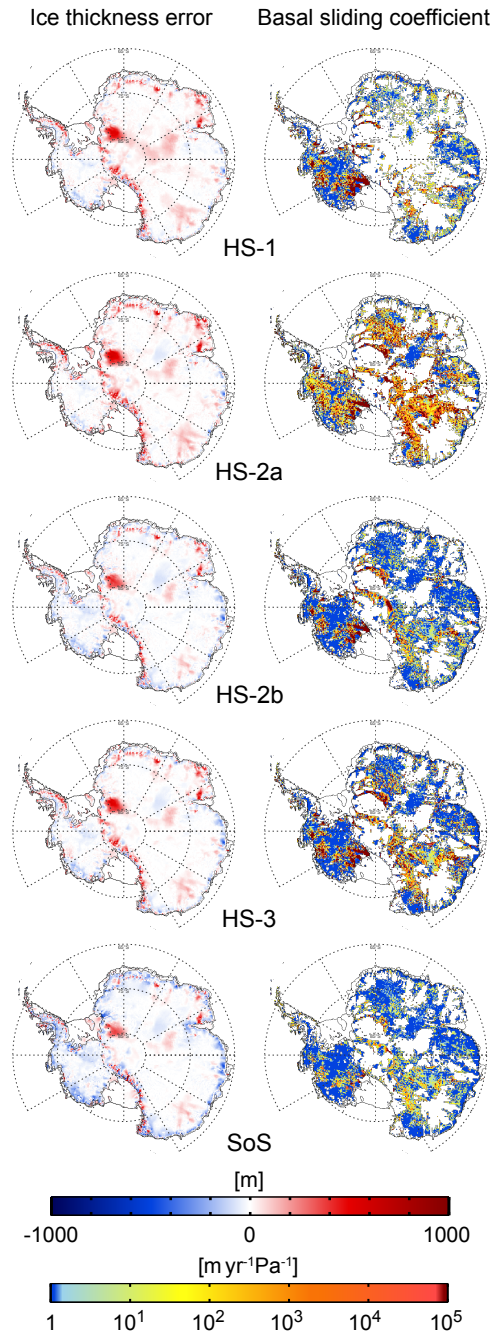
Figure 3. Comparison of the ice sheet states derived from different schemes at the end of equilibrium simulations. **Left column:** Differences between the modelled and observed ice thickness, in m. **Right column:** Inferred distributions of basal sliding coefficients, in $\mathrm{m\,yr^{-1}\,Pa^{-1}}$. Non-coloured areas mark the locations where basal sliding is not identified and the calibration procedure does not operate. Color-code saturates at the upper and lower limits allowed for the calibration procedure.
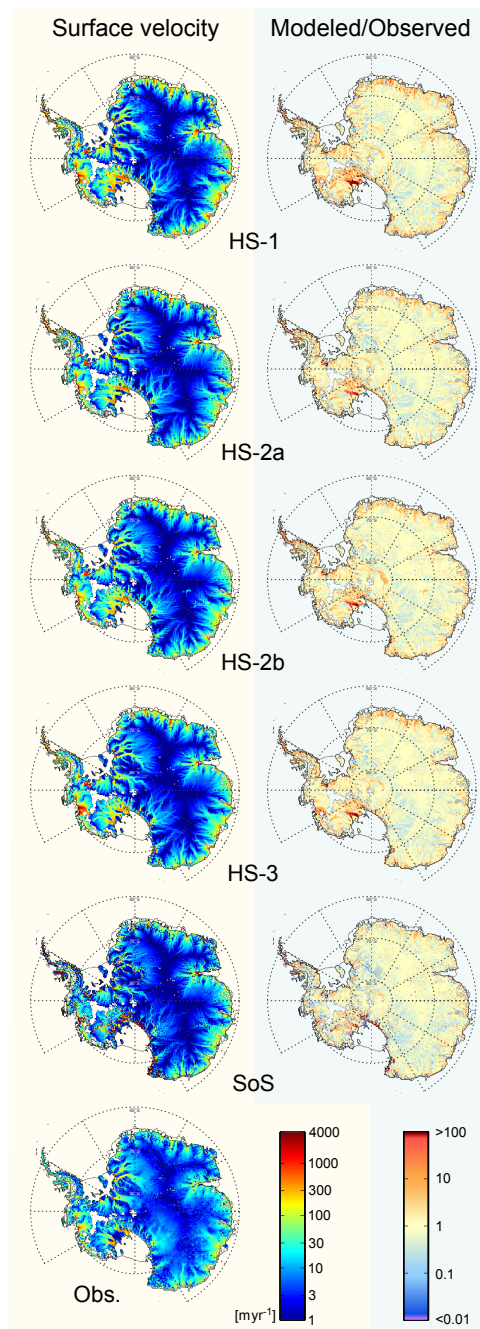
Figure 4. Left column: Equilibrium surface velocities across the grounded ice areas derived from different schemes, in $\mathrm{m\,yr^{-1}}$, compared to the observational data set from Rignot et al. (2011) (bottom), regridded to the model resolution of $20\,\mathrm{km}$. **Right column:** Ratios of the modelled to observed surface velocities, plotted on a logarithmic scale. Velocities smaller than $2\,\mathrm{m\,yr^{-1}}$ are excluded. Color-code saturates at ratios larger than 100 or smaller than 0.01.
