

1 **Joint Inversion for Surface Accumulation and**
2 **Geothermal Flux from Ice-Penetrating Radar**
3 **Observations at Dome A, East Antarctica. Part II: Ice**
4 **Sheet State and Geophysical Analysis**

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8 **Key Points:**

- 9 • Our model matches the observed water and freeze-on locations, predicts new ar-
10 eas to look for water, and estimates freeze-on volume.
11 • Our model predicts that ice up to 1.5 Ma suitable for coring may be found un-
12 der the divide, assuming divide stability in the geologic past.
13 • Our geothermal flux estimate is higher than most previous estimates, reflecting
14 the variability in geothermal flux, even in old cratons.

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Abstract

Dome A is the peak of the East Antarctic Ice Sheet (EAIS), underlain by the rugged Gamburtsev Subglacial Mountains (GSM). The rugged basal topography produces a complex hydrological system featuring basal melt, water transport and storage, and freeze-on. In a companion paper, we used an inverse model to infer the spatial distributions of geothermal flux and accumulation rate that best fit a variety of observational constraints. Here, we present and analyze the best-fit state of the ice sheet in detail. Our model agrees well with the observed water bodies and freeze-on structures, while also predicting a significant amount of unobserved water and suggesting a change in stratigraphic interpretation that reduces the volume of the freeze-on units. We predict that a weak Raymond effect underneath the ice divide has been mostly masked by the high-amplitude variability in the layers produced by draping over subglacial topography. Our model stratigraphy agrees well with observations, and we predict- assuming that the ice divide has been stable over time- that there will be two distinct patches of ice older than 1 Ma suitable for ice coring underneath the divide. Finally, our geothermal flux estimate is substantially higher than previous estimates for this region. Correcting for the bias induced by unresolved narrow valleys still leaves our result in the high end of past estimates, with substantial local anomalies that are hotter still. Fundamentally, the observational evidence of a complex basal hydrological system is inconsistent with a simple picture of a uniformly cold East Antarctic craton.

Plain Language Summary

In a companion paper, we combined a model with observations to figure out the best-fit maps of geothermal heat flow and snowfall rate in the highest and coldest part of Antarctica, Dome A. In this paper, we analyze the best-fit model in detail. The observations show liquid water moving around underneath the ice sheet and traveling from melting regions to freezing regions. Our model does a good job of matching those observations, while also suggesting new locations where water underneath the ice may be found and also suggesting that the refrozen ice may be smaller than previously believed. Our model predicts that ice more than one million years old, which would be very useful for collecting ice cores, might be found intact and in order within a narrow region underneath the ice ridge. However, if the ice ridge has moved over time then our model might be wrong about this. Our best-fit map of geothermal heat flow is hotter than previous estimates, even after we try to correct it to account for a possible source of error. This result emphasizes the fact that even old cold areas of Earth's crust can have local areas that are hotter.

1 Introduction

Dome A is the highest and coldest part of the East Antarctic Ice Sheet (EAIS), with a maximum surface elevation of nearly 4100 m and an annual average surface temperature of roughly -60 °C (Fretwell et al., 2013; Comiso, 2000). It is underlain by the Gamburtsev Subglacial Mountains (GSM), a rugged mountain range with ~ 2500 m of relief that is completely covered by the ice sheet (Fig 1). The GSM is believed to have formed by rejuvenation of a Proterozoic crustal root during Permian and Cretaceous rifting (Ferraccioli et al., 2011). Despite the fact that the GSM have been covered by the EAIS continuously since 34 ma, their topography is still dominated by a pre-glacial fluvial valley network modified slightly by valley glaciers in the very early stages of Antarctic glaciation (Bo et al., 2009; Rose et al., 2013). The modern ice divide is located roughly over the northern spur of the GSM, while the southern spur of the mountains is associated with a number of locations where the hydraulic potential would force subglacial water, if present, to flow uphill towards the thinner ice located over the mountain peaks (Fig 1).

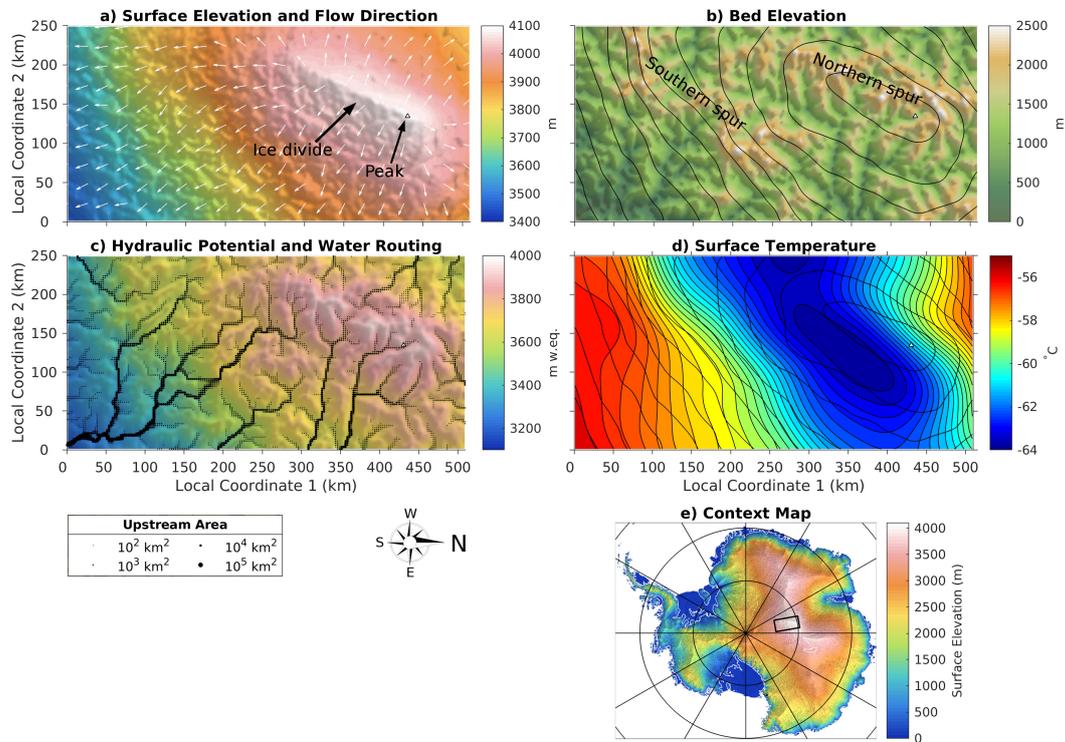


Figure 1. Glaciological setting and boundary conditions for the model. a) Hillshaded ice sheet surface elevation, with flow direction vectors from the smoothed surface overlain; b) hillshaded bed elevation, with 50 m surface contours overlain; c) hillshaded hydraulic potential, with the results of a water routing algorithm overlain; d) surface temperature, with 50 m contours of surface elevation; and e) context map showing model domain (black rectangle) overlain on a plot of hillshaded surface elevation for all of Antarctica. Triangles show Kunlun station near the peak of Dome A. All hillshading performed using two perpendicular light sources at the top and right of the page. Compass rose shows orientation of model domain relative to true north. All subsequent map figures retain this orientation, as well as the local coordinate system aligned with the rectangular domain. Text labels in (a) and (b) indicate geographic features referred to later in the text. Surface and bed elevations are from BEDMAP2 (Fretwell et al., 2013); hydraulic potential has been computed from those quantities using densities of $\rho_i = 917$ and $\rho_w = 1000 \text{ kg m}^{-3}$, followed by filling of closed basins; and surface temperature is derived from (Comiso, 2000) via the ALBMAP_v1 compilation (Le Brocq et al., 2010), followed by a uniform downward adjustment of $5.75 \text{ }^\circ\text{C}$ to correct for the difference between modern temperatures and long-term average temperatures, with the correction derived based on the EPICA Dome C ice core (Jouzel et al., 2007).

65 During the 2008-2009 austral summer field season, the Antarctica's GAmurtsev
66 Province (AGAP) expedition surveyed Dome A and the GSM with a suite of airborne
67 geophysics instruments at a flight line spacing of 5 km with 33 km perpendicular tie lines.
68 In addition to revealing the basic contours of the mountainous subglacial topography (Fig
69 1b), the ice-penetrating radar data collected during this survey also revealed large plumes
70 of refrozen (accreted) ice being added to the ice sheet base and carried downstream in
71 the ice flow (Bell et al., 2011). The source regions for the most prominent of these freeze-
72 on plumes are associated with the termini of organized networks of small subglacial wa-
73 ter bodies in places where the hydraulic potential forces water to flow uphill (Wolovick
74 et al., 2013). This freeze-on process is believed to protect the mountain peaks from ero-
75 sion by the subglacial water that is found in the deep valleys, thus preserving the GSM
76 against erosion (Creyts et al., 2014). However, much about the ice sheet state in the GSM
77 remains unknown, including critical boundary conditions like the geothermal flux and
78 the long-term accumulation rate, as well as the aspects of the ice sheet state that depend
79 on those boundary conditions, such as the thermal structure of the ice sheet and bed,
80 the basal melt rate and water flux, the ice flow and deformation fields, and the age-depth
81 scale.

82 In a companion paper (Wolovick et al., in review) we used a formal inverse model
83 to assimilate the AGAP radar observations of subglacial water, freeze-on, and internal
84 layers, in order to solve for best-fit geothermal flux and long-term average accumulation
85 rate fields for Dome A. We also used the continental-scale geothermal flux estimate from
86 Martos et al. (2017), which is based on the AGAP aeromagnetic observations for the part
87 of their model within our domain, as an additional constraint. Our forward model com-
88 puted a self-consistent coupled steady state between the ice sheet flow field, thermal struc-
89 ture, and basal hydrological system. Our basal hydrology model included both melt
90 and freeze-on, allowing us to track water transport from source to sink while conserv-
91 ing mass and energy at the bed. We also computed the age structure of the ice sheet and
92 the thickness of freeze-on ice. Our inverse model minimized a compound misfit function
93 accounting for all of the available constraints using an evolutionary algorithm followed
94 by local optimization. Readers interested in exploring the details of the forward model,
95 data constraints, and inverse model are encouraged to read the companion paper. We
96 show the best-fit results of the inversion and their uncertainty fields in Figure 2.

