https://doi.org/10.5194/egusphere-2022-236
Preprint. Discussion started: 23 May 2022

Geothermal heat flux is the dominant source of uncertainty in englacial-temperature-based dating of ice-rise formation

Aleksandr Montelli[1,2], Jonathan Kingslake[1]

[1]Lamont-Doherty Earth Observatory, Columbia University, Palisades, New York, USA
[2]Scott Polar Research Institute, University of Cambridge, Cambridge, UK

Correspondence to: Aleksandr Montelli (aim39@cam.ac.uk)

Abstract. Ice rises are areas of locally grounded, slow-moving ice adjacent to floating ice shelves. Temperature profiles measured through ice rises contain information regarding changes to their dynamic evolution and external forcings, such as past surface temperatures, past accumulation rates and geothermal heat flux. While previous work has used borehole temperature-depth measurements to infer one or two such parameters, there has been no systematic investigation of parameter sensitivity to the interplay of multiple external forcings and dynamic changes. A one-dimensional vertical heat flow forward model developed here examines how changing forcings affect temperature profiles. Further, using both synthetic data and previous measurements from the Crary Ice Rise in Antarctica, we use our model in a Markov Chain Monte-Carlo inversion to demonstrate that this method has potential as a useful dating technique that can be implemented at ice rises across Antarctica. However, we also highlight the non-uniqueness of previous ice rise formation dating based on temperature profiles, showing that using nominal values for forcing parameters, without taking into account their realistic uncertainties, can lead to underestimation of dating uncertainty. In particular, geothermal heat flux represents the dominant source of uncertainty in ice-rise age estimation. For instance, in Crary Ice Rise higher heat flux values (i.e., about 90 mW m^{-2}) yield grounding timing of 1400±800 years, whereas lower heat flux of around 60 mW m^{-2} implies earlier ice rise formation and lower uncertainties in the ice rise age estimations (500±250 years). We discuss the utility of this method in choosing future ice drilling sites and conclude that integrating this technique with other indirect dating methods can provide useful constraints on past forcings and changing boundary conditions from in-situ temperature-depth measurements.

1. Introduction

Present-day englacial temperatures are the product of the millennial-scale histories of ice flow and thermal boundary conditions experienced by an ice sheet (Robin, 1955). Temperature measurements from boreholes drilled through ice sheets have been widely used to extract important paleoclimatic archives, such as surface temperature and accumulation history, as well as information about the conditions at the base of an ice sheet (i.e., glacial thermal regime and geothermal heat flux),

https://doi.org/10.5194/egusphere-2022-236
Preprint. Discussion started: 23 May 2022

both in Antarctica and Greenland (e.g., Dahl-Jensen et al., 1986; 1998; Engelhardt, 2004a; Orsi et al., 2012; Cuffey et al., 2016). In Antarctica, the ice sheet contains ice rises - regions of slow-flowing, locally elevated, grounded ice embedded within or adjacent to fast-flowing, floating ice shelves; one way they form is through an ice shelf grounding on marine bed (e.g., Martin and Sanderson, 1980; Matsuoka et al., 2015; Wearing and Kingslake, 2020). This shift in boundary condition at the base of an ice-sheet results in a transient evolution of the temperature-depth profile within an ice rise (MacAyeal and Thomas, 1980; Bindschadler et al., 1990). Therefore, due to their proximity to marine ice-sheet periphery and negligible horizontal flow, ice rises can retain an imprint of past grounding line migration on millennial timescales, a record that is otherwise largely inaccessible beneath the ice sheet or its fringing ice shelves (e.g., Conway et al., 1999; Matsouka et al., 2015; Kingslake et al., 2018; Neuhaus et al., 2020).

Past work used temperature-depth measurements within ice rises in Antarctica to estimate the timing of ice-shelf grounding. For instance, Bindschadler et al (1990) developed an advection-diffusion thermal model of Crary Ice Rise, West Antarctica (Fig. 1). The model calculated the initial steady-state ice-shelf temperature profile from the specified set of parameters, including ice thickness, surface temperatures and accumulation rates. The steady-state ice-shelf profile used thermal properties of the bed (geothermal heat flux, diffusivity and conductivity of the bedrock) and was perturbed using a specified vertical ice velocity function to calculate transient thermal evolution after ice-shelf grounding. The modelled temperature profiles were then compared to in-situ borehole measurements at Crary Ice Rise, and minimizing the mismatch between the measured and synthetic profiles yielded the best age estimate of the ice rise in its thickest part to be 1100 years. Recent work by Neuhaus et al (2020) built upon the model of Bindschadler et al (1990) to evaluate the timing of grounding at three sites in the Ross Sea sector of Antarctica, where previous measurements showed anomalously high basal temperature gradients (Engelhardt, 2004a). These results largely corroborate estimates of late-Holocene re-advance in the region (Kingslake et al., 2019) and associated grounding at these sites between 1100 and 500 years ago (Neuhaus et al., 2020).

Thus, the methods used in these studies have potential if future boreholes are drilled at Antarctic ice rises in locations suspected of undergoing significant dynamics change. Yet, the uncertainties inherent in these approaches must be carefully assessed to target drilling, maximize the utility of borehole drilling, and increase the accuracy of ice-dynamic and paleoclimatic inferences. Previous work has included sensitivity tests where some predefined variables, such as accumulation, ice thickening, and melt rate, were assigned several different values to examine how they affected the final temperature profile, and their relation to inferred timing of grounding (Bindschadler et al., 1990; Neuhaus et al., 2020). Yet, there is a lack of systematic investigations of temperature profile sensitivity to cumulative effects of multiple external forcings and dynamic changes, particularly given that some of the predefined variables (e.g., geothermal heat flux) may have considerable uncertainties (e.g., Fudge et al., 2019). In addition, time-variable parameters, such as ice thickness, accumulation and surface temperature may significantly increase the dimensionality of the problem, solutions to which need optimized inversion methods, as opposed to exhaustive global search algorithms where highly dimensional inversion tasks become computationally unfeasible (Mosegaard and Tarantola, 1995).

https://doi.org/10.5194/egusphere-2022-236
Preprint. Discussion started: 23 May 2022

Here we use a Monte-Carlo forward modelling approach to investigate how the interplay between forcings (i.e., surface temperatures, accumulation rates, heat flux and thickness histories) affect the englacial temperatures. Using previous measurements from the Crary Ice Rise, West Antarctica (Fig. 1), we also implement a Markov Chain Monte-Carlo inversion to explore the contributions of multiple parameter uncertainties to the inferred timing of ice-rise formation.

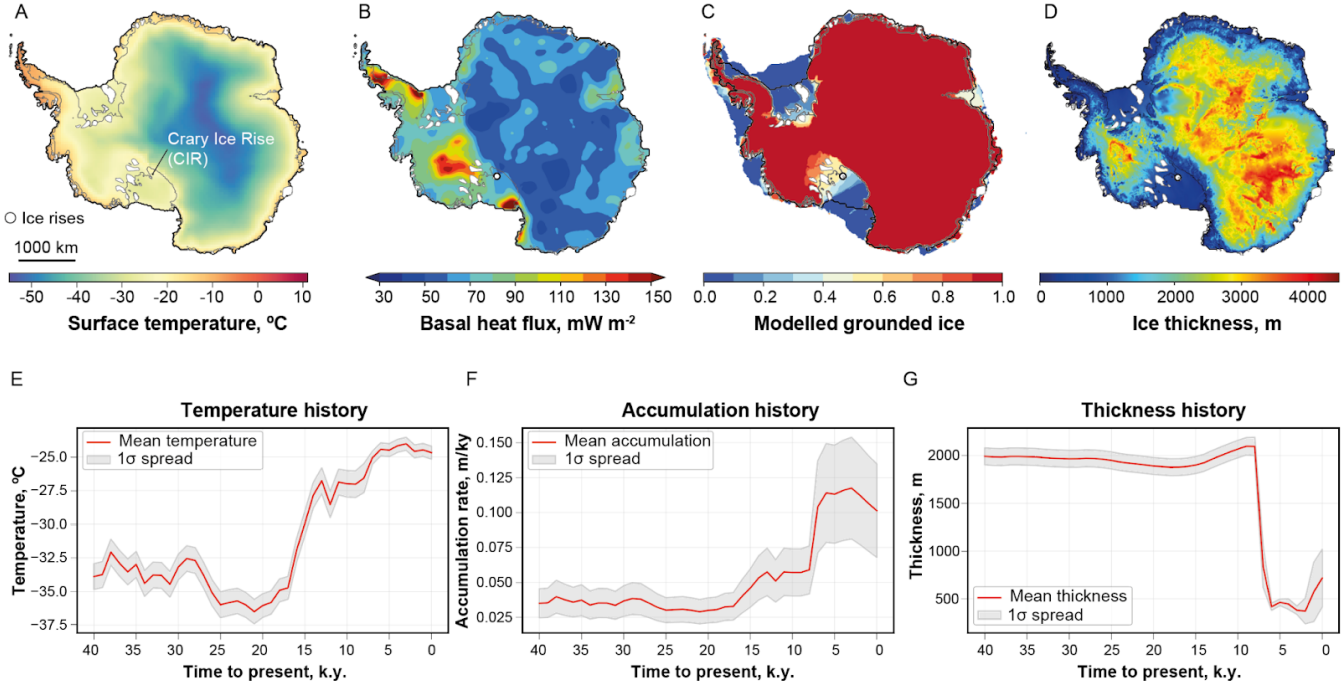


Fig. 1. Present-day distribution of key parameters affecting temperature-depth profiles in Antarctica: A. ERA-Interim annual mean surface temperatures (Albrecht et al., 2020a). B. Basal geothermal heat flux reconstructed from magnetic data (Martos et al., 2017). C. Ensemble-score weighted grounded mask for the present-day ice-sheet reflecting possible histories of grounding line migration since deglaciation (Albrecht et al., 2020b): Red color indicates grounded areas which are covered by grounded ice in all simulations. Blue color indicates areas which are covered by few simulations with low scores. D. Ice thickness (Fretwell et al., 2013). The grounding line and coastlines are shown in black. White dot shows the location of Crary Ice Rise, where previously timing of grounding was inferred using borehole temperature measurements (Bindschadler, 1990). E-G. 40 k.y. - histories of surface temperature, accumulation and thickness in the vicinity (i.e., within about 200 km) of Crary Ice Rise obtained from large-scale Parallel Ice Sheet Model (PISM) simulations (Kingslake et al., 2018; Albrecht et al., 2020a; 2020b). White areas in A-D indicate locations of ice rises from an Antarctic-wide inventory presented in Matsouka et al. (2015).