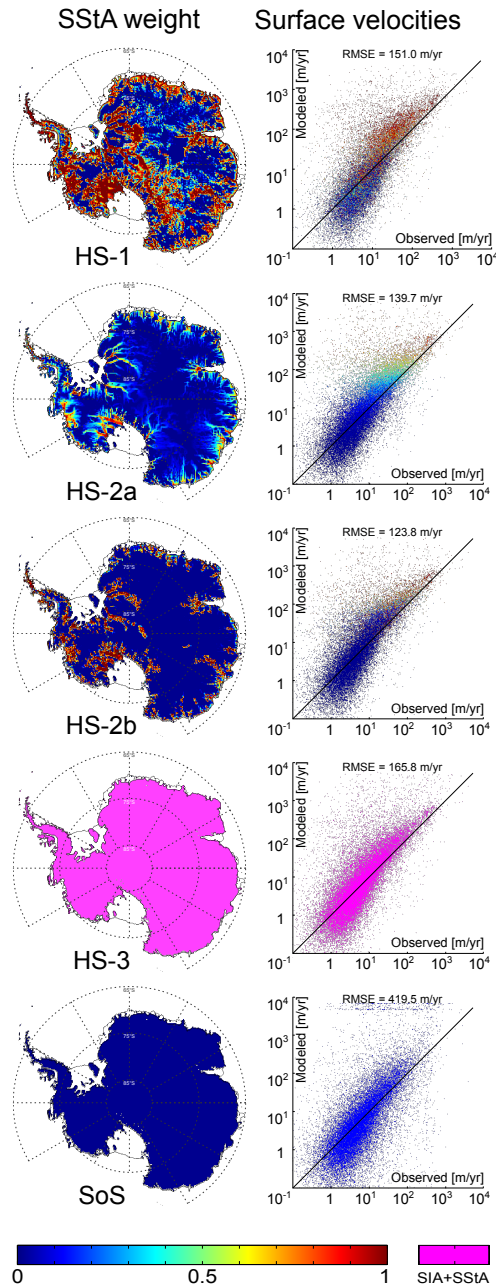
Figure 5. Left column: Equilibrium SStA weights derived from different schemes. The HS-3 does not use a weighting function and simply adds velocities derived from both shallow approximations. **Right column:** Scatter plots of modelled vs. observed surface velocities, in $\mathrm{m\,yr}^{-1}$. Each point is color-coded according to the corresponding SStA weight.