97 Here, we examine the best-fit state of the ice sheet in detail. We explore the ther-
98 mal structure of the ice sheet and bed, and show how our model correctly captures the
99 placement of the prominent observed water networks and freeze-on plumes, while also
100 predicting the existence of large wet-based areas in the main trunk valleys of the GSM
101 for which the radar evidence of water is weak or ambiguous. Our model also suggests
102 several reasonable changes in interpretation relative to our earlier published works. We
103 also explore the flow, deformation, and age structure of the ice sheet in detail, showing
104 how our model predicts that a weak version of the Raymond effect (Raymond, 1983) ought
105 to be present underneath the ice divide. This weak Raymond effect creates conditions
106 in which extremely old ice- up to ~ 1.5 Ma- may be found in stratigraphic order for ice
107 coring. We discuss the implications of our best-fit accumulation rate pattern for divide
108 stability and ice coring. We then discuss how our model may be limited by the need to
109 use gridded topography, which necessarily removes short-wavelength variability from the
110 bed, and we estimate the bias introduced by this limitation in our best-fit estimate of
111 geothermal flux. Finally, we compare our best-fit and bias-corrected geothermal flux es-
112 timates with other estimates of geothermal flux in the region, and we discuss the impli-
113 cations of those differences for East Antarctic geology.

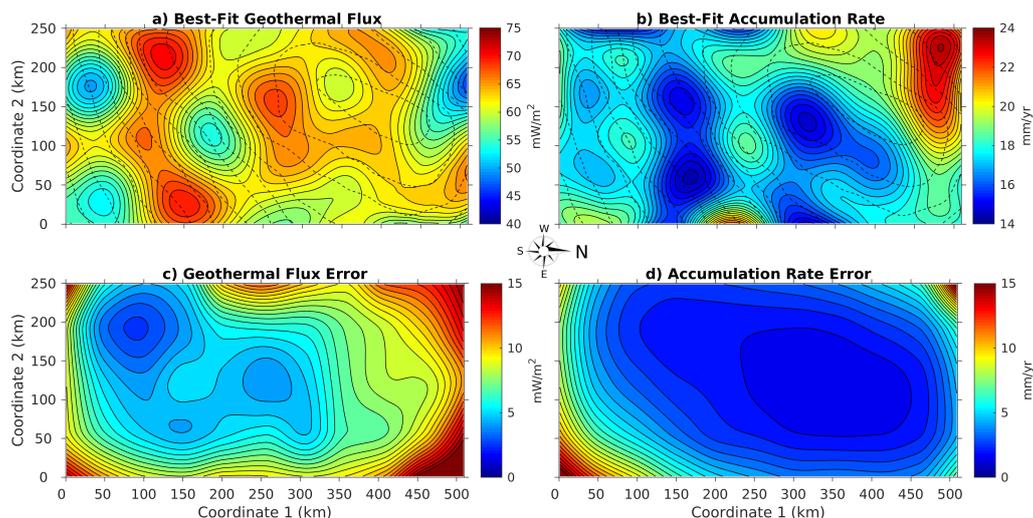


Figure 2. Summary of the results of the inversion described in the companion paper (Wolovick et al., in review). Top row shows best-fit fields for a) geothermal flux and b) accumulation rate, while bottom row (c and d) shows their respective error estimates. Dashed lines in top row are 50m surface elevation contours.

114 2 Results

115 2.1 Thermal Structure, Basal Hydrology, and Associated Data Types

116 The thermal structure of the ice sheet is dominated by conduction, with nearly linear
 117 profiles from the bed to the surface (Fig 3a). Because the accumulation rate is so low
 118 (Fig 2b), the thermal profiles are only slightly curved by downward advection. Horizontal
 119 variability in the thermal state of the ice sheet is strongly influenced by the preglacial
 120 valley network of the GSM, with deviations from the simple valley structure where water
 121 is forced up and over subglacial ridges (Fig 3, c.f. Fig 1c). To first order, the basal
 122 temperature of the ice sheet is determined by basal topography, with ice at the melting
 123 point in deep valleys and as much as 25°C below the melting point on the peaks (Fig
 124 3b, c.f. Fig 1b). While basal temperature cannot rise above the melting point, this pattern
 125 is continued in the basal melt rate, with the highest melt rates found in the deepest
 126 parts of the valleys (Fig 3c). However, the uphill flow of water under the influence
 127 of ice overburden pressure feeds freeze-on units and produces important deviations from
 128 the simple topographic picture. At five of the major networks identified by (Wolovick
 129 et al., 2013), identified by the letters A-E, water flows uphill through the valley network
 130 and over topographic peaks (Fig 3d). As the water flows uphill, it freezes and releases
 131 latent heat (Fig 3c), warming the ice and maintaining the ice base at the melting point
 132 through otherwise cold-bedded regions (Fig 3b). In some cases (networks A and C), the
 133 water is used up by freezing and the network terminates, while in the others, freeze-on
 134 only consumes a fraction of the available water and the remainder continues downstream,
 135 eventually joining the water produced by local melting in the southeast corner of the domain.
 136 The southeast corner of the domain has few observations of basal water, but the deep ice
 137 there ensures that conditions are warm-bedded in the best-fit model. Overall,
 138 the model predicts a large amount of unobserved water in the deep valleys of the GSM.
 139 Either drilling to the bed or more advanced radar processing and analysis techniques would
 140 be needed to confirm whether these valleys do indeed contain unponded water.

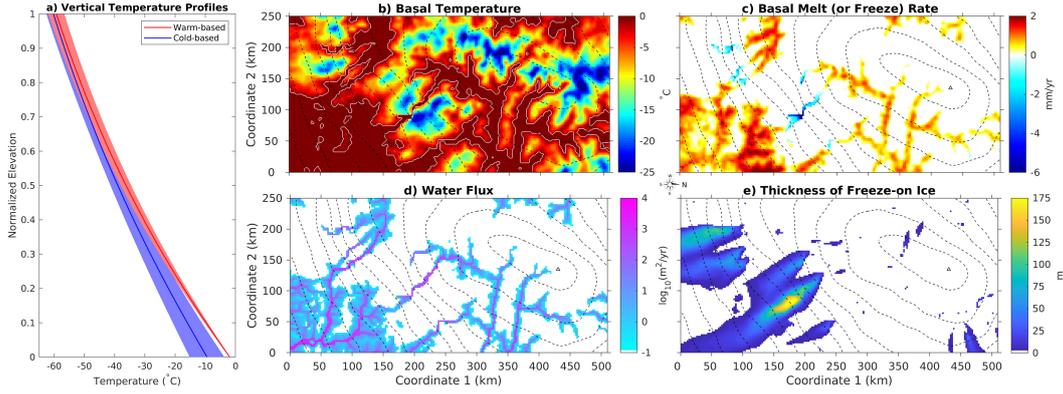


Figure 3. Thermal structure in the best-fit model. Panel a) shows vertical temperature profiles through the ice column. Profiles represent horizontal averages for warm-based and cold-based regions separately, and shading represents the standard deviation. Map plots show b) basal temperature, c) basal melt rate, d) basal water flux, and e) freeze-on thickness. Basal temperature in (b) has been corrected for the pressure dependence of the melting point and white line encloses the region at the melting point. Dashed lines show 50 m surface contours and triangle shows Kunlun station near the peak of Dome A. Letters A-E indicate major water networks and freeze-on plumes in observations.

141 Comparing our best-fit model results with the observational constraints allows us
 142 to verify the quality of our predicted thermal and hydrological structure. All of the major
 143 observed water networks are captured by the model (Fig 4b) with the exception of
 144 the uppermost reaches of network E (Fig 4c). The major freeze-on plumes are well-represented
 145 as well (Fig 5b), with most of the model freeze-on thickness falling in the “correctly lo-
 146 cated” category (Fig 5c). In fact, most of the incorrectly located model freeze-on appears
 147 to be edge effects around the margins of the correctly located freeze-on (Fig 5d). Whether
 148 because of artificial diffusion in the model advection scheme, or because of minor wa-
 149 ter routing errors in the freeze-on source regions, the model freeze-on plumes often ex-
 150 tend more than one grid cell beyond the borders of the observed freeze-on plumes, lead-
 151 ing to a penalty in the misfit function, most prominently at plume D. The model does
 152 not produce much freeze-on thickness in regions completely disconnected from the ob-
 153 served freeze-on plumes, giving us confidence that it is accurately representing the ther-
 154 mal state of the ice sheet as represented by the water and freeze-on observations. In ad-
 155 dition, both the model produced by the evolutionary algorithm and the locally optimized
 156 model predict very similar distributions of basal temperature and freeze-on (not shown),
 157 giving us confidence in the robustness of our results.

158 Consistent with the results of our preliminary forward model tests, the freeze-on
 159 packages produced by these water networks in our model are smaller than their observed
 160 counterparts (Bell et al., 2011), generally achieving maximum thickness between 50-100
 161 m, with only the freeze-on package produced by network D reaching 175 m (Fig 3d). How-
 162 ever, the observed freeze-on unit at that location had a maximum thickness of nearly 1000m
 163 and a mean thickness of 400-500m. The observed freeze-on units are universally higher
 164 in the ice column than their modeled counterparts, despite the fact that the inversion
 165 tried to maximize freeze-on volume without any penalty for producing too much freeze-
 166 on at the observed locations. The inability of the model to produce larger freeze-on
 167 units is a fundamental consequence of the large latent heat of ice; it is simply not possible to
 168 remove latent heat fast enough to produce larger units in a steady-state model.