2. Methods

In this section we outline the numerical forward model and inversion method. The forward model builds upon and extends the models used by Alley and Koci (1990), Orsi et al (2012) and Neuhaus et al (2020). The inversion method is similar to that previously used to reconstruct past surface temperatures in Greenland and Antarctica (e.g., Dahl-Jensen et al., 1998; 1999). Extending previous work, we include the effects of different vertical velocity functions, we use an optimised set of

https://doi.org/10.5194/egusphere-2022-236
Preprint. Discussion started: 23 May 2022

some parameters (e.g., pressure melting/freezing point of ice/seawater), and we introduce temporal variability to ice thickness, surface temperature and accumulation rates, as well as multiple phases of grounding and ungrounding and corresponding changes to the boundary conditions.

2.1. Forward model

2.1.1. Model equations

Since we assume that the ice column is located close to the center of the ice rise where the horizontal velocity and the internal heat production can be neglected (e.g., Dahl-Jensen et al., 1999), we use the following form of the vertical diffusion-advection equation to simulate the time- and depth-evolution of temperature T:

$$\frac{\partial T}{\partial t} = \alpha \frac{\partial^2 T}{\partial z} - w \frac{\partial T}{\partial z} \quad (1)$$

where t is time, z is height above the bed, w is vertical ice velocity (positive upwards) and α is thermal diffusivity:

$$\alpha = k\rho c \quad (2)$$

where ρ is density, k is the thermal conductivity, c is the specific heat capacity. We assume that α is uniform and constant.

For sensitivity experiments, we implement three analytical approximations for vertical velocity w within grounded ice. The Dansgaard and Johnsen (1969) vertical flow approximation can be formulated as:

$$w(z, z_k) = -w_s(t) \frac{2z - z_k}{2H - z_k} \qquad \text{for } z_k \leq z \leq H \quad (3)$$

$$w(z, z_k) = -w_s(t) \frac{z^2}{(2H - z_k) z_k} \qquad \text{for } 0 \leq z \leq z_k \quad (4)$$

where $z = 0$ represents the ice-bedrock interface, H is the ice thickness and z_k is a free parameter (Martín & Gudmundsson, 2012). Lliboutry (1979) provides another commonly used approximation (Wearing & Kingslake, 2019, Fudge et al., 2019):

$$w(z, t) = w_s(t) \left[1 - \frac{n+2}{n+1} \left(1 - \frac{z}{H(t)} \right) + \frac{1}{n+1} \left(1 - \frac{z}{H(t)} \right)^{n+2} \right] \quad (5)$$

where n is a rheological parameter (Glen's flow law exponent, $n=3$).

Finally, a simplified approximation where vertical velocity varies linearly with depth from zero at the base to a maximum value at the surface, following Bindschadler et al (1990) was implemented for sensitivity experiments and estimation of associated uncertainties:

https://doi.org/10.5194/egusphere-2022-236
Preprint. Discussion started: 23 May 2022

$$w(z,t) = w_s(t)\frac{z}{H(t)} \quad (6)$$

When time-varying ice thickness is introduced, the difference between accumulation and thickening rates determines the vertical velocity at the surface:

$$w(z,t) = a(t) - \frac{\partial H(t)}{\partial t} \quad (7)$$

For temperature calculations within the underlying bedrock, no vertical advection is assumed (w=0), reducing Eq. 1 to only account for heat diffusion.

2.1.2. Spatiotemporal domain and boundary conditions

We define a one-dimensional spatial domain that extends vertically from the bedrock base to the ice surface. The vertical coordinate is z, which increases upwards, and z_b and z_s are the elevations of the ice base and ice surface, respectively. We implement a simple finite-difference scheme to solve Eq. 1. We discretize the spatial domain with a minimum of 50 nodes within the ice column and 50 nodes with fixed 10 m vertical spacing in the subjacent bedrock (Neuhaus et al., 2020). We used a time step of 1 year. In a scenario where ice thickness varies through time, vertical grid spacing in the ice column is adapted and temperature values are interpolated accordingly at each time step. The model can be run under two different assumptions about whether the ice column is a part of a grounded ice rise or a floating ice shelf.

In the ice-shelf scenario (Fig. 2A) boundary conditions at the ice surface are set by time-variable surface temperatures $T(t)$ and accumulation rates $a(t)$. The temperature at the top of bedrock/base of ice column $T(z=z_b)$ is forced to equal the freezing point of seawater (calculated as a function of ice thickness, to account for its pressure dependence), and the temperature gradient in the bedrock is calculated based on the geothermal heat flux G (Millero, 1978; Determann and Gerdes, 1994). We assume that the rates of accumulation and basal melting balance the depth-uniform vertical velocity (Holland and Jenkins, 1999).

When modelling temperatures within a grounded ice sheet (i.e., an ice rise; Fig. 2B), the boundary condition at the base of the ice column is set such that the vertical gradient in T corresponds to a diffusive heat flux (there is no advection because $w(z_b) = 0$) that balances the geothermal heat flux G through the underlying bedrock. Thickness history $H(t)$ and the vertical velocity function (Eqs. 3-6) within the ice rise are also used. The boundary conditions at the top of the ice column are prescribed based on the temperature and accumulation rate histories; prior values of parameters that dictate these boundary conditions are discussed further in Section 2.1.3.

https://doi.org/10.5194/egusphere-2022-236
Preprint. Discussion started: 23 May 2022

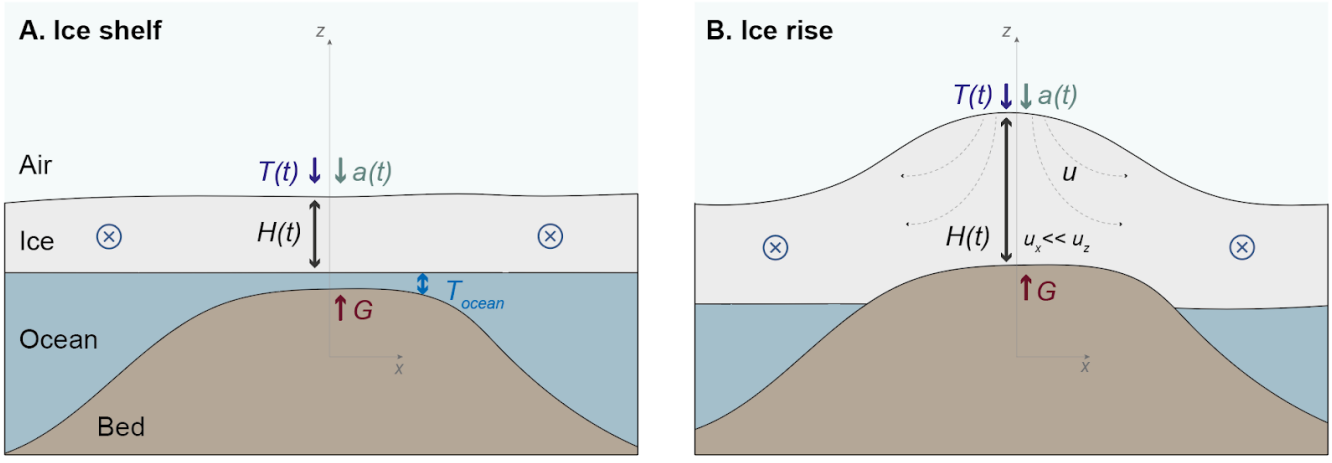


Fig. 2. Geometry of the ice shelf - ice rise heat flow model. The simple 1D vertical model is implemented along a stationary crest of hypothetical ice rise (z axis on the figure) and bedrock beneath it. *T(t)* and *a(t)* indicate surface temperature and accumulation forcings, respectively. *G* is geothermal heat flux, *u* is ice flow velocity (along the crest of ice rise, horizontal flow velocity u_x is negligibly small comparable to its vertical component u_z, or *w*).

2.1.3. Input prior parameters

Parameters characterising the properties of ice and the bedrock are assumed to be constant (Table 1). Values of geothermal heat flux at specific locations are sampled from the Antarctic-wide heat flux data compilation derived from spectral analysis of airborne magnetic data (Martos et al., 2017). Surface temperature evolution, accumulation rate histories and ice thickness histories are sampled from distributions provided by an ensemble of simulations using the Parallel Ice Sheet Model (PISM) (Fig. 1E-F, Kingslake et al., 2018; Albrecht et al., 2020a, 2020b). PISM is a three-dimensional, thermomechanical ice-sheet model that solves a hybrid shallow approximation of Stokes flow. PISM produces continental-scale (16 km grid cell size), long-term (multi-millenial length) simulations of ice-sheet thickness. Associated with these fields are surface-temperature reconstructed from the West Antarctic Ice Sheet (WAIS) divide ice-core (Cuffey et al., 2016), scaled by modelled ice-surface elevation and accumulation patterns simulated by the regional climate model RACMOv2.1 (Ligtenberg et al., 2013), scaled by 2% per degree of temperature change from present (Kingslake et al., 2018). Other specifications and parameters used for the chosen PISM model output, including mantle viscosity and flexural rigidity are described in detail in Kingslake et al. (2018). Despite inherent uncertainties associated with large-scale numerical models, which have relatively coarse resolution compared to the length-scales of ice rises, PISM outputs provide the first-order estimates of prior information about the time-variable parameters (i.e., temperature, accumulation and ice thickness) at any specified location on the Antarctic Ice Sheet (AIS) (Fig. 1E-F).

