Characteristics of the contemporary Antarctic firn layer simulated with IMAU-FDM v1.2A (1979-2020)

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Abstract. Firn simulations are essential for understanding Antarctic ice sheet mass change as they enable us to convert satellite altimetry observed volume changes to mass changes, column thickness to ice thickness, and to quantify the meltwater buffering capacity of firn. Here, we present and evaluate a simulation of the contemporary Antarctic firn layer using the updated semi-empirical firn model IMAU-FDM for the period 1979-2020. In IMAU-FDM, we have improved the We have improved previous fresh snow density and firn compaction parameterizations, and used improved updated atmospheric forcing. In addition, the model has been tuned and evaluated against 148 in situ calibrated and evaluated using 112 firn core density observations across the ice sheet. The updated model captures the observed strong spatial variation in firn thickness and density. The temporal variation can be split into a rather stable seasonal cycle driven by snowfall, compaction and melt seasonal cycles, and more irregular decadal variations driven by snowfall anomaliesWe found that 63 to 68 % of the seasonal and decadal surface height variability is due to variations in firn air content rather than firn mass. Comparison of simulated surface elevation change with altimetry shows that the decadal trends agree reasonably well, and that the a previously published multi-mission altimetry product for the period 2003-2015 shows that performance of the updated model has improved, notably in Dronning Maud Land and Wilkins Land. Wilkes Land. However, a substantial trend difference (>10 cm yr⁻¹) remains in the Antarctic Peninsula and Ellsworth Land, mainly caused by uncertainties in the spin-up forcing. By estimating previous climatic conditions from ice core data, these trend differences can be reduced by 38 %.

1 Introduction

The Antarctic ice sheet (AIS) is the largest freshwater reservoir on Earth, and is losing mass since at least 2002 (Shepherd et al., 2018; Rignot et al., 2019), thereby contributing to around 10 % of global average sea level rise since 1993 (Oppenheimer et al., 2019). About 99 % of the AIS is covered by a firn layer (Winther et al., 2001), which represents the