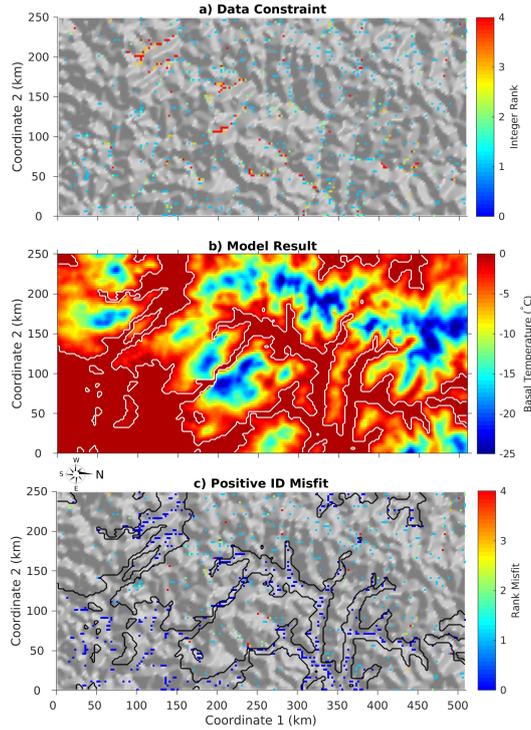


Figure 4. Best-fit model compared to the water constraint. a) Water observations overlain on hillshaded bed topography; b) model result for basal temperature relative to the melting point, with white contour enclosing region at the melting point; c) model misfit to the data plotted on top of hillshaded bed topography. Plot shows the severity of misfit for each observation; where the observed water falls within the model water (given by the black contour), the misfit is zero; where the observed water is not captured by the model, the misfit is given by the original confidence rank of the water observation.

169 We therefore interpret this discrepancy as evidence that additional processes inter-
 170 tervene in reality to uplift freeze-on ice above the basal layer and into the mid-depths
 171 of the ice column. This represents a change in interpretation from Bell et al. (2011). They
 172 interpreted the radargrams to mean that the transparent ice below the observed freeze-
 173 on reflectors was composed of freeze-on ice; ie, the observed reflector represented the up-
 174 per surface of an otherwise transparent package. We interpret the observed reflectors as
 175 representing the entirety of the freeze-on ice; ie, we interpret the observed reflector as
 176 representing volume scattering from impurities or sediment within the refrozen ice, with
 177 transparent and severely deformed meteoric ice below it. Additional processes not in-
 178 cluded in our model, such as time-variable basal slip (Wolovick et al., 2014; Wolovick &
 179 Creyts, 2016), small-scale rheological heterogeneity (NEEM Community Members, 2013),
 180 or complex flow patterns associated with the rugged topography of the GSM like viscous
 181 ice eddies (Meyer & Creyts, 2017), are then required to uplift the freeze-on ice several
 182 hundred meters above the bed and produce large englacial folds, where those are present.

183 2.2 Ice Sheet Flow Field

184 The vertical structure of the ice sheet flow field is marked by a strong concentra-
 185 tion of shear in the lower 20-30% of the ice column, and nearly uniform velocity above
 186 that (Fig 6a). Though our model does not include basal sliding, the concentration of shear

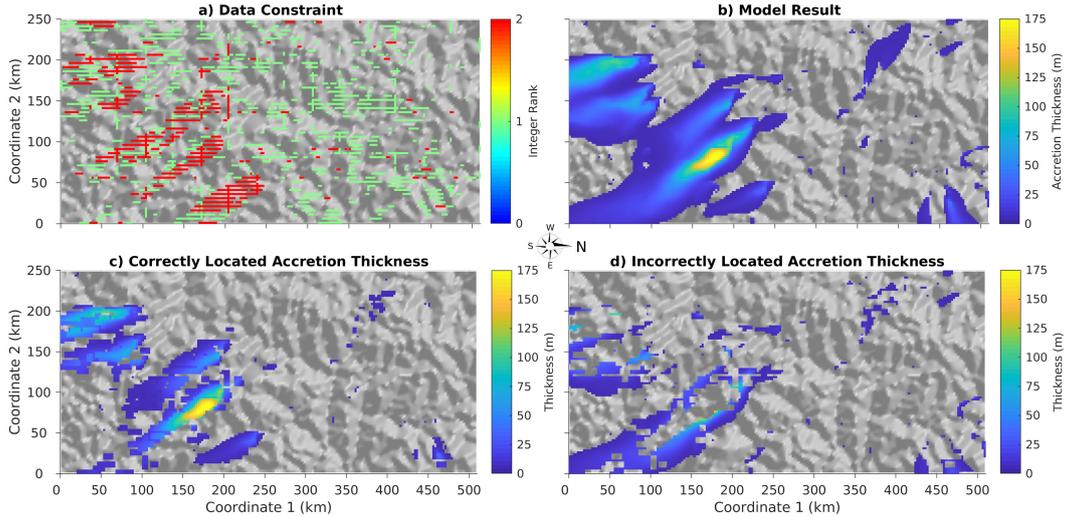


Figure 5. Best-fit model compared to the freeze-on constraint. a) Freeze-on observations overlain on hillshaded bed topography. b) Model result for freeze-on thickness. The color scale has been truncated where freeze-on thickness is less than 1 m and hillshaded bed topography is shown underneath. c) Correctly located freeze-on thickness, scaled by one half of the confidence rank of the observations at that location. The spatial integral of this field gives the “good” freeze-on volume in the positive ID misfit (Wolovick et al., (in review), Eq 15). d) Incorrectly located freeze-on thickness. The integral of this field gives the “bad” freeze-on volume in the negative ID misfit (Wolovick et al., (in review), Eq 16).

187 near the bed nonetheless increases in warm-bedded regions, as the increase in basal tem-
 188 perature also increases the rheological contrast with the colder and stiffer ice above (Fig
 189 6a). In most of the domain, the upper 70-80% of the ice column is nearly rigid, moving
 190 downstream as a coherent unit while the warm soft layers underneath deform to accom-
 191 modate the complex basal topography. This pattern is only broken in a narrow region
 192 near the divide, where a weak version of the Raymond effect (Raymond, 1983) causes
 193 deformation to spread out more evenly through the ice column (Fig 6a).

194 The horizontal structure of the ice sheet flow field follows from the governing as-
 195 sumptions of the balance flux algorithm we used to determine column-average velocity.
 196 Flow is slowest around the dome and increases with distance downstream, reaching a max-
 197 imum of just over 3 ma^{-1} in the southeast corner of the domain (Fig 6a). While these
 198 velocities are very slow compared with rapidly sliding ice streams and outlet glaciers, the
 199 flow is nonetheless concentrated in narrow fingers of (relatively) faster-moving ice sur-
 200 rounded by slower-flowing areas. If this concentrated structure is real, it would continue
 201 the trend of “patterned enhanced flow” observed by Rignot et al. (2011) in areas of East
 202 Antarctica downstream of our model domain. The patterned flow that they observed was
 203 in ice moving several tens of ma^{-1} , about an order of magnitude faster than in our do-
 204 main, but still well below the threshold of streaming flow. They attributed the patterned
 205 flow to basal sliding occurring even in slower-moving flank regions of the ice sheet, but
 206 our results suggest that basal slip is not necessary for flow to concentrate in tendril-like
 207 patterns. However, care should be taken in interpreting this aspect of our results. Our
 208 model uses a balance flux algorithm to compute ice flow, and balance flux algorithms
 209 make the simplifying assumption that the flow vector points exactly downhill (at least
 210 with respect to the smoothed ice surface). In reality there may be small angular devi-
 211 ations away from the perfect downhill direction, and spatial variations in the degree of

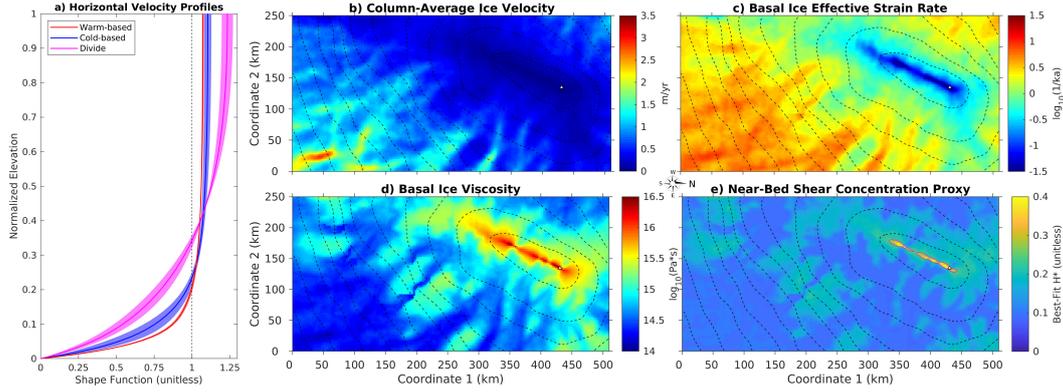


Figure 6. Flow structure in the best-fit model. Panel a) shows vertical profiles of the horizontal velocity shape function (ie, \hat{u} from Wolovick et al., (in review), Eq 12). Profiles represent horizontal averages for warm-based, cold-based, and near-divide regions separately, and shading represents the standard deviation. Map plots show b) column-average velocity (\bar{u}), c) effective strain rate ($\dot{\epsilon}_E$) of the basal ice, d) basal ice viscosity (μ), and e) dimensionless corner elevation for best-fit D-J model (Dansgaard & Johnson, 1969). The corner elevation quantifies the degree to which shear is concentrated near the bed, with smaller values indicating that shear is more concentrated. The “near-divide” profiles in (a) are defined by corner elevation greater than 0.25, corresponding to areas within the purple contour in (e). Dashed lines show 50 m surface contours and triangle shows Kunlun station near the peak of Dome A.