Table 1. Numerical values of the parameters used in the simulations

https://doi.org/10.5194/egusphere-2022-236
Preprint. Discussion started: 23 May 2022

Parameter	Description	Value	Units
ρ_i	Density of ice	918	kg m^{-3}
ρ_b	Density of bedrock	2750	kg m^{-3}
k_i	Heat capacity of ice	2.3	J kg^{-1} K^{-1}
k_b	Heat capacity of bedrock	2.8	J kg^{-1} K^{-1}
c_i	Thermal conductivity of ice	2000	W m^{-1} K^{-1}
c_b	Thermal conductivity of bedrock	790	W m^{-1} K^{-1}

2.1.4. Experimental design and computational environment

Prior to imposing time-variable forcings, we allow the simulation to reach a steady state using surface temperatures, thickness, and accumulation rates equal to their values at the beginning of the chosen simulation period (Fig. 1E-G). The transient temperature profile is then simulated using the histories of surface temperatures, ice thickness and accumulation rates throughout the specified period. During the transient simulation, once a grounding/ungrounding event is introduced, the boundary conditions at the base of the ice column switch accordingly (see Section 2.1.2). The calculated temperature profile at the last time step of the chosen period represents the final product of a single simulation. For forward Monte-Carlo sensitivity experiments, we performed an ensemble of simulations, in which uncertain parameters were repeatedly randomly sampled from a range of their prior distributions. The resultant temperature profiles were then compared to examine how altering our prior parameters affected the final temperature profiles. In order to evaluate the fit between two temperature profiles, we used a root-mean squared mismatch,

$$\boldsymbol{RMS} = \sqrt{\frac{\sum_{i=1}^{n}(T_{m_i} - T_{p_i})^2}{n}} \quad (8)$$

where $\boldsymbol{T}_{m_i}$ and $\boldsymbol{T}_{p_i}$ are the measured and predicted temperatures at grid point i, and n is the number of grid points.

To increase computation efficiency in our forward Monte-Carlo models, we used Dask, a flexible library for parallel computing in Python, which was implemented using clusters provided by the Pangeo project (Odaka et al., 2019). Pangeo is a developing ecosystem of scalable, open-source tools for cloud-based parallel computation and interactive large-scale computation and data analysis. Automatic parallelization on tens to hundreds of workers significantly increases computation performance, as compared to a standard approach using a desktop computer, and is particularly applicable to tasks that are easily parallelized, such as forward Monte-Carlo simulations.

https://doi.org/10.5194/egusphere-2022-236
Preprint. Discussion started: 23 May 2022

2.2. Inverse model

Where borehole temperature-depth measurements are available, they can be used to infer the history of dynamic change and evolution of boundary conditions (e.g., Dahl-Jensen et al., 1999) via inversion (e.g., Mosegaard and Tarantola, 1995). In this section, we outline what observational data and inversion methods we use to infer past ice-rise evolution.

2.2.1. Temperature-depth data

Among numerous ice rises mapped across the AIS only a few have been sampled by borehole temperature measurements (Matsouka et al., 2015). These sites include Siple Dome and Crary Ice Rise of the Ross Ice Shelf, as well as Mill Island of the Shackleton Ice Shelf in East Antarctica (e.g., Koci and Bindschadler, 1989; Engelhardt et al., 2004b; MacGregor et al., 2007; Roberts et al., 2013). We digitized Bindschadler et al.'s (1990) temperature observations from Crary Ice Rise and used them as input data for our Bayesian inversion method (i.e., probabilistic data analysis that involves using the prior information and computing the posterior probability distribution for the parameters of the model; Section 2.2.2). Due to the quality of available data, digitisation implies inherent uncertainties (up to 0.1ºC RMS difference between independently digitised temperature profiles). Furthermore, as uncertainties associated with measurement calibrations can reach up to 0.1°C (e.g., Orsi et al., 2012), we consider 0.1ºC as an upper threshold value for estimating the degree of fit between measured and predicted temperature profiles. In addition, we used the forward model outlined in Section 2.1 with predetermined parameters to produce synthetic temperature-depth data. These synthetic profiles were then used as an input for validation of our inverse method.

2.2.2. Markov Chain Monte Carlo inverse method

The Markov Chain Monte Carlo (MCMC) method tests randomly selected combinations of prior variables using a random walk through a high-dimensional parameter space. The variables are assumed to be 'unknown' parameters and are prescribed prior distributions (or simply a range of realistic prior values). The forward model uses these parameters from their prior distributions as inputs, and its output is compared to the measured (or synthetic) temperature profiles. In each step of random walk with predefined length (i.e., total number of parameter combinations), a perturbed model of the current model is proposed. The MCMC then uses a likelihood function to estimate the agreement between the modelled and measured profiles (Mosegaard and Tarantola, 1995; Dahl-Jensen et al., 1998) using:

$$\boldsymbol{L} = -\mathbf{0.5}\sqrt{\frac{\sum_{i=1}(T_{m_i}-T_{p_i})^2}{T_{error}{}^2}} \tag{9}$$

where $\boldsymbol{T_{m_i}}$ and $\boldsymbol{T_{p_i}}$ are the measured and predicted temperatures for each point of the grid, and T*error* is the uncertainty of measurements.

https://doi.org/10.5194/egusphere-2022-236
Preprint. Discussion started: 23 May 2022

The perturbed model is either 'rejected', in which case a new random perturbation is applied from the same starting location in parameter space, or 'accepted', in which case this location becomes the new starting location for the next random perturbation. Even when a new perturbation yields better fit to the observations, the model may be rejected, or if the new perturbation produces larger misfit, the model may be accepted. Whether a model is accepted is based on an acceptance probability (see Dahl-Jensen et al., 1999), which introduces a degree of stochasticity, ensuring that the random walk avoids entrapment in local likelihood maxima (Mosegaard and Tarantola, 1995). Eventually, the paths converge towards the regions corresponding to model parameters that yield the lowest misfit between modelled and observation temperature profiles.

In practice, MCMC can be used to infer any inputs to the forward model. For instance, in the Dye 3 borehole drilled through the 3-km thick Greenland Ice Sheet, the unknown temperature history has been divided into 125 intervals, which, together with heat flux (also assumed to be unknown) yielded a 126-dimensional parameter space (Dahl-Jensen et al., 1999). Here, we focus on inferring past dynamic changes (i.e., timing of grounding), but also assume heat flux, temperature, accumulation, and thickening rates to be unknown in order to explore the possible combinations of realistic parameter values that could provide a close fit between the field observations and modelled temperature-depth profiles. To perform the MCMC we used *Emcee*, a Python-based software package which implements the affine-invariant ensemble sampler (Goodman & Weare, 2010; Foreman-Mackey et al., 2014).

3. Results

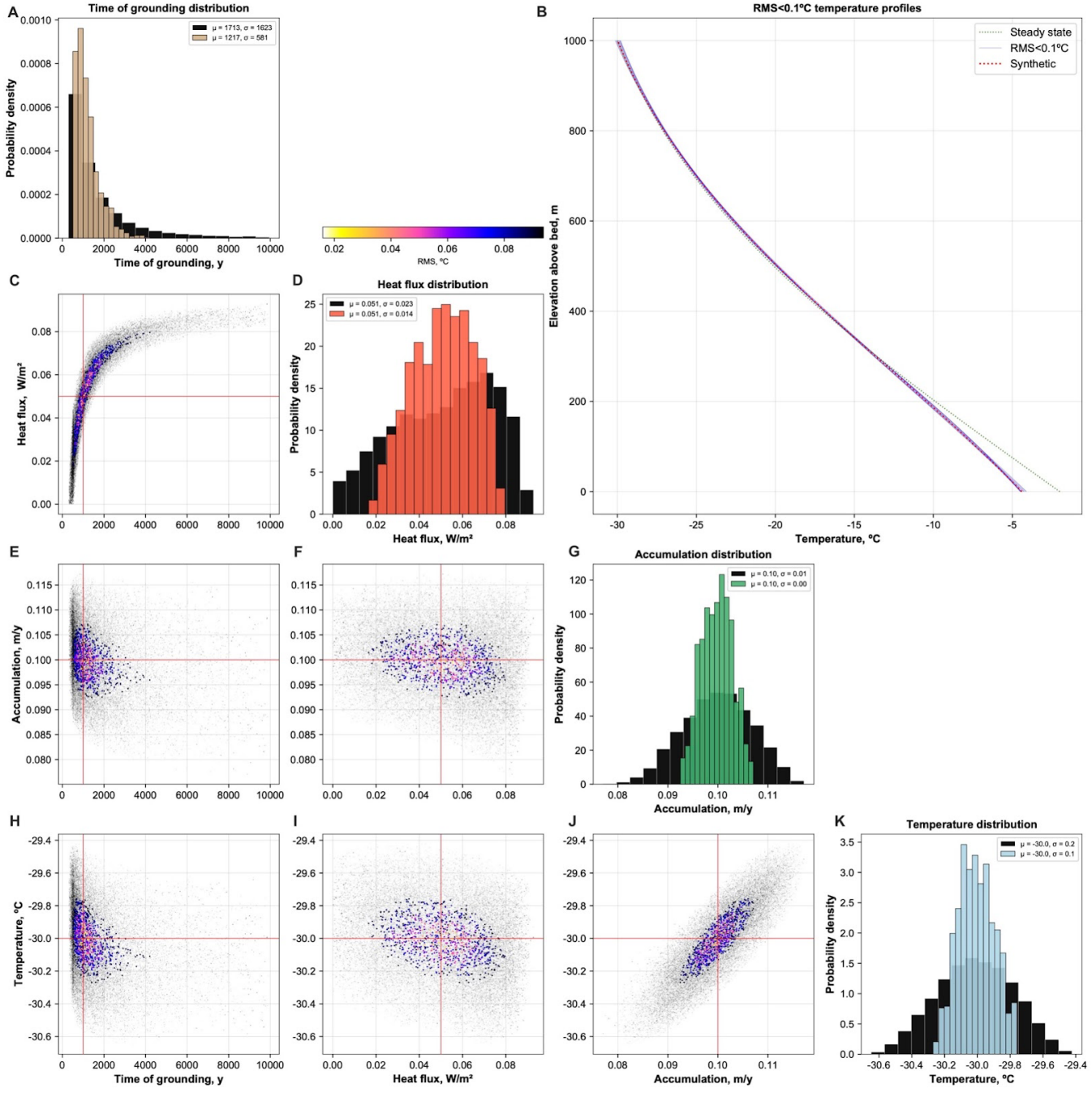
3.1. Inverse modelling

3.2.1. Synthetic data

Prior to using actual borehole measurements, we tested the MCMC inversion method on a synthetic temperature profile calculated using a simplified forward model that assumes a hypothetical 1000-m thick, 1 k.y. old ice rise (i.e., formed when an ice shelf grounded 1000 years ago), forced by a heat flux of 50 mW m^{-2}, a constant surface temperature of -30º C and an accumulation of 0.1 m yr^{-1}. This experiment also allows us to evaluate some aspects of temperature-depth sensitivity to the key forcing variables (Fig. 3). The variables are prescribed the widest range of possible values as their prior distributions. From sampling of 100,000 different combinations of these variables from their prior range, over 9,000 yield an RMS mismatch with the synthetic temperature profile of less than 0.1ºC, with best fit samples (RMS ~ 0.01ºC) yielding parameters that very closely match the parameters used to produce the synthetic profile (Fig. 3). Nevertheless, even in this idealised setup, a wide range of ice-rise age estimations (mean±standard deviation of ~1200±580 years,) closely match the synthetic profile (i.e., RMS <0.1ºC). Moreover, while derived accumulation rates and surface temperatures are tightly clustered (i.e., within a few percent) of their prescribed values, heat flux values are more widely distributed 50±20 mW m^{-2} (Fig. 3D). In summary, this synthetic test shows that our inverse approach effectively recovers surface parameters

https://doi.org/10.5194/egusphere-2022-236
Preprint. Discussion started: 23 May 2022

(accumulation and temperature), but uncertainty inherent in the system may introduce unavoidable uncertainty in estimated geothermal heat fluxes and ice-rise grounding times.