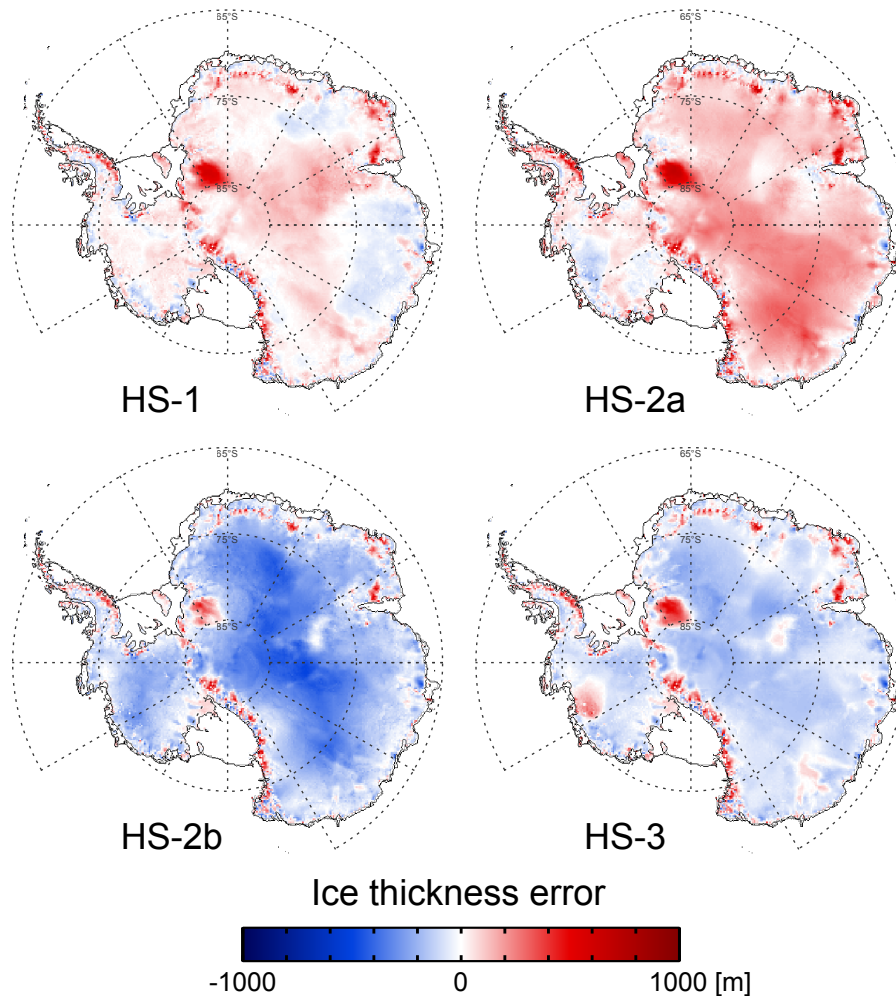
Figure 6. Difference between the modelled and observed ice thickness for all hybrid schemes, in m, after additional $100\,\text{kyr}$ runs in which a median of the inferred distributions of C_0 is prescribed at the ice sheet base. The values of C_0 derived from the SoS are not included in the median distribution.
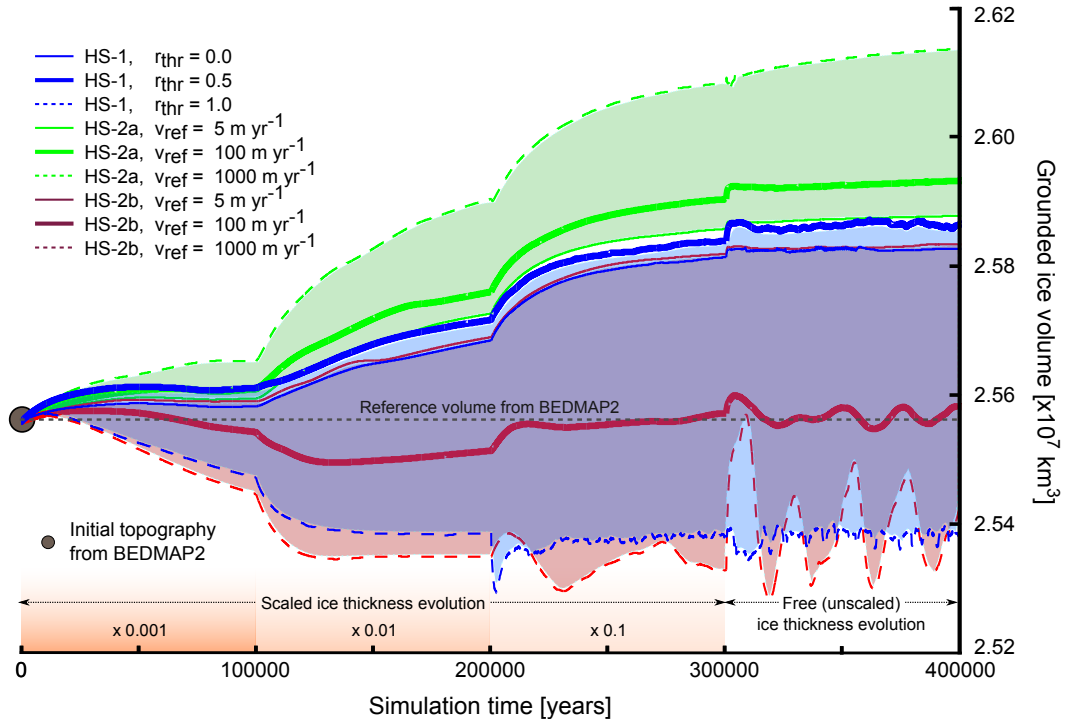
Figure 7. The evolution of the total grounded ice volume during the calibration procedure, in km^3, for different values of free parameters included in the computation of the SStA weight. The volume spread contains the reference values used for the experiments discussed in Sections 5.1–5.3 (solid thick lines), and values that are representative of the lower and upper limits of possible parameter ranges (thin solid and dashed lines, respectively).
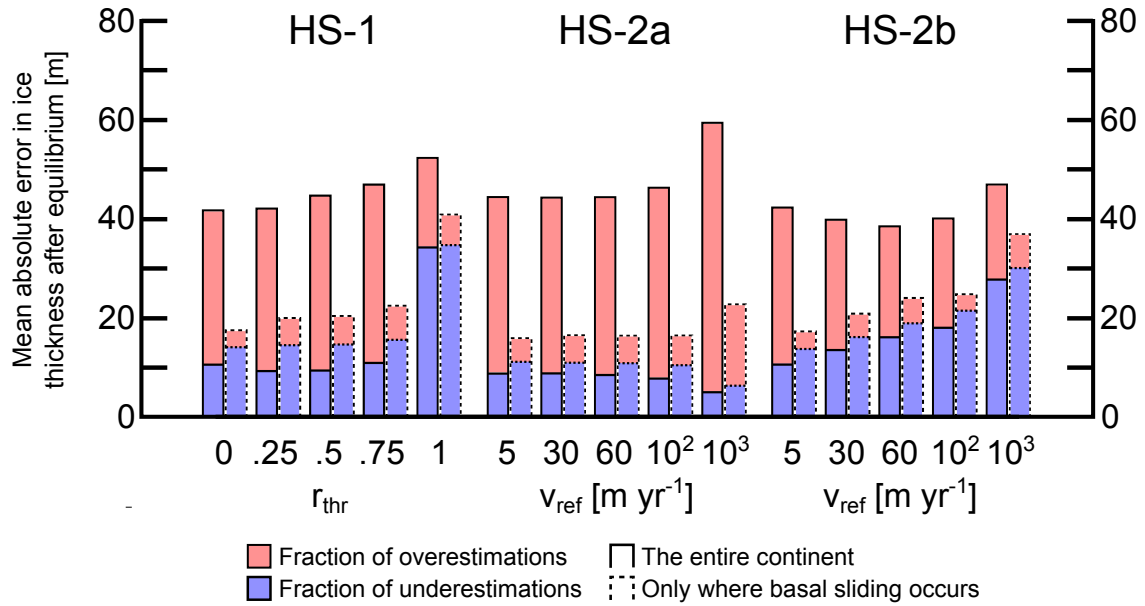
Figure 8. Mean differences between the modelled and observed ice thickness at the end of the equilibrium simulations, in $\times 10^7 \, \text{km}^3$, for different values of free parameters included in the computation of the SStA weight. As in Fig. 2, a mean error is calculated for the entire ice sheet (left bars), and separately over the areas where basal sliding is identified (right bars). Each bar is divided into relative errors arising from under- (blue) and overestimations (red).