212 longitudinal coupling will also produce variability in the appropriate horizontal averaging
 213 scale over which “downhill” ought to be defined. These small deviations from the as-
 214 sumed downhill direction can become magnified during the balance flux integration, po-
 215 tentially producing erroneous structure in the model velocity field.

216 Regardless of whether the tendrils of relatively faster flow away from the dome are
 217 accurate, the slow flow and weak Raymond effect near the dome are a robust result of
 218 our model. The ice surface forms a divide ridge for about 100km south and to the east
 219 of the peak of Dome A, a geometry known to give rise to the Raymond effect. The Ray-
 220 mond effect (Raymond, 1983) is the phenomenon whereby the non-Newtonian rheology
 221 of ice combines with the very low rates of vertical shear underneath a divide to produce
 222 a core of stiff ice near the bed, thus resisting vertical thinning and causing the upward-
 223 ping of internal layers. The Raymond effect is strongest in an isothermal ice sheet which
 224 is frozen to a flat bed. In Dome A, the Raymond effect is weakened by topographic vari-
 225 ability in the bed and by the extreme thermal contrast between the bed and the surface.
 226 This thermal contrast- approximately 60°C- ensures that the ice near the bed remains
 227 softer than the ice near the surface regardless of the strain rate, while the large-amplitude
 228 variability in the subglacial mountains ensures that the dominant signal in the internal
 229 layers is always going to be draping over the topography. Nonetheless, strain rates are
 230 substantially lower at the bed underneath the divide than they are at the bed on the flanks
 231 (Fig 6b), and this leads the basal ice to be stiffer there than it is away from the divide
 232 (Fig 6c), producing more distributed shear within the ice column (Fig 6a). We can quan-
 233 tify the impact that this (relative) basal stiffening has on the englacial velocity distri-
 234 bution by fitting a best-fit Dansgaard-Johnson (D-J) model (Dansgaard & Johnson, 1969)
 235 to the shape function for horizontal velocity. A D-J model approximates the horizon-
 236 tal velocity field with a piecewise linear function of depth, where the velocity increases
 237 linearly from the bed up to a corner elevation H^* , and is constant above that. The best-
 238 fit value of H^* thus provides a convenient way to quantify the degree to which shear is

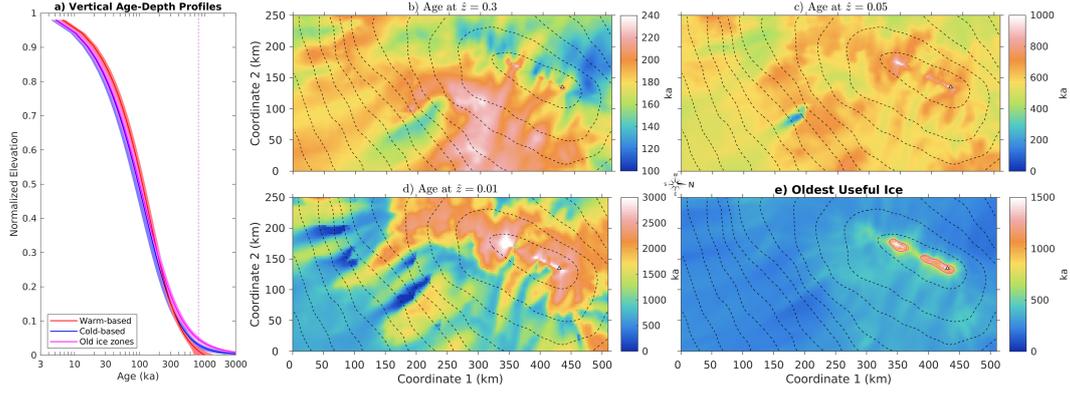


Figure 7. Age structure in the best-fit model. Panel a) shows vertical profiles of the age-depth scale. Profiles represent horizontal averages for warm-based, cold-based, and old-ice regions separately, and shading represents the standard deviation. Mean and standard deviation were both computed on a logarithmic scale, and only meteoric ice was included in the calculation. Map plots show ice age at normalized elevations of b) 30% ice thickness, c) 5% ice thickness, d) 1% ice thickness, and d) oldest useful ice for paleoclimate records. The oldest useful ice is computed using an accumulated shear steepening function that measures the likelihood that small stratigraphic perturbations will overturn, as well as a maximum temporal smoothing due to isotopic diffusion of 10 ka. Note changes in color scale between each map panel. “Old-ice regions” in panel (a) defined by areas with oldest useful ice greater than 800 ka, marked by the red contour in (e). The 800 ka threshold was chosen to correspond to the oldest ice in the EPICA Dome C ice core (Jouzel et al., 2007). Dashed lines show 50 m surface contours and triangle shows Kunlun station near the peak of Dome A.

239 concentrated near the bed. While the best-fit H^* is less than 20% of the ice thickness
 240 in the vast majority of the domain, we can see it rise to about 40% of ice thickness
 241 underneath the divide (Fig 6a,e), reflecting a flow regime in which shear is less concentrated
 242 near the bed, a hallmark of the Raymond effect. As we will show next, this weak Ray-
 243 mond effect has profound consequences for the age structure of the ice sheet.

244 2.3 Age Structure and Internal Layers

245 The age-depth profile of the ice sheet shows relatively weak spatial variability at
 246 mid-depths, with profiles that increasingly diverge near the bed (Fig 7a). If we take a
 247 horizontal slice at 30% of the ice thickness (ie, roughly at the top of the basal shear layer;
 248 c.f. Fig 6a), we see that the age structure of the ice sheet is dominated by the twin in-
 249 fluences of basal topography and horizontal advection (Fig 7b). In broad subglacial val-
 250 leys, the characteristic vertical strain rate (defined by the ratio of accumulation rate to
 251 ice thickness) is smaller, producing older ice at similar positions within the ice column.
 252 Conversely, subglacial peaks produce a larger characteristic strain rate and younger ice.
 253 This phenomenon can also be seen in the way that warm-bedded regions have older ice
 254 than cold-bedded regions at mid-depths in Fig 7a. However, this effect of the local ver-
 255 tical strain rate is modified by the presence of horizontal advection: downstream of promi-
 256 nent peaks, especially around (200,100), we can see streaks of younger ice, while promi-
 257 nent valleys produce downstream streaks of older ice. The greatest degree of short-wavelength
 258 spatial variability in the mid-depth layers can be seen underneath the divide, where slow
 259 flow diminishes the importance of horizontal advection, while away from the divide the
 260 age structure of the ice sheet is much more smoothed out. However, the situation reverses

261 near the bed. At normalized elevations of 5% and 1% (corresponding to average eleva-
 262 tions of 125 and 25 m above the bed, respectively), the youngest ice is found in the val-
 263 leys and the oldest ice on the peaks (Fig 7c,d). This is due to basal melting destroying
 264 older ice within the valleys, and causing the age-depth profiles to intersect the bed at
 265 finite age (Fig 7a). In addition, very young ice can be found above the freeze-on regions
 266 (Fig 7c,d; c.f. Fig 3e), where the “age” in this case represents time since freeze-on rather
 267 than time since surface deposition. The oldest ice is found in a thin region underneath
 268 the divide corresponding to the weak Raymond effect discussed in the previous section.
 269 At 1% elevation, ice older than 2 Ma is widespread and ice older than 3 Ma can be found
 270 as well (Fig 7d). The oldest ice is found underneath the divide, with a gap correspond-
 271 ing to a basal valley in which basal melt occurs (Fig 7d, c.f. Fig 3b,c).

272 However, just because extremely old ice is present does not mean that it is neces-
 273 sarily useful to ice coring. To obtain a useful climate record from an ice core, the ice must
 274 both be in stratigraphic order and it must not have been thinned so much that useful
 275 temporal resolution has been lost. To estimate the oldest useful ice in the model, we used
 276 thresholds reflecting both of these criteria. For our stratigraphic continuity threshold,
 277 we know that vertical shear increases the likelihood that small stratigraphic perturba-
 278 tions will steepen and overturn (Waddington et al., 2001; Jacobson & Waddington, 2004,
 279 2005); while we cannot resolve these small perturbations in our large-scale model, we can
 280 resolve the vertical shear, so we computed an accumulated shear steepening index to es-
 281 timate the likelihood of small-scale overturning. This steepening index can be thought
 282 of as finite-amplitude rotation in a vertical plane; beginning from zero when the ice is
 283 deposited at the surface, it accumulates along particle paths and grows most rapidly when
 284 vertical shear exceeds vertical thinning. Because we cannot resolve small-scale layer over-
 285 turn in our model, we do not know a priori what value of the steepening index to use as
 286 our threshold; so we empirically estimate a threshold value by comparing our model to
 287 the observed height of the echo-free zone (EFZ), the region near the bed in radar echograms
 288 where smooth continuous stratigraphy is no longer observed. The observed EFZ varies
 289 between 4% of the ice thickness under the divide (where the picks may be too high owing
 290 to the difficulty in detecting continuous layers close to a variable bed reflector) to
 291 a maximum of 38% of the ice thickness, with a mean value of 18% of the ice thickness.
 292 The best-fit threshold produces an RMS misfit with respect to the observed EFZ of 4%
 293 of the ice thickness, which we consider to be acceptable. For our layer thickness thresh-
 294 old, we assumed that the limiting factor would be isotopic diffusion rather than analyt-
 295 ical equipment, on the argument that technology can always improve the spatial reso-
 296 lution possible at ice core laboratories, but that no technological improvements can re-
 297 verse the loss of information due to diffusion. Many authors have computed isotopic dif-
 298 fusion using a simple $\sqrt{\kappa t}$ scaling, however the diffusion coefficient κ is temperature-dependent
 299 and varies by over three orders of magnitude as the ice layers descend from the surface
 300 to the bed. We therefore computed the characteristic spreading length of an impulsive
 301 isotopic perturbation as follows,

$$\frac{D\sigma}{Dt} = \frac{\kappa}{\sigma}, \quad (1)$$

302 where σ is the characteristic spreading length perpendicular to the layers, $\frac{D}{Dt}$ is the
 303 material derivative, and κ is an Arrhenius function of temperature. The boundary con-
 304 ditions are that σ is zero at the surface and bed. To derive the above equation, we as-
 305 sumed that an initial delta-function isotopic perturbation would spread out perpendic-
 306 ular to the locally planar layers with a Gaussian shape with a standard deviation given
 307 by σ . We solved for the steady-state σ field using the same advection solver used for the
 308 rest of the model, computed the layer thickness directly from the vertical gradient of the
 309 age field, and then conservatively took the ratio of 2σ over the layer thickness to rep-
 310 resent the finest temporal resolution that could be achieved before diffusion smoothed
 311 out climate signals. To compute the oldest useful ice we assumed that the coarsest use-

312 ful temporal resolution would be 10ka, such that the Nyquist frequency would catch the
 313 precession cycle at about 20ka, while the important obliquity cycle would have 4 sam-
 314 ples per cycle.