245 **Fig. 3. Inversion of the synthetic temperature profile data. Synthetic temperature-depth profile (red dotted line in panel B) was produced using the prescribed set of parameter values (shown by the intersection of red solid lines in the scatter plots, C, E, F, H-**

https://doi.org/10.5194/egusphere-2022-236
Preprint. Discussion started: 23 May 2022

J). The scatter plots illustrate the random walk in parameter space: in grey are the tested combinations of parameter values that yielded a less than 0.3ºC RMS misfit with synthetic temperature profile, while in colour are the combination of parameter values that yield best-fit with synthetic profile (i.e., <0.1ºC RMS). The color of the points indicate the RMS. Diagonally placed histograms (A,D,G K) show posterior distributions for each parameter: black bars show parameter distributions that yielded RMS< 0.3ºC RMS; coloured bars show distributions of parameter values that yielded best-fit with synthetic temperature profile (i.e., <0.1ºC RMS). Insets in histograms display the mean, μ, and standard deviation, σ, of each distribution. B. Randomly selected resultant temperature profiles for 0.1ºC RMS misfits.

3.2.2. Crary Ice Rise borehole measurements

To implement our MCMC inversion with englacial temperature data from Crary Ice Rise, we prescribe a wide range of prior values for model parameters. This allows us to probabilistically evaluate the possible combinations of realistic parameter values that could closely match the measured temperature-depth profiles (Fig. 4). For example, instead of assigning a heat flux of 77 mW m^{-2}, as used by Bindschadler et al (1990), we obtain prior distributions of these values within a 200-km region around Crary Ice Rise using Martos et al.'s (2017) heat flux reconstruction (Section 2.1.3). Similarly, based on large-scale PISM model simulation outputs in the Crary Ice Rise area, we allow accumulation rates to vary between 5 and 20 cm yr^{-1} (Fig. 4E-G). Here, our results are based on tests of 250,000 combinations of five parameters: surface temperature, accumulation rate, initial thickness, timing of grounding and geothermal heat flux.

Within the assigned prior parameter limits, the results show a wide range of posteriors that fit the observed temperature-depth measurements to within measurement errors reported by Bindschadler et al (1990) (i.e., RMS ~0.1ºC). Thus, over 25,000 samples yield a relatively low mismatch (i.e., 0.1<= RMS<=0.15ºC) with Crary Ice Rise measurements, with timing of grounding of 1000±300 years (mean±standard deviation; Fig. 4A).

The distributions of the coloured points in the scatter plots in Fig. 4 provide insight into the dependence of each parameter and its uncertainty on the other parameters. For example, as with synthetic data inversion (Section 3.2.1), the inferred ages of ice rise formation is strongly dependent on heat flux (Fig. 4C). Higher heat flux (i.e., about 90 mW m^{-2}) yields grounding timing of 1400±800 years, whereas lower heat flux of around 60 mW m^{-2} implies earlier ice rise formation and lower uncertainties in the ice rise age estimations (500±250 years). For comparison, using their deterministic least square approach with fixed parameter values, Bindschadler et al (1990) estimated an age of ice rise formation of 1100 years with a standard error between measured and simulated temperature profiles of 0.08ºC. Our best fit (i.e., RMS of 0.1ºC) temperature profile is inferred when the timing of grounding is 1300 years (with heat flux of 86 mW m^{-2}, accumulation of 0.14 m yr^{-1} and initial thickness of 460 m being the estimated best fit values). Here, we focus on the interplay of all parameters and do not assess the inferred initial thickness required to explain grounding at a particular Crary Ice Rise site on the shelf. However, more extra information on the past sea level, as well as the history of bedrock isostatic fluctuations can be easily incorporated into our model to improve initial thickness estimation. Our MCMC inversion experiments highlight that inferring timing of ice-rise formation from englacial temperature measurements requires accurate knowledge of heat flux. However, Figures 3 and 4

https://doi.org/10.5194/egusphere-2022-236
Preprint. Discussion started: 23 May 2022

show that the inverse approach does not illustrate the sensitivity of the temperature profile to model parameters, and we discuss this in Section 3.3.

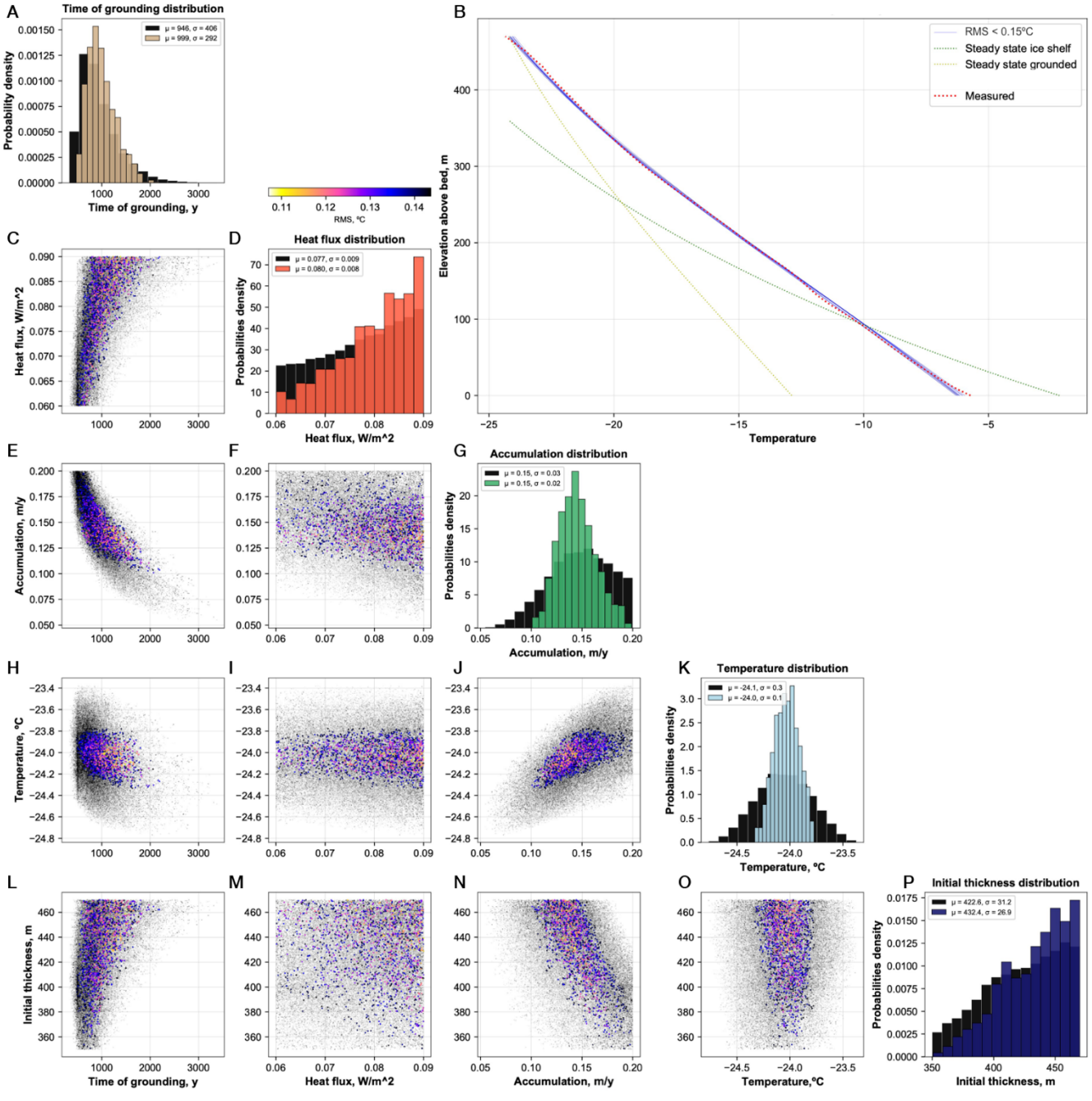


Fig. 4. Inversion of the temperature-depth measurements near Crary Ice Rise Site D. Grey cloud of points illustrate the random walk in parameter space, showing the tested combinations of parameter values that yielded a less than 0.3°C RMS misfit with

https://doi.org/10.5194/egusphere-2022-236
Preprint. Discussion started: 23 May 2022

observations, while the coloured data points show the combination of parameter values that yield best-fit with measured temperature profile (i.e., less than 0.15ºC RMS). Diagonally placed histograms correspond to the posterior distributions for each parameter. B.Randomly selected resultant temperature profiles for 0.3ºC and 0.15ºC RMS misfits.