315 Using these steepening and layer thickness thresholds, we were able to estimate the
 316 oldest useful ice (Fig 7e). While in most of the domain intense vertical shear in the lower
 317 quarter of the ice column ensures that it would be difficult to obtain an intact climate
 318 record extending more than a few hundred thousand years into the past, underneath the
 319 divide the situation is different. There are two distinct patches of extremely old ice un-
 320 derneath the divide where useful ice older than 1 Ma can be expected, and in fact our
 321 model predicts that useful ice up to 1.5 Ma may be present as well, a key finding as lo-
 322 cating a continuous ice core record back to 1.5 Ma has been identified as an important
 323 target for understanding the climate of the Quaternary (Fischer et al., 2013). These two
 324 patches are elongated in the direction of the ice surface ridge; the larger of the two con-
 325 tains the present-day ice peak while the smaller patch is centered around a small sub-
 326 subsidiary peak in the ice surface topography. So long as the position of the ice divide has
 327 remained stable over time, both of these two patches are likely to contain extremely old
 328 ice in stratigraphic order.

329 It is therefore important to explore what the results of our inversion can say about
 330 the history of divide migration around Dome A. The pattern of accumulation rate re-
 331 solved by the inverse model displays a precipitation shadow effect associated with the
 332 ice sheet surface topography. We plot surface elevation contours in Fig 2b to facilitate
 333 this comparison: to the northwest of the ice divide (ie, in the direction leading towards
 334 Lambert Glacier, the Amery Ice Shelf, and the ocean), we can see the highest accumu-
 335 lation rates in the domain, about 24 mm a^{-1} . Meanwhile, to the southeast of the divide
 336 (ie, on the side facing the ice sheet interior) there is a pattern of lower accumulation sub-
 337 parallel to the ice divide, about $15\text{-}17 \text{ mm a}^{-1}$, and two additional local minima (down
 338 to $\sim 14 \text{ mm a}^{-1}$) located on the inland side of the divide as well. Furthermore, this pre-
 339 cipitation shadow is located within the center of the model domain, where errors are low
 340 (Fig 2d). Thus, while we do not have confidence in structure produced by the inversion
 341 near the edges of the domain, we do have confidence that the precipitation shadow as-
 342 sociated with ice surface topography is robust. Since these results represent a steady-
 343 state accumulation rate field averaged over the last 161 ka (the age of the oldest dated
 344 layer used in our inversion), they imply that the precipitation shadow created by the modern-
 345 day ice sheet topography has been stable for at least one and a half glacial cycles. A sta-
 346 ble precipitation shadow would argue against large-scale divide migration. Our results
 347 do not have sufficient precision to rule out small-scale divide migration, on the order of
 348 50 km in the across-divide direction. In addition, we would probably not see much dif-
 349 ference in our results if the peak of Dome A migrated up and down the present-day ridge
 350 over time (Fig 1a). However, large-scale migration, such as a jump from the present-day
 351 ice divide which is anchored on the northern spur of the GSM to an alternate ice divide
 352 which is anchored on the southern spur of the mountains (Fig 1b), can be ruled out, at
 353 least within the last one and a half glacial cycles.

354 A detailed look at the fit between the modelled and observed internal layers (Fig
 355 8) does not provide evidence for divide migration either. The precipitation shadow pat-
 356 tern is clearly visible in the raw data for all layers (Fig 8a,d,g,j,m), where it is manifested
 357 as higher layers (redder colors) on the southeast side of the ridge and lower layers (blue
 358 colors) on the northwest side. The model captures the long-wavelength spatial structure
 359 of the observed layers quite well, although the observed layers generally have more short-
 360 wavelength variability than the model. The misfits for the dated layers (Fig 8c,f,i,l) are
 361 dominated by a temporal pattern in which the model layers are systematically too high
 362 in the ice column for the younger layers (38 and 48 ka; Fig 8c,f) and systematically too
 363 low in the ice column for the older layers (90 and 161 ka; Fig 8i,l). This temporal pat-
 364 tern is explained by the fact that the steady state model is solving for a temporally av-

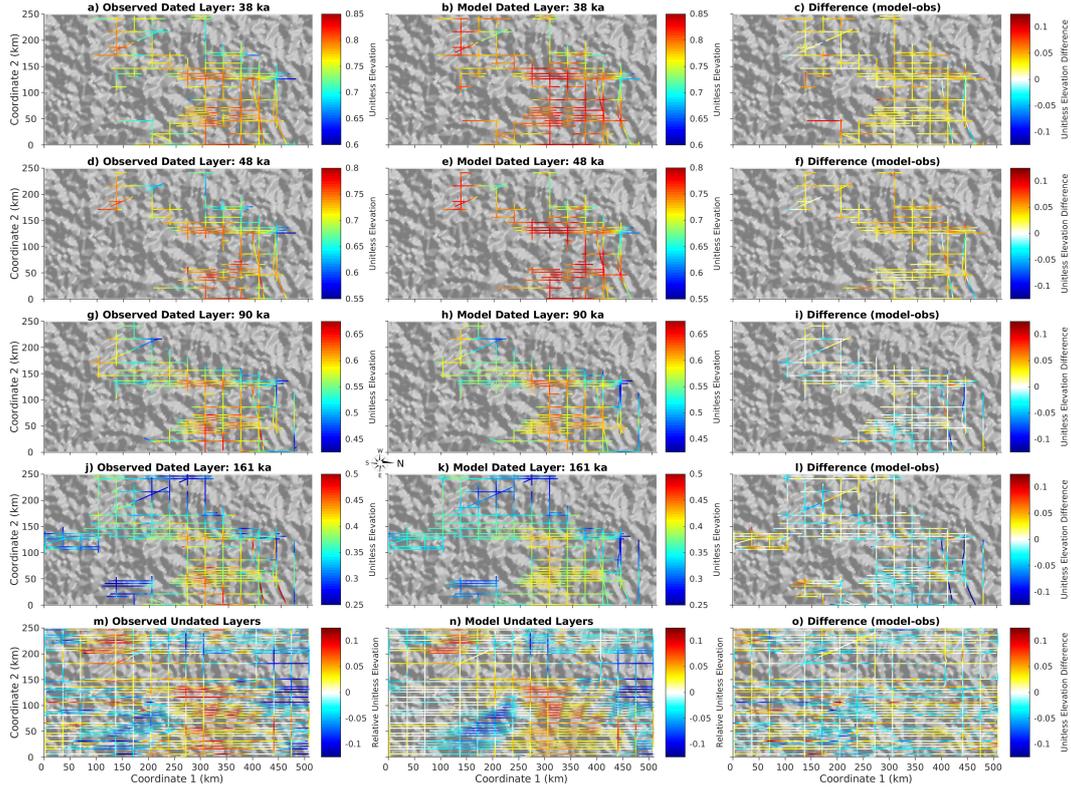


Figure 8. Best-fit model compared to the internal layer observations. Left-hand column (a,d,g,j,m) shows the observed dated and undated layers; middle column (b,e,h,k,n) shows the corresponding model layers; right-hand column (c,f,i,l,o) shows the difference, positive when the model layer is higher in the ice column. Each row represents a particular layer (38 ka, 48 ka, 90 ka, 161 ka, and undated). For the dated layers, the model layer is taken to be the isochron with the corresponding age, while for the undated layers, the model isochron selected for comparison is the isochron that best matches the average depth of the observed layer. Thus, the undated layers (m,n,o) are only sensitive to spatial gradients in layer depth rather than the absolute value of depth. Hillshaded bed topography is shown in the background of all plots.

365 eraged accumulation rate: if accumulation is higher during interglacials, then more re-
 366 cent layers should be pushed down lower in the ice column than would be implied by the
 367 temporally averaged accumulation rate, while older layers should respond more to the
 368 lower accumulation rate during glacial periods and thus be higher in the ice column. The
 369 spatial patterns in the misfit do not appear to be correlated with the precipitation shadow.
 370 The largest misfit values for the undated layers are seen near the freeze-on associated
 371 with network D, where unresolved small-scale deformation processes produce large englacial
 372 folds not captured by the model. Outside of this location, the magnitude of the layer depth
 373 misfit is generally on the order of 5% of the ice thickness or less.