3.2. Englacial temperature profile sensitivity

To explore temperature profile sensitivity to model parameters, a series of Monte Carlo forward simulations were conducted for each parameter under consideration, and the resultant temperature profiles (Figs. 5A-C) were directly compared for different ice thicknesses (Figs. 5D-F). Temperature effects of changing heat flux and surface temperatures by a fixed value are similar for ice of a given thickness (e.g., for 2-km thick ice 1mW m^{-2} change in heat flux yields about 0.4 ºC (RMS) change in temperature profile, Fig. 5F). In contrast, fixed changes to accumulation rates and timing of grounding reflect a more complicated relation to resultant temperature profiles (Figs. 5D,E). For example, a 0.01 m $k.y.^{-1}$ accumulation change yields a considerably larger effect on englacial temperatures when accumulation rates are low (e.g., for a 2-km thick ice, about 0.4 ºC (RMS) per 0.01 m $k.y.^{-1}$ for accumulation rates around 0.1 m $k.y.^{-1}$, compared to 0.1 ºC (RMS) for accumulation rates of about 0.4 m $k.y.^{-1}$). Similarly, Figure 5E shows that the temperature profiles are much more sensitive to grounding time for younger ice rises.

https://doi.org/10.5194/egusphere-2022-236
Preprint. Discussion started: 23 May 2022

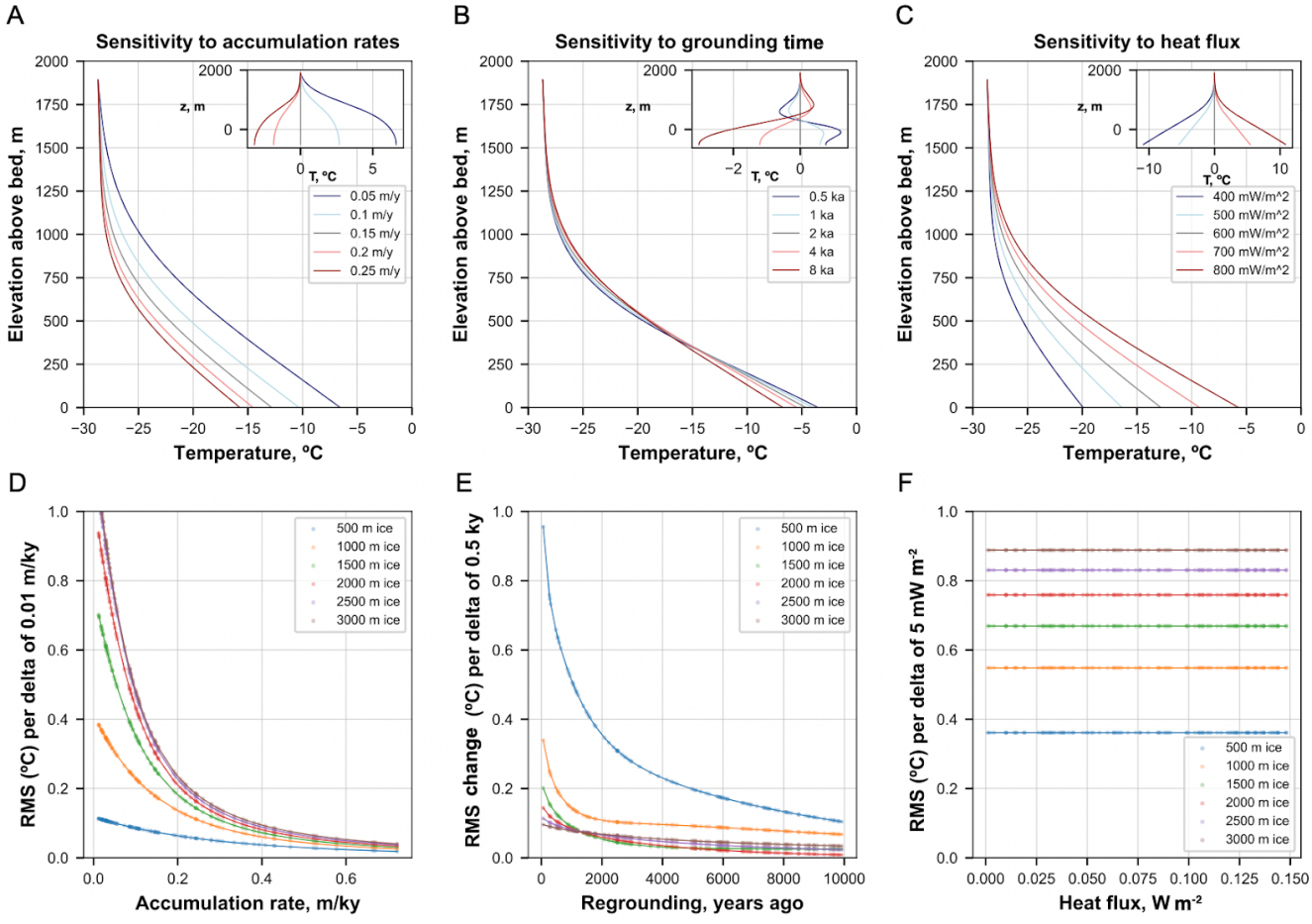


Fig. 5. Englacial vertical temperature profile sensitivity to model parameters. A. Example of five temperature profiles (with differences in the inset) obtained using a forward model for grounded ice with only accumulation rate varied between simulations, while other variables remain constant at continent-wide mean values from the input data. B. Example of five temperature profiles (with differences in the inset) obtained using forward model for grounded ice with only timing of grounding varied, while other variables remain constant at their mean value. C. Example of five temperature profiles (with differences in the inset) with only heat flux varied, while other variables remain constant at their mean value. Bottom row: Results of forward simulations showing how a change (termed 'delta' on the vertical axis label) in D. accumulation rate of 0.01m k.y.$^{-1}$ E. timing of grounding of 500 years, and F. geothermal heat flux of 1mW m^{-2} affects temperature profiles within grounded ice (expressed as root-mean-squared change (RMS) between two profiles).

We also investigate how different velocity functions and temporal variability of thickness, accumulation and surface temperatures may affect the distribution of temperature within grounded ice of various thicknesses (Fig. 6). For these simulations, variable temperature and accumulation for the last 40 k.y. were obtained from PISM simulation outputs (Fig.1E-F).

https://doi.org/10.5194/egusphere-2022-236
Preprint. Discussion started: 23 May 2022

Advection effects of chosen vertical velocity approximations on englacial temperature profiles are illustrated in Figure 6A, D. Temperatures are increasingly affected in the lower part of the ice column (inset on Fig. 6A). Deviations between profiles produced with Dansgaard and Johnsen (1969) and Lliboutry (1979) velocity approximations also increase with ice thickness. For example, for 500 m thick ice the RMS difference between two profiles is 0.18ºC, and 2.55 ºC for 2500 m ice column (Fig. 6D).

Inversion experiments described in Section 3.1 assumed constant surface temperatures, T_s, and accumulation rates, a, over the simulation period. To examine how temporal variability of these parameters impacts the temperature-depth data, we ran forward Monte-Carlo simulations and compared resultant profiles for four instances (T_s and a constant, T_s constant and a variable in time, a constant and T_s variable in time, both variable in time- see Fig. 6B). These experiments reveal considerable deviations between profiles forced only by one time-variable parameter, and while these effects typically do not exceed 1ºC RMS for ice thinner than 1 km, they result in as much as few degrees RMS difference between two profiles within ice about 3 km thick (Figs. 6B,E).

Finally, the effect of the ice thickening rate are most significant in relatively thin ice (up to 2ºC RMS change for ice 500 m thick, Figs. 6C,F), but decrease rapidly as ice becomes thicker. Overall, forward sensitivity experiments outlined in this section provide insight into what model parameters have the stronger effect on the ice column, how these effects manifest in the shape of temperature profile, and how this is affected by ice thickness.

https://doi.org/10.5194/egusphere-2022-236
Preprint. Discussion started: 23 May 2022