374 **3 Discussion**

375 **3.1 Ice Cores and Old Ice**

376 Our model strongly suggests that the ice divide extending southwest of Dome A
 377 is a promising location to drill an old ice core, with two distinct regions of very old ice

378 found near the bed along this ridge (Fig 7). The reasons for this are straightforward and
379 robust. Our model includes all of the terms in the strain rate tensor when computing
380 ice viscosity, even if it does not compute the full stress balance (Wolovick et al., in re-
381 view). It therefore includes a simplified representation of the mechanism behind the Ray-
382 mond effect, which brings old ice layers closer to the surface underneath ice divides (Ray-
383 mond, 1983). Our model also accounts for the temperature effect on rheology, which weak-
384 ens the Raymond effect when ice near the bed is warmer and therefore softer than ice
385 near the surface. A classic Raymond arch may not be visible in the stratigraphy under-
386 neath Dome A (Zhao et al., 2018; Wang et al., 2018) owing to the twin effects of ther-
387 mal weakening and topographic draping, but the prediction that a weak Raymond ef-
388 fect should nonetheless be present is a simple consequence of the divide geometry and
389 ice rheology. In addition, vertical shear in flank positions, magnified by ice flow over a
390 highly variable bed, should cause small perturbations to overturn in the lower portion
391 of the ice column, producing the echo-free zone. In some cases, we can even observe small-
392 scale folding in the radar data as the upper boundary of the echo-free zone eats into the
393 overlying smooth stratigraphy (not shown). Under the divide, this vertical shear should
394 be greatly reduced, producing an environment much more conducive to stratigraphic con-
395 tinuity. Our prediction that extremely old ice in stratigraphic order may be found un-
396 derneath the ice divide (with small gaps corresponding to deep valleys where melting oc-
397 curs) is thus a robust consequence of the divide geometry itself.

398 The one major complication to this generally hopeful picture for ice coring is the
399 potential for divide migration. Our model predicts that regions off of the divide should
400 experience a large amount of vertical shear, and thus small stratigraphic perturbations
401 are likely to overturn. Thus, if the old-ice patches were once in a flank regime, their lay-
402 ers would probably be disturbed. However, the present-day ice divide is aligned with a
403 prominent ridge in the underlying mountains (Fig 1a,b), and it is reasonable to assume
404 that the subglacial mountains stabilize the position of the ice divide. Under this hypoth-
405 esis, it would take a large change in forcing (most likely far-field changes in surface slope
406 and ice flux that diffuse inland from the coast; see Gillet-Chaulet & Hindmarsh (2011)
407 for detailed discussion of the influence of far-field forcing on ice divide geometry) to cause
408 the ice divide to jump to a different stable position, such as the high basal topography
409 on the southern spur of the GSM (Fig 1b). The fact that our inverted accumulation rate
410 contains a precipitation shadow pattern aligned with the present-day divide (and the fact
411 that this pattern is seen in the raw layer data as well) gives us confidence that the ice
412 divide has not made this jump within the last one and a half glacial cycles. In addition,
413 the observed freeze-on reflectors are continuous for 100-150 km downstream of their source
414 regions (Fig 5a), which implies stability in both the ice flow direction and their freez-
415 ing source regions for roughly 100-150 ka. Large-scale divide migration would interrupt
416 the continuity of the observed freeze-on reflectors by both changing the ice flow direc-
417 tion and by removing the surface gradient that drives water uphill into their freezing source
418 regions. Of course, we cannot rule out earlier migrations before the earliest dated layer
419 and freeze-on reflector used in our inversion, but we expect the glacial-interglacial cy-
420 cle to be the largest forcing change experienced by the ice sheet in the recent geologic
421 past, so the fact that the divide made it through a complete cycle without migrating is
422 a hopeful sign for its stability in the longer term.

423 However, while our model may rule out large-scale divide migration in the last one
424 and a half glacial cycles, and while we may be hopeful that this result also indicates a
425 lack of large-scale migration in previous cycles, that still leaves open the possibility that
426 stratigraphic continuity could have been disturbed by small-scale divide migration. The
427 ice divide could migrate ~ 50 km or less in the northwest direction and remain roughly
428 positioned over the northern spur of the GSM (Fig 1a,b). None of the arguments against
429 large-scale migration that we advanced above apply to small-scale migration: such a mi-
430 gration would be small enough that it would still be consistent with the hypothesis that
431 divide position is stabilized by subglacial topography; our inversion has a course enough

432 resolution that we would probably not be able to distinguish the resulting precipitation
 433 shadow from one caused by the modern surface topography; and the ice flow and hydro-
 434 logic routing in the vicinity of the freeze-on regions would probably remain roughly the
 435 same. Yet such a migration would put our putative old ice patches temporarily in a flank
 436 regime, where they would be vulnerable to stratigraphic overturn due to vertical shear.
 437 Fortunately, stratigraphic overturn would not be guaranteed in this scenario. The ro-
 438 tation rate of layer perturbations is proportional to the horizontal velocity of the ice sheet,
 439 and such small divide migrations would still leave our old ice patches in a region of slow
 440 flow near the divide (Fig 6b), so it is possible that stratigraphic continuity could survive
 441 if the divide excursion was more rapid than the overturn time. More detailed dynamic
 442 modeling is needed to determine the duration and magnitude of small divide excursions,
 443 and whether those excursions would be sufficient to cause stratigraphic overturn.

444 But even assuming a stable ice divide, it would be inappropriate to use our model
 445 to choose a specific drilling location. This is due both to processes which we leave out
 446 of the model, and to the inherent deficiencies of a gridded bed map in a region with a
 447 large degree of topographic variability. The two biggest omissions from our model as it
 448 relates to old ice is the lack of a full stress balance and the lack of an anisotropic rhe-
 449 ology. Near the divide, the shallow ice approximation becomes a poor representation of
 450 the ice sheet stress state. When surface gradients are very low and flow is very slow, the
 451 stress regime becomes dominated by far-field horizontal stresses rather than vertical shear.
 452 While we did include the full strain rate tensor in our viscosity computation (Wolovick
 453 et al., in review), that is not the same as solving the Full Stokes equations for the ice flow,
 454 or allowing the flow vector to deviate from the assumed downhill direction. Furthermore,
 455 near a divide the ice can develop a preferential crystal orientation fabric, which produces
 456 an anisotropic rheology that can have major impacts on the age-depth scale, especially
 457 near the bed (Zhao et al., 2018). Finally, even if we had run a Full Stokes model with
 458 an anisotropic rheology, our model would still have been limited by the incomplete na-
 459 ture of the gridded topography.

460 3.2 Unresolved Basal Topography

461 The GSM underneath Dome A contain an enormous degree of short-wavelength
 462 topographic variability, and a large portion of this variability is not captured by BEDMAP2
 463 (Fig 9). The gridded topography cannot capture this short-wavelength variability for two
 464 reasons: 1) the AGAP flight line spacing is 5 km, so content at shorter wavelengths than
 465 that cannot be represented even with a perfect gridding algorithm; and, 2) the actual
 466 gridding algorithm used must make compromises between regions with sparse data cov-
 467 erage, where greater spatial smoothing is appropriate, and regions with dense data cov-
 468 erage like the GSM, where less spatial smoothing would have been better. Local devi-
 469 ations from the gridded topography are largest in the northwest of the domain, reach-
 470 ing RMS values of ~ 250 m (Fig 9c), and smallest in the southeast corner, where they
 471 drop to 130 m. Individual peaks and valleys often differ from the gridded topography
 472 by double the RMS value. A close examination of the along-track radar data reveals that
 473 numerous short-wavelength valleys with aspect ratios of around 1/5 (depth over width)
 474 are ubiquitous, and a few valleys with aspect ratios up to 1/3 or even slightly higher can
 475 be found as well.

476 None of these short-wavelength topographic features are included in the smooth
 477 gridded topography, but all of them would have important effects on the local ice flow
 478 in their vicinity. In a strongly shearing flank flow regime, these short-wavelength topo-
 479 graphic features should promote complex folds and stratigraphic overturn near the bed,
 480 and some of the narrower and deeper valleys may even host viscous ice eddies with coun-
 481 terflow at the base (Meyer & Creyts, 2017). In the divide flow regime, the lack of strong
 482 vertical shear means that these short-wavelength valleys and mountains are unlikely to
 483 cause overturning, but they should still play a strong role in modifying the flow and de-

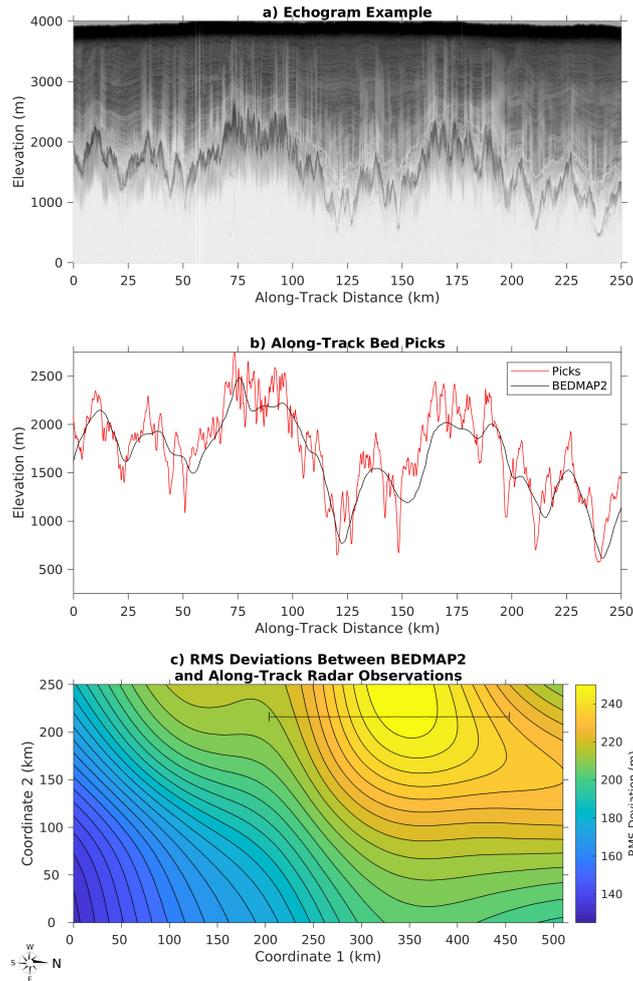


Figure 9. Short-wavelength deviations between bed topography measured along-track and gridded bed topography in BEDMAP2. a) Example echogram featuring copious topographic variability. b) Bed picks for the echogram in (a) with BEDMAP2 bed elevation overlain. c) Map of RMS deviations between gridded topography and true topography measured along-track. The “mean” stage of the RMS calculation was performed with a gaussian weighting function with a 50 km standard deviation (~ 100 km wavelength). Horizontal line shows the location of the example echogram.

484 formation regime of the ice near the bed. Our large-scale model is sufficient to make a
 485 broad-scale prediction that there are two promising old ice patches aligned with the di-
 486 vided, but any attempt to predict the precise details of the age-depth scale would need
 487 to resolve this short-wavelength topography and the effect that it has on ice flow. Closely
 488 spaced survey lines along the whole length of the divide- on the order of 500 m to 1 km
 489 apart, for the entire 100 km long divide- are an essential prerequisite to any attempt to
 490 predict the location of the oldest ice in detail. Furthermore, this unresolved short-wavelength
 491 topography is likely to have an effect on the ice sheet thermal structure, which will in
 492 turn lead to a bias in our inverted geothermal flux estimate, which we try to estimate
 493 below.

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3.3 Estimating Geothermal Flux Bias

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Our geothermal flux estimates are higher than all of the prominent prior estimates for this region (Shapiro & Ritzwoller, 2004; Fox-Maule et al., 2005; An et al., 2015; Martos et al., 2017). Fundamentally, our high estimates are being driven by the observations of subglacial water. The observations demand that subglacial water be present in the model both directly, through the water constraint, and indirectly, through the requirement that liquid water must exist at the ice sheet base in order to produce freeze-on plumes matching the freeze-on constraint. However, the ice sheet thermal state is heavily dependent on ice thickness. Thicker ice provides better insulation from the cold surface temperatures, encouraging warmer conditions, while thinner ice increases the rate of conductive cooling, encouraging colder conditions. Thus, deviations between the gridded ice thickness and the true ice thickness have the potential to change the thermal state of the bed. If the observations demand water at a particular location, but the model ice thickness is too low, then the model will compensate by making the geothermal flux higher than it should be in order to ensure that the bed is still warm. Because subglacial water is preferentially located in topographic minima, this effect will produce a systematic bias in our inferred geothermal flux estimate, since the real subglacial water bodies are located in narrow valleys which are not fully resolved by the smoothed grid.

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We can estimate this bias by looking at the mismatch between the gridded topography used as a model input and the true topography measured along-track in the radar data at the locations of the observed water bodies. To do this, we first constructed a smoothed estimate of the ice thickness bias between the water observations and the gridded topography (Fig 10a). This estimate is different from the RMS estimate we presented previously (Fig 9c) because we are only interested in the deviations at the locations of the observed water. We computed this estimate by first computing the difference between the along-track radar measurements of ice thickness at the locations of the water observations and the corresponding gridded value, and then constructing a weighted average based on the observational confidence (Fig 4a) and a Gaussian distance weighting with the same 50 km standard deviation (100 km wavelength) as the rest of the inversion. Next, we computed a smoothed ice thickness product at the same wavelength, and used that smoothed ice thickness product both to normalize the ice thickness bias (Fig 10b) and to compute a characteristic conductive heat flux under the assumption that the bed is tied to the melting point everywhere (Fig 10c). The product of those two quantities then represents the bias in conductive heat flux (Fig 10d). Since this bias in conductive heat flux should only affect our inversion results when either the water or freeze-on observations are responsible for setting the lower bound on geothermal flux, we then computed the combined contribution of those two data types to setting the lower bound (Fig 10e). Finally, we estimated the bias in geothermal flux (Fig 10f) by taking the product of that contribution with the bias in conductive heat flux. We show our final bias-corrected estimate of geothermal flux in Fig 10g. We assume that there is a 50% uncertainty in the above procedure for estimating the geothermal flux bias, and we add this error to our previous estimate of geothermal flux uncertainty (Fig 2c) in quadrature to produce our final uncertainty estimate for geothermal flux (Fig 10h).

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With a mean value of 4.5 mW m^{-2} , this estimated bias accounts for most of the 5.4 mW m^{-2} mean difference between our geothermal flux results and those of Martos et al. (2017). Given that our estimate of this bias was only approximate, and given that a proper estimate of this bias would require a high-resolution model using a high-resolution topographic grid as input (with a high-resolution field survey necessary to produce that grid), we interpret our bias-corrected results as being consistent with Martos et al. (2017), at least in a spatially averaged sense.

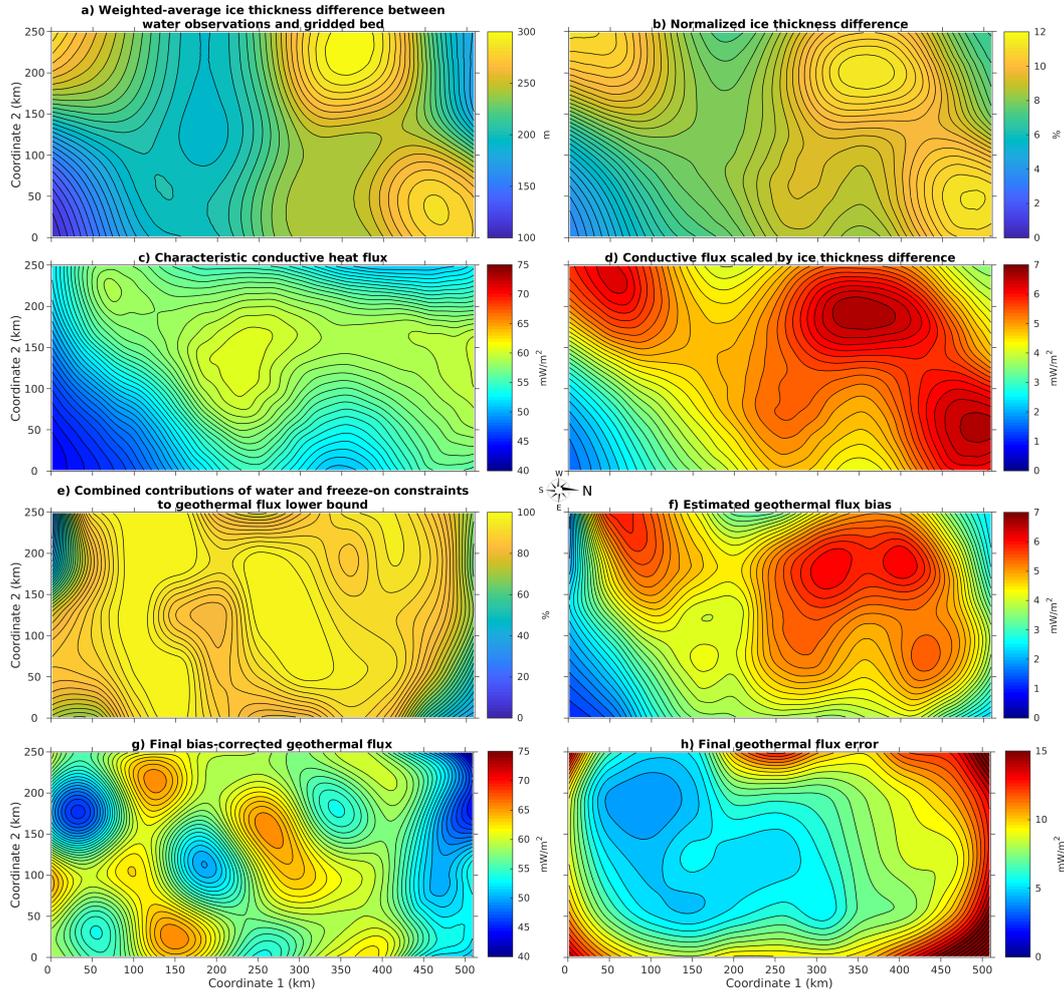


Figure 10. Bias estimation for geothermal flux. a) Offset between ice thickness at the along-track locations of the water observations and the corresponding ice thickness in the gridded topography, weighted by observation confidence (Fig 4a) and spatially smoothed; b) the same quantity, normalized by smoothed ice thickness; c) characteristic conductive heat flux; d) conductive heat flux scaled by ice thickness offset (ie, the product of (b) and (c)); e) combined contributions of water and freeze-on observations towards constraining the lower bound on geothermal flux in the inversion; f) our estimate of the bias in our inverted geothermal flux (the product of (d) and (e)); g) our final bias-corrected geothermal flux estimate (ie, Fig 2a minus (f)); and h) the uncertainty in our final estimate of geothermal flux.

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3.4 East Antarctic Geology

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Nevertheless, our inversion has revealed that geothermal flux in the GSM may well be higher than previously believed, especially in localized regions. While the mean offset between our results and the Martos estimate may be explained by the thermal effect of unresolved narrow valleys, the Martos estimate was itself on the high end of previously published geothermal flux estimates for this region. Other prominent estimates put the mean geothermal flux in our domain at 47 mW m^{-2} (An et al., 2015), 48 mW m^{-2} (Shapiro & Ritzwoller, 2004), or 53 mW m^{-2} (Fox-Maule et al., 2005). In contrast, the Martos estimate- which was the only one of the previous estimates to be constrained by local aeromagnetic data- put the mean heat flow in our domain at 57 mW m^{-2} . Our bias-corrected estimate puts the mean heat flow at 58 mW m^{-2} , with local excursions up to 66 mW m^{-2} (prior to bias correction, these figures were 62 and 72 mW m^{-2} , respectively). Thus, our inversion suggests that heat flow in the GSM may be closer to the global continental average (65 mW m^{-2} , (Jaupart & Mareschal, 2007)) than to values typical of Proterozoic continental cratons (48 mW m^{-2} , (Jaupart & Mareschal, 2007)).