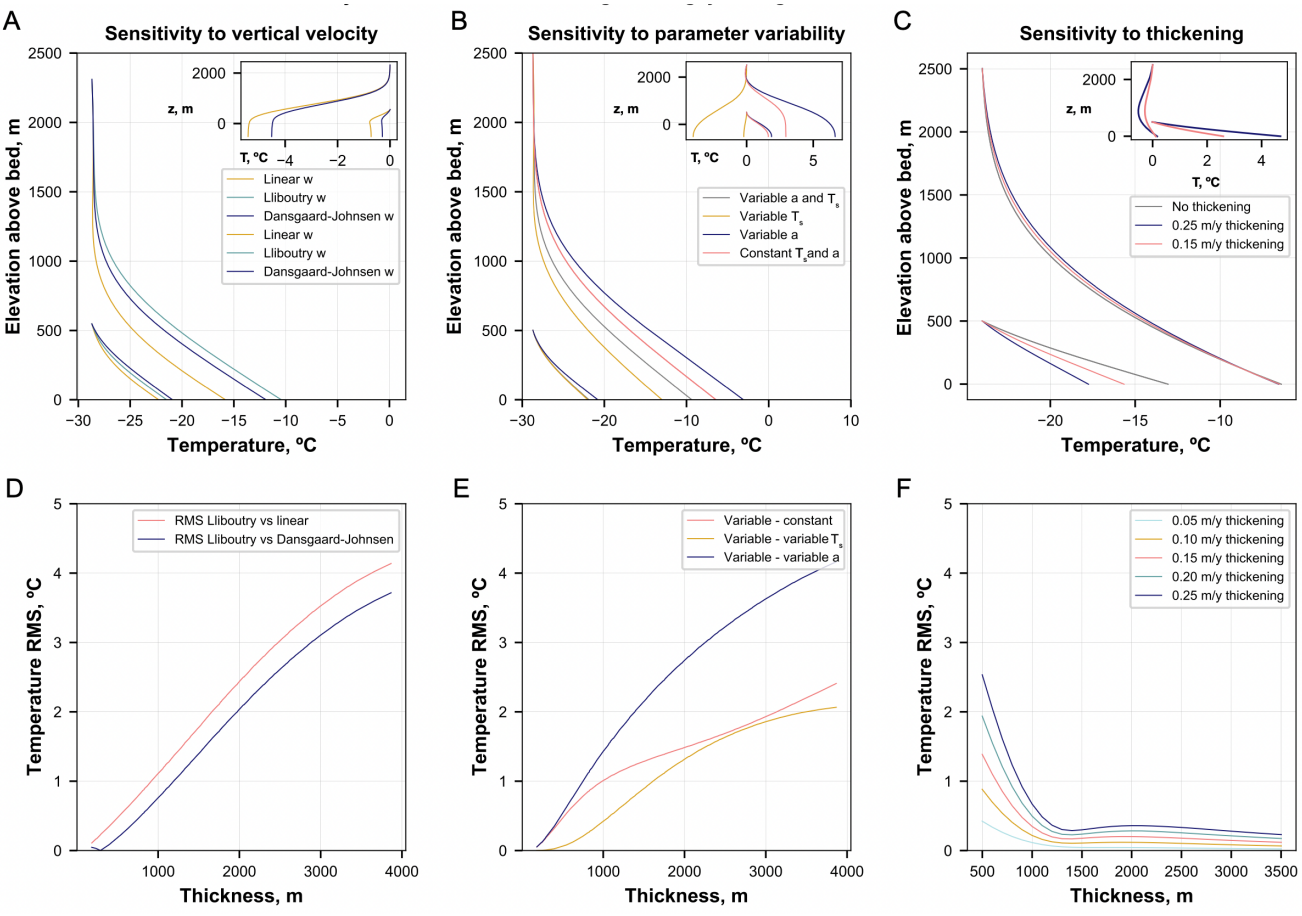


Fig. 6. Englacial vertical temperature profile sensitivity to vertical velocity function and accumulation/surface temperature temporal variability. A. Comparison of three temperature profiles obtained using a forward model for grounded ice with the Lliboutry, Dansgaard-Johnsen and linear vertical velocity approximations for 500 and 2500 m thick ice. B. Comparison of four temperature profiles obtained using forward model for grounded ice with and without temporal variability of surface temperature and accumulation. C. Comparison of three temperature profiles obtained using a forward model for grounded ice assuming three different thickening rates. Bottom row: Results of forward simulations showing how for a range of ice thicknesses temperature profile differ depending on D. various vertical velocity models; E. temporally variable and constant parameters, and F. ice thickening rates.

4. Discussion

4.1. Inversion of ice-rise age

Bayesian inversion of englacial temperature-depth profiles indicates that inferences of ice rise age (i.e., timing of grounding) may significantly vary depending on the values of other forcing parameters. Among these, geothermal heat flux may have a significant effect on the inferred timing of grounding, with lower heat flux yielding earlier age estimates, along with much smaller corresponding uncertainties (that is, narrower range of solutions acceptable within a prescribed degree of misfit, Figs. 3, 4). In case of the Crary Ice Rise inversion experiment, we infer a range of possible ice rise age that encompasses

https://doi.org/10.5194/egusphere-2022-236
Preprint. Discussion started: 23 May 2022

the value of 1100 years previously reported by Bindschadler et al (1990), but increases from 500±250 to 1400±800 years ago, as the corresponding values of heat flux decrease (assumed here to vary between 60 mW m^{-2} and 90 mW m^{-2}).

The effect of heat flux may be significant when prior knowledge about its values is poor or unconstrained. For instance, in the synthetic data inversion experiment with prescribed grounding timing of 1000 years, heat flux of 30 mW m^{-2} yields ice rise age estimation of 520±80 years, whereas a value of 80 mW m^{-2} corresponds to ice grounding d 4030±610 years ago, with both combinations falling within misfit of 0.1ºC RMS (Figs. 3, 7). This result is important because previous deterministic approaches, for example the one implemented by Bindscahdler et al (1990), assumed a fixed value of heat flux. However, recent studies have shown that despite its importance to understanding ice sheet evolution (Pollard et al., 2005; Seroussi et al., 2017), heat flux beneath ice sheets remains relatively poorly constrained, with discrepancies between continental-scale reconstructions and targeted heat flux measurements, and associated uncertainties of up to ±15 mW m^{-2} (e.g., An et al., 2015; Martos et al., 2017; Fudge et al., 2019). Furthermore, heat flux at the base of an ice sheet may show substantial lateral variations (i.e., up to 200 mW m^{-2}) over relatively short distances of less than 100 km, as for example observed below the Whillans Ice Stream (Begeman et al., 2017), as well as in other locations beneath the Antarctic and Greenland ice sheets (e.g., Cuffey et al., 1995; Dahl-Jensen et al., 2003; Schroeder et al., 2014; White-Gaynor et al., 2019). Careful estimation of heat flux uncertainties and their incorporation in temperature-depth inverse models are therefore essential for accurate estimation of past dynamic changes to ice rises.

The uncertainties in ice rise age estimates are also determined by the degree of accepted misfit between predicted and measured temperature profiles, which in turn relies on the accuracy of englacial borehole thermometry. Depending on the instrumentation used, previous temperature-depth data have been collected with typical uncertainties of around 0.1ºC (Orsi et al., 2012; Yang et al., 2017, Fudge et al 2019), and occasionally up to 0.05ºC and even 0.02ºC, as demonstrated by the work on the Antarctic Ice Sheet and Himalayan glaciers (Van Ommen et al 1999; Miles et al., 2017; Talalay et al 2020). Improvements in accuracy of measurements can put tighter constraints on inverted parameters: for example, in our synthetic data inversion experiments, decreasing uncertainties from 0.1ºC to 0.05ºC more than doubles the accuracy of the inferred timing of grounding, as well as significantly narrows the range of accepted heat flux values (Fig. 7A).

https://doi.org/10.5194/egusphere-2022-236
Preprint. Discussion started: 23 May 2022

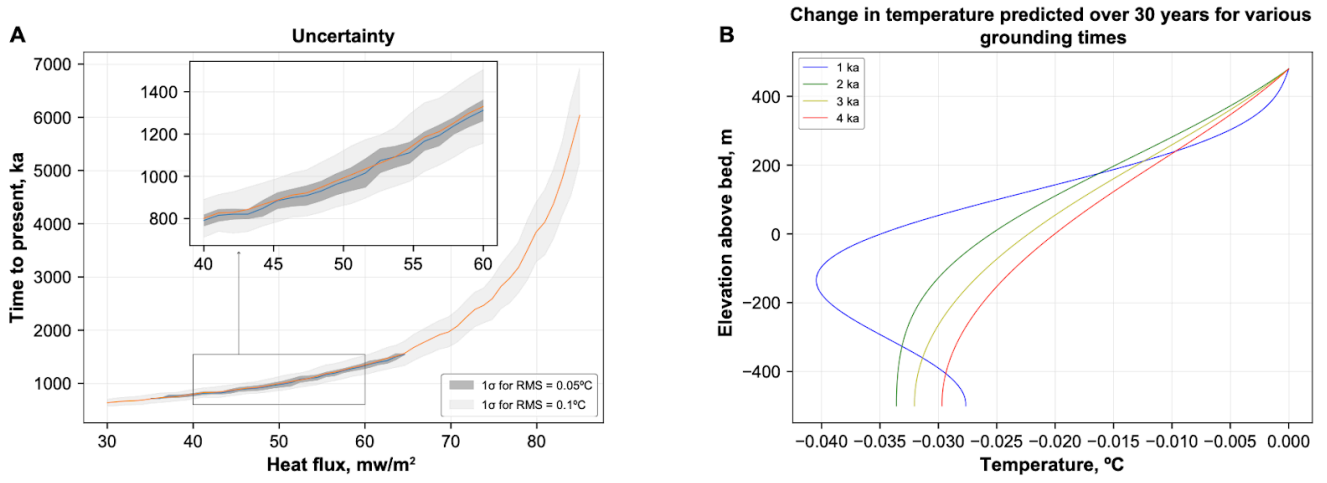


Fig. 7. Uncertainties associated with ice rise age inversion from temperature-depth profiles. A. Combinations of values of heat flux and timing of grounding used in the synthetic inversion experiment that yield 0.1º RMS (mean indicated by orange line with light gray band around it indicating standard deviation) and 0.05ºC RMS (mean indicated by blue line with dark gray band around it indicating standard deviation) misfit with 'measured' temperature profiles, respectively. B. Temperature profile difference curves showing potential detectability of grounding-induced changes in the lower part of the ice column by repetitive englacial thermometry (for a time interval between measurements of 30 years for four different timings of grounding).

Higher accuracy of englacial thermometry implies that the grounding-induced evolution of the temperature profile can be detected with two measurements separated by a few decades, which could potentially be utilised in previously drilled boreholes (Fig. 7B). In the case of Crary Ice Rise, a series of forward models shows that 30-year increase in the timing of grounding (e.g., difference between temperature profiles within 1030 and 1000 years old ice rise) may yield a difference in englacial temperatures that could be detected in the lower part of the ice column as well as the upper section of the underlying sediment/bedrock (Fig. 7B). Therefore, if a sufficiently deep borehole is drilled measurements through the underlying bedrock may provide useful additional constraints on both the heat flux and the timing of ice-rise formation. However, rapid attenuation of temperature differences with time since grounding implies that this method could only be applicable for ice rises that experienced recent grounding (i.e., less than 1 ka) assuming measurement uncertainties of 0.02ºC, as reported from a borehole drilled through the Dome Summit South in East Antarctica (Van Ommen et al., 1999). This loss of temporal resolution also implies increasing uncertainties associated with inferring the timing of grounding for ice rises that are older than approximately 4 k.y. (Fig. 5E). Additionally, our model does not account for the freezing of groundwater and its corresponding effect on cooling. A more sophisticated forward model that takes this into account would be required to investigate this effect in water-saturated bedrock or sediments with high porosity.

4.2. Inversion of time-variable forcings

The Bayesian inversion method presented here has potential use in providing information about other model parameters, including time-variable thickness, temperature and accumulation rates, as previously done in multiple locations in Greenland

https://doi.org/10.5194/egusphere-2022-236
Preprint. Discussion started: 23 May 2022

and Antarctica (e.g., Dahl-Jensen et al. 1998; 1999; Waddington, 2005; Orsi et al., 2012; Cuffey et al., 2016). Since the addition of temporal variability to external forcings significantly increases the dimensionality of the inversion problem, with associated exponential growth of the computation cost, here we focus largely on thermal effects of dynamic ice-rise evolution. However, our forward Monte-Carlo experiments show the variable impact of these forcings on englacial temperature distribution and their variability with ice thickness and depth within the ice column (Figs 5,6). For example, the effects of relatively small perturbations to accumulation rates are generally greater for thicker ice, whereas the opposite is true for timing of grounding and thickening rates, which play a more important role when ice is thin. Anomalies associated with changes to most parameters under consideration typically increase with depth (insets of Figs.5A-C and Figs. 6A-D), with the exception of surface temperatures, which exert a strong influence on the temperature-depth distribution in the upper part of the ice column (e.g., Dahl-Jensen et al. 1998). This implies that the upper part of the ice column is likely to contain a more recent and less diffuse record of past surface temperatures (e.g., Orsi et al., 2012).

Matsouka et al.'s (2015) Antarctic ice-rise inventory shows that ice rises rarely exceed 500 m in thickness, suggesting that these features may store a stronger signal of past dynamic changes and thickening history, whereas thick, permanently grounded ice domes may retain more information about accumulation and temperature histories and are thus more appropriate locations for paleoclimatic inferences from englacial temperature measurements (e.g., Dahl-Johnsen et al., 1999; Engelhardt, 2004; Orsi et al., 2012; Cuffey et al., 2016). Forward model experiments with different vertical velocity approximations show that the impact of vertical ice flow parametrization becomes more significant for thicker ice (Fig. 6A,D). Yet, previous studies from Greenland and Antarctica have also shown that analytical ice flow approximations from Dansgaard and Johnsen (1969) and Lliboutry (1979) cannot fully capture the nonlinearity of vertical velocity profiles in ice divide/ice rise setting, and phase-sensitive radar measurements can provide useful additional constraints on vertical ice-sheet velocities (e.g., Gillet-Chaulet et al., 2011; Kingslake et al., 2014). Therefore, integration of these techniques would help improve inferences of external forcings from borehole temperatures, in particular surface temperature and accumulation histories from deep boreholes.

4.3. Implications for choosing borehole drilling sites

Building on Bindscadler et al.'s (1990) foundational work and Neuhaus et al.'s (2020) more recent study, we have demonstrated how this method has potential as a useful dating technique that can be implemented at ice rises across Antarctica where direct geological sampling methods are inaccessible (e.g., Bentley et al., 2010; Spector et al., 2018). Integrating this technique with other methods, such as: (1) indirect estimates of timing of grounding from radar observations and modelling (e.g., Schroeder et al., 2014; Kingslake et al., 2016; Wearing and Kingslake, 2019), (2) parametrization of vertical velocities (Kingslake et al 2014), and (3) adoption of more tightly restricted, informative prior constraints using geochemical ice-core data, (e.g., for past temperature proxies; Cuffey et al., 2016), will allow for more accurate inferences of dynamic ice-rise/ice-sheet evolution and grounding line migration (e.g., Orsi et al 2012). Moreover, we have demonstrated

https://doi.org/10.5194/egusphere-2022-236
Preprint. Discussion started: 23 May 2022

an approach for better quantifying uncertainties in these inferences. Furthermore, borehole measurements through the upper tens of meters of underlying sediment/bedrock could place additional important constraints, both on the geothermal heat flux and ice-rise evolution. This technique could even provide insights into dynamic ice-sheet evolution if future boreholes are drilled through floating ice and sediment in the vicinity of the grounding line, in places where recent ungrounding has left a pronounced vertical temperature anomaly within both ice and sediment/bedrock columns.

Our results prompt the question of what characteristics make a location favourable for borehole drilling and measuring temperature-depth data within an ice rise. Due to the diffusive nature of the englacial thermal signal, and as synthetic data experiments have shown, the temporal resolution decreases with time since grounding (Figs. 3, 5B,E, 7A). Therefore, ages of ice rises that are over 4 k.yr. old may be difficult to determine accurately, subject to other parameters like heat flux and thickening rates. Therefore, areas that are located close to the present-day grounding line (and thus more likely to have been formed relatively recently) with well-constrained, low values of heat flux and low thickening rates could represent optimal locations for implementation of this method and could yield accurate (i.e., on the order of 10%) ice-rise age estimations. Kingslake et al (2018), Venturelli et al. (2020) and Neuhaus et al. (2020) have shown evidence of potential regrounding across large areas of West Antarctica. Preliminary investigations, juxtaposing these maps of potential grounding line migrations (Kingslake et al., 2018; Albrecht et al., 2020b; Fig. 1D) with Matsouka et al.'s (2015) ice-rise inventory (Figs.1A-D) and Martos et al. (2017) heat flux model shows that several ice rises, for example Korff Ice Rise (Kingslake et al., 2016), could prove to be optimal locations for application of this technique. Future work could systematically quantify the suitability of these locations following this approach.

5. Conclusions

In this paper, we combined Bayesian inversion and forward Monte-Carlo methods to make a comprehensive evaluation of uncertainties inherent in inferences of ice-rise dynamic evolution from temperature-depth profiles. Tested with both synthetic datasets and borehole temperature measurements from Crary Ice Rise, Ross Sea Embayment, our method explores the interplay of surface temperature, rates of accumulation and thickening, geothermal heat flux and parameterized vertical velocities. We show that depending on the accuracy of borehole thermometry, the same temperature profile (within the accuracy of measurements) may result from a range of forcing parameters, of which geothermal heat flux through underlying bedrock plays a particularly important role. The key implication is that careful model parametrization and evaluation of uncertainties are essential to infer dynamic ice-rise evolution from borehole thermometry. We highlight that uncertainties in inferred ice-formation time may increase significantly with ice-rise age. Accuracy of inversion relies on the low measurement uncertainties (i.e., <0.05ºC) and can be high (i.e., uncertainties <10%) for relatively young ice rises (i.e., formed <4 ka) that are grounded in areas where heat flux is low, and its value is well-constrained.

https://doi.org/10.5194/egusphere-2022-236
Preprint. Discussion started: 23 May 2022

Code availability. The code related to this article is available on-line at: https://github.com/sashamontelli/borehole_temperature_models/blob/master/Annotated%20temperature%20profile%20MCMC%20inversion.ipynb

Author contributions. A.M. co-designed this research, performed the analysis, and wrote the manuscript. J.K. co-designed this research and contributed to the writing and editing of the manuscript.

Competing interests. The authors declare that they have no conflict of interest.

Acknowledgements. A.M. is grateful to the Schmidt Science Fellowship for funding of this research. We thank Thorsten Albrecht for granting access to the PISM model outputs and Nicholas Holschuh for helpful discussions of temperature measurements from Crary Ice Rise.

Financial support. This research has been supported by the Schmidt Science Fellowship to A.M.

References

Albrecht, T., Winkelmann, R. and Levermann, A., 2020a. Glacial-cycle simulations of the Antarctic Ice Sheet with the Parallel Ice Sheet Model (PISM)–Part 1: Boundary conditions and climatic forcing. The Cryosphere, 14(2), 599-632.

Albrecht, T., Winkelmann, R. and Levermann, A., 2020b. Glacial-cycle simulations of the Antarctic Ice Sheet with the Parallel Ice Sheet Model (PISM)–Part 2: Parameter ensemble analysis. The Cryosphere, 14(2), 633-656.

Alley, R.B. and Koci, B.R., 1990. Recent warming in central Greenland?. Annals of Glaciology, 14, 6-8.

An, M. J., Wiens, D. A., Zhao, Y., Feng, M., Nyblade, A., Kanao, M., et al. (2015). Temperature, lithosphere-asthenosphere boundary, and heat flux beneath the Antarctic Plate inferred from seismic velocities. Journal of Geophysical Research: Solid Earth, 120, 8720–8742.

Begeman, C.B., Tulaczyk, S.M. and Fisher, A.T., 2017. Spatially variable geothermal heat flux in West Antarctica: evidence and implications. Geophysical Research Letters, 44(19), 9823-9832.

Bentley, M. J., C. J. Fogwill, A. M. Le Brocq, A. L. Hubbard, D. E. Sugden, T. J. Dunai, and S. P. Freeman (2010), Deglacial history of the West Antarctic Ice Sheet in the Weddell Sea Embayment: Constraints on past ice volume change, Geology, 38(5), 411–414.

Bindschadler, R.A., Roberts, E.P. and Iken, A., 1990. Age of Crary Ice Rise, Antarctica, determined from temperature-depth profiles. Annals of Glaciology, 14, 13-16.

Cuffey, K. M., Clow, G. D., Alley, R. B., Stuiver, M., Waddington, E. D., & Saltus, R. W. (1995). Large Arctic temperature change at the Wisconsin-Holocene glacial transition. Science, 270(5235), 455–458.

Cuffey, K.M., Clow, G.D., Steig, E.J., Buizert, C., Fudge, T.J., Koutnik, M., Waddington, E.D., Alley, R.B. and Severinghaus, J.P., 2016. Deglacial temperature history of West Antarctica. *Proceedings of the National Academy of Sciences*, *113*(50), 14249-14254.

Dahl-Jensen, D. and Johnsen, S.J., 1986. Paleotemperatures still exist in the Greenland ice sheet. Nature, 320(6059), 250-252.

Dahl-Jensen, D., Mosegaard, K., Gundestrup, N., Clow, G.D., Johnsen, S.J., Hansen, A.W. and Balling, N., 1998. Past temperatures directly from the Greenland ice sheet. Science, 282(5387), 268-271.

https://doi.org/10.5194/egusphere-2022-236
Preprint. Discussion started: 23 May 2022

Dahl-Jensen, D., Morgan, V.I. and Elcheikh, A., 1999. Monte Carlo inverse modelling of the Law Dome (Antarctica) temperature profile. Annals of Glaciology, 29, 145-150.

Dahl-Jensen, D., Gundestrup, N., Gogineni, S. P., & Miller, H. (2003). Basal melt at NorthGRIP modeled from borehole, ice-core and radio-echo sounder observations. Annals of Glaciology, 37(37), 207–212.

Determann, J. and Gerdes, R., 1994. Melting and freezing beneath ice shelves: Implications from a three-dimensional ocean-circulation model. Annals of Glaciology, 20, 413-419.

Engelhardt, H., 2004a. Thermal regime and dynamics of the West Antarctic ice sheet. Annals of Glaciology, 39, 85-92.