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Furthermore, our inverted geothermal flux would have been even higher had we not included the Martos estimate in our cost function. In our companion paper (Wolovick et al., in review) we quantitatively attributed our inversion results to each of the misfit components: in two thirds of the domain, the Martos aeromagnetic model is responsible for 75% or more of the misfit increase at the upper bound on geothermal flux. Only in a few locations- most prominently in the freeze-on plume at the terminus of network A- does the freeze-on constraint provide a meaningful upper bound on geothermal flux. In the rest of the domain, the radar data are responsible for setting the lower bound on geothermal flux, not the upper bound. This is significant because the bias in our geothermal flux estimate arising from the presence of narrow unresolved valleys is really a bias in the lower bound only: deep subglacial valleys make it *possible* to have subglacial water at a lower geothermal flux than would be possible in shallower valleys; but of course the deep valleys will still have subglacial water if the geothermal flux is higher as well. Without the aeromagnetic constraint, the upper bound of our geothermal flux estimate would be unconstrained in a large fraction of the domain. In effect, by including the Martos estimate we gave the conventional wisdom about East Antarctic geology a 20% stake in our misfit function. If we had not credited the conventional wisdom in this way, our estimate would have been much higher.

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Our inversion has also revealed a large degree of spatial variability in geothermal flux that was not known before. The basic pattern of spatial variability we found remains the same both before and after bias correction (Fig 2a, c.f. Fig 10g). The highest values still occur in the catchment region of water network A (Fig 10g, c.f. Fig 3). All of the major water networks are still associated with dipole patterns in geothermal flux, with higher values found in their upstream catchments and lower values found near the freezing zones at their termini. This sort of spatial variability should be expected in geothermal flux; indeed, had our inverse model allowed structure at smaller wavelengths, we likely would have resolved even more variability. In places where it can be more easily measured on land, geothermal flux is known to have a high degree of spatial heterogeneity, even in old continental cratons (Jaupart & Mareschal, 2007). Differences in tectonic, magmatic, and hydrothermal history, and differences in the distribution of radioactive elements within the crust, all combine to produce a great deal of local variability in continental heat flow. Even within the population of Proterozoic cratons, heat flow varies: while the mean of this class of continental crust may be only 48 mW m^{-2} , individual cratons may be as high as 90 mW m^{-2} or as low as 36 mW m^{-2} (Jaupart & Mareschal, 2007, Table 3). The Shapiro & Ritzwoller (2004) estimates of heat flow may have been the lowest in our domain, but even they were explicit that the “error” estimate they provided was actually an estimate of the local spatial variability that should be expected about their central value. Our results do not contradict the conventional wisdom that

597 East Antarctica is an old Proterozoic continental craton. They do, however, challenge
 598 the frequently unexamined corollary that its heat flow must therefore be uniformly low.
 599 Variability at all spatial scales is a central feature of geothermal flux measurements. Some
 600 old cratons have quite high heat flow, while others are very low. A similar degree of vari-
 601 ability in East Antarctica should be the expectation, not a surprise.

602 Ice sheet modelers typically ignore the potential for spatial variability in geother-
 603 mal flux, since we usually lack the means to constrain it; but the basal hydrology and
 604 thermal structure of the ice sheet will respond to this variability nonetheless. While we
 605 cannot speak with specificity about the pattern of geothermal flux outside of our domain,
 606 we can make general inferences about how we expect the thermal structure of the ice sheet
 607 and bed to behave. First of all, we expect basal hydrology to redistribute latent heat un-
 608 derneath the ice sheets, so that cold regions downstream of melting zones will be warmer
 609 than they otherwise would be. Ice sheet thermal models without basal hydrology and
 610 freeze-on will underestimate basal temperatures downstream of melting zones, especially
 611 along narrow hydrological flow paths. Secondly, we expect to see more heterogeneity in
 612 the basal thermal state than would be implied by spatially smoothed estimates of geother-
 613 mal flux. For example, one often-cited study of Antarctic basal temperatures (Pattyn,
 614 2010) ran sensitivity tests with multiple geothermal flux maps, but all of the maps that
 615 they tested were course-resolution products with ad hoc localized adjustments to account
 616 for radar observations of subglacial lakes. As a result, they predicted that large swaths
 617 of the Antarctic ice sheet are basically guaranteed to be warm-bedded, and that the melt
 618 rate in these regions ought to be uniformly positive. In reality, there are likely to be nu-
 619 merous local variations in basal temperature, and even within regions that are uniformly
 620 warm-based, there are likely to be local transitions between melting and freezing condi-
 621 tions. The ice sheet base is probably a hodgepodge of conditions: cold-based, warm-based
 622 but freezing, and warm-based melting. This variability is consistent with observed 10
 623 km-scale variability in basal morphology (Bingham et al., 2017), which also acts as a driver
 624 for the water routing system. The expectation of heterogeneity ought to inform our in-
 625 terpretation of the geomorphic and detrital record of past ice sheet dynamics, and also
 626 our interpretation of model results for present-day ice sheets. When interpreting the re-
 627 sults of ice sheet thermal models, we should keep in mind that, while thermal uniformity
 628 may be a technically correct result given smooth boundary conditions used as model in-
 629 put, in reality we expect the ice sheet base to be heterogeneous.

630 4 Conclusions

631 We have used a formal inverse model to assimilate radar observations of subglacial
 632 water, freeze-on, and internal layers into a thermomechanical model of the ice sheet and
 633 basal hydrology around Dome A in East Antarctica. Using this inverse model, we have
 634 estimated the geothermal flux and long-term accumulation rate boundary conditions act-
 635 ing on the ice sheet over the last 161 ka. Based on those boundary conditions, we were
 636 able to produce a self-consistent estimate of the ice sheet flow field, thermal structure,
 637 and basal water flow informed by those observations.

638 Our inferred distribution of basal water and freeze-on largely matches the obser-
 639 vations, but the use of a self-consistent physical model allows us to constrain melt rates
 640 and freeze-on volumes. We find that it is unlikely that the full thickness underneath the
 641 reflectors identified by Bell et al. (2011) is composed of freeze-on; instead, a change of
 642 interpretation is warranted, such that the observed reflectors likely reflect volume scat-
 643 tering from within the body of the freeze-on unit, with transparent meteoric ice under-
 644 neath. We also predict a large amount of basal water in the trunk valleys of the GSM.
 645 While this water does not appear to be ponded in the same way as the clear subglacial
 646 lakes in the networks identified by Wolovick et al. (2013), it is likely that a large volume
 647 of subglacial water flows down these valleys and out towards downstream regions of the

648 ice sheet. This prediction could be tested by drilling to the bed, by more advanced radar
649 techniques, or by active source seismic surveys.

650 Our inferred accumulation rate field contains a precipitation shadow pattern aligned
651 with the modern-day ice divide, suggesting that there has not been any large-scale di-
652 vide migration in the last one and a half glacial cycles. The same pattern also appears
653 in the raw layer data used to constrain the model, giving us added confidence in its va-
654 lidity. The continuity of the observed freeze-on units also argues against large-scale di-
655 vide migration within the last glacial cycle. While we cannot say anything with certainty
656 about divide migrations in previous cycles, the fact that the divide went through a full
657 glacial cycle's worth of forcing changes without large-scale migration is a hopeful sign
658 for divide stability in the longer term. If the divide has indeed been stable on longer timescales,
659 then our model predicts that extremely old ice- perhaps up to 1.5 Ma- may be found in-
660 tact within two distinct patches in an elongated pattern stretching for roughly 100 km
661 underneath the divide.

662 Our inferred geothermal flux field is generally warmer than prior published esti-
663 mates for this region (Martos et al., 2017; Shapiro & Ritzwoller, 2004; An et al., 2015;
664 Fox-Maule et al., 2005). While the average offset with the aeromagnetic estimate (Mar-
665 tos et al., 2017) can be explained by the difference between the smoothed gridded to-
666 pography used as model input and the deep subglacial valleys present in reality, the aero-
667 magnetic estimate was the highest of the prior estimates, and a number of local hot spots
668 remain even after we corrected for this bias. Fundamentally, the radar observations of
669 copious subglacial water and basal freeze-on require a dynamic bed with many valleys
670 at the melting point; this sort of dynamic basal environment is inconsistent with an ex-
671 tremely low geothermal flux from a uniformly cold craton. Spatial heterogeneity in geother-
672 mal flux is an important control on the thermal structure and basal hydrology of the ice
673 sheet, and therefore observational datasets that sample aspects of the basal hydrolog-
674 ical system can in turn be used to constrain spatial heterogeneity in geothermal flux.

675 In this paper and in our companion paper (Wolovick et al., in review), we have demon-
676 strated that it is possible to combine many disparate sources of information into a sin-
677 gular self-consistent picture of the ice sheet and basal hydrology. The combination of ob-
678 servational constraints with a physically consistent forward model can yield important
679 insights about the ice sheet state and history, allowing us to estimate the forcings that
680 have acted on the ice sheet, along with the uncertainty and skewness of the probability
681 distribution of those forcings, as well as giving us a detailed picture of the resulting state
682 of the ice sheet. This nuanced picture allows us to predict the ice sheet thermal struc-
683 ture and flow field, the basal melt (or freezing) rate and water flux, and the distribution
684 of old ice and deformation within the body of the ice sheet.

685 Acknowledgments

686 This work was supported by the National Natural Science Foundation of China (No. 41941006)
687 and COLD:Finnish Academy, #322430. Undated layer picks and echo-free zone picks
688 were produced by Sara Wolovick. We thank all of the participants in the AGAP project
689 for their hard work in collecting and analyzing the datasets used in this paper. MJW
690 thanks the members of the Polar Geophysics Group at the Lamont-Doherty Earth Ob-
691 servatory for years of stimulating discussion about Dome A and other places. Inversion
692 results, data constraints, and best-fit model are available at DOI:10.5281/zenodo.4072255.

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