Engelhardt, H., 2004b. Ice temperature and high geothermal flux at Siple Dome, West Antarctica, from borehole measurements. Journal of Glaciology, 50(169), p.251-256.

Foreman-Mackey, D., Hogg, D.W., Lang, D. and Goodman, J., 2013. emcee: the MCMC hammer. Publications of the Astronomical Society of the Pacific, 125(925), p.306.

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), 375-393.

Fudge, T.J., Markle, B.R., Cuffey, K.M., Buizert, C., Taylor, K.C., Steig, E.J., Waddington, E.D., Conway, H. and Koutnik, M., 2016. Variable relationship between accumulation and temperature in West Antarctica for the past 31,000 years. Geophysical Research Letters, 43(8), 3795-3803.

Fudge, T.J., Biyani, S.C., Clemens-Sewall, D. and Hawley, R.L., 2019. Constraining geothermal flux at coastal domes of the Ross Ice Sheet, Antarctica. Geophysical Research Letters, 46(22), 13090-13098.

Gillet-Chaulet, F., R. C. A. Hindmarsh, H. F. Corr, E. C. King, and A. Jenkins (2011), In-situ quantification of ice rheology and direct measurement of the Raymond Effect at Summit, Greenland using a phase-sensitive radar, Geophys. Res. Lett., 38, L24503, doi:10.1029/2011GL049843.

Goodman, J. and Weare, J., 2010. Ensemble samplers with affine invariance. Communications in applied mathematics and computational science, 5(1), 65-80.

Holland, D.M. and Jenkins, A., 1999. Modeling thermodynamic ice–ocean interactions at the base of an ice shelf. Journal of Physical Oceanography, 29(8), 1787-1800.

Kingslake, J., Hindmarsh, R.C., Aðalgeirsdóttir, G., Conway, H., Corr, H.F., Gillet-Chaulet, F., Martín, C., King, E.C., Mulvaney, R. and Pritchard, H.D., 2014. Full-depth englacial vertical ice sheet velocities measured using phase-sensitive radar. Journal of Geophysical Research: Earth Surface, 119(12), 2604-2618.

Kingslake, J., Martín, C., Arthern, R.J., Corr, H.F. and King, E.C., 2016. Ice-flow reorganization in West Antarctica 2.5 kyr ago dated using radar-derived englacial flow velocities. Geophysical Research Letters, 43(17), 9103-9112.

Kingslake, J., Scherer, R.P., Albrecht, T., Coenen, J., Powell, R.D., Reese, R., Stansell, N.D., Tulaczyk, S., Wearing, M.G. and Whitehouse, P.L., 2018. Extensive retreat and re-advance of the West Antarctic Ice Sheet during the Holocene. Nature, 558(7710), 430-434.

Koci, B. and Bindschadler, R., 1989. Hot-water drilling on Crary ice rise, Antarctica. Annals of Glaciology, 12, 214-214.

https://doi.org/10.5194/egusphere-2022-236
Preprint. Discussion started: 23 May 2022

Ligtenberg, S.R.M., Van de Berg, W.J., Van den Broeke, M.R., Rae, J.G.L. and Van Meijgaard, E., 2013. Future surface mass balance of the Antarctic ice sheet and its influence on sea level change, simulated by a regional atmospheric climate model. Climate dynamics, 41(3-4), 867-884.

Lliboutry, L. A. (1979). A critical review of analytical approximate solutions for steady state velocities and temperature in cold ice sheets. Zeitschrift für Gletscherkunde und Glazialgeologie, 15(2), 135–148.

MacAyeal, D.R. and Thomas, R.H., 1980. Ice–shelf grounding: ice and bedrock temperature changes. Journal of Glaciology, 25(93), 397-400.

MacGregor, J.A., Winebrenner, D.P., Conway, H., Matsuoka, K., Mayewski, P.A., Clow, G.D., 2007. Modeling englacial radar attenuation at Siple Dome, West Antarctica, using ice chemistry and temperature data. J. Geophys. Res. Earth Surf. 112 (F3), F03008. http:// dx.doi.org/10.1029/2006JF000717.

Martin, P.J. and Sanderson, T.J.O., 1980. Morphology and dynamics of ice rises. Journal of Glaciology, 25(91), 33-46.

Martín, C., & Gudmundsson, G. H. (2012). Effects of nonlinear rheology, temperature and anisotropy on the relationship between age and depth at ice divides. The Cryosphere, 6(5), 1221–1229. https://doi.org/10.5194/tc-6-1221-2012.

Martos, Y.M., Catalán, M., Jordan, T.A., Golynsky, A., Golynsky, D., Eagles, G. and Vaughan, D.G., 2017. Heat flux distribution of Antarctica unveiled. Geophysical Research Letters, 44(22), 11-417.

Matsuoka, K., Hindmarsh, R.C., Moholdt, G., Bentley, M.J., Pritchard, H.D., Brown, J., Conway, H., Drews, R., Durand, G., Goldberg, D. and Hattermann, T., 2015. Antarctic ice rises and rumples: Their properties and significance for ice-sheet dynamics and evolution. Earth-science reviews, 150, 724-745.

Miles, K.E., Hubbard, B., Quincey, D.J., Miles, E.S., Sherpa, T.C., Rowan, A.V. and Doyle, S.H., 2018. Polythermal structure of a Himalayan debris-covered glacier revealed by borehole thermometry. Scientific reports, 8(1), 1-9.

Millero, F.J. 1978. Freezing point of sea water. Eighth Report of thecJoint Panel of Oceanographic Tables and Standards. Appendix 6. UXESCO Tech. Pap. lI//a). Sci., 28, 29-31.

Mosegaard, K. and Tarantola, A., 1995. Monte Carlo sampling of solutions to inverse problems. Journal of Geophysical Research: Solid Earth, 100(B7), 12431-12447.

Neuhaus, S.U., Tulaczyk, S.M., Stansell, N.D., Coenen, J.J., Scherer, R.P., Mikucki, J.A. and Powell, R.D., 2020. Did Holocene climate changes drive West Antarctic grounding line retreat and re-advance?. The Cryosphere Discussions, 1-30.

Odaka, T.E., Banihirwe, A., Eynard-Bontemps, G., Ponte, A., Maze, G., Paul, K., Baker, J. and Abernathey, R., 2019. The Pangeo Ecosystem: Interactive Computing Tools for the Geosciences: Benchmarking on HPC. In Tools and Techniques for High Performance Computing (190-204). Springer, Cham.

Orsi, A.J., Cornuelle, B.D. and Severinghaus, J.P., 2012. Little Ice Age cold interval in West Antarctica: evidence from borehole temperature at the West Antarctic Ice Sheet (WAIS) divide. Geophysical Research Letters, 39(9).

Roberts, J.L., Moy, A.D., van Ommen, T.D., Curran, M.A.J., Worby, A.P., Goodwin, I.D., Inoue, M., 2013. Borehole temperatures reveal a changed energy budget at Mill Island, East Antarctica, over recent decades. Cryosphere 7 (1), 263–273. http://dx.doi.org/10. 5194/tc-7-263-2013.

Robin, G.D.Q., 1955. Ice movement and temperature distribution in glaciers and ice sheets. Journal of Glaciology, 2(18), 523-532.

https://doi.org/10.5194/egusphere-2022-236
Preprint. Discussion started: 23 May 2022

Schroeder, D. M., Blankenship, D. D., Young, D. A., & Quartini, E. (2014). Evidence for elevated and spatially variable geothermal flux beneath the West Antarctic Ice Sheet. Proceedings of the National Academy of Sciences of the United States of America, 111(25), 9070–9072. https://doi.org/10.1073/pnas.1405184111

Seroussi, H., Ivins, E.R., Wiens, D.A. and Bondzio, J., 2017. Influence of a West Antarctic mantle plume on ice sheet basal conditions. Journal of Geophysical Research: Solid Earth, 122(9), 7127-7155.

Spector, P., Stone, J., Pollard, D., Hillebrand, T., Lewis, C. and Gombiner, J., 2018. West Antarctic sites for subglacial drilling to test for past ice-sheet collapse. The Cryosphere, 12(8), 2741-2757.

Talalay, P., Li, Y., Augustin, L., Clow, G.D., Hong, J., Lefebvre, E., Markov, A., Motoyama, H. and Ritz, C., 2020. Geothermal heat flux from measured temperature profiles in deep ice boreholes in Antarctica. The Cryosphere, 14(11), 4021-4037.

Van Ommen, T.D., Morgan, V.I., Jacka, T.H., Woon, S. and Elcheikh, A., 1999. Near-surface temperatures in the Dome Summit South (Law Dome, East Antarctica) borehole. Annals of Glaciology, 29, 141-144.

Venturelli, R.A., Siegfried, M.R., Roush, K.A., Li, W., Burnett, J., Zook, R., Fricker, H.A., Priscu, J.C., Leventer, A. and Rosenheim, B.E., 2020. Mid-Holocene grounding line retreat and readvance at Whillans Ice Stream, West Antarctica. *Geophysical Research Letters*, *47*(15), p.e2020GL088476.

Waddington, E.D., Conway, H., Steig, E.J., Alley, R.B., Brook, E.J., Taylor, K.C. and White, J.W.C., 2005. Decoding the dipstick: thickness of Siple Dome, West Antarctica, at the last glacial maximum. Geology, 33(4), 281-284.

Wearing, M.G. and Kingslake, J., 2019. Holocene Formation of Henry Ice Rise, West Antarctica, Inferred From Ice-Penetrating Radar. Journal of Geophysical Research: Earth Surface, 124(8), 2224-2240.

White-Gaynor, A. L., Nyblade, A. A., Aster, R. C., Wiens, D. A., Bromirski, P. D., Gerstoft, P., et al. (2019). Heterogeneous upper mantle structure beneath the Ross Sea Embayment and Marie Byrd Land, West Antarctica, revealed by P-wave tomography. Earth and Planetary Science Letters, 513, 40–50.

Yang, J.W., Han, Y., Orsi, A.J., Kim, S.J., Han, H., Ryu, Y., Jang, Y., Moon, J., Choi, T., Hur, S.D. and Ahn, J., 2018. Surface temperature in the twentieth century at the Styx Glacier, northern Victoria Land, Antarctica, from borehole thermometry. Geophysical Research Letters, 45(18), 9834-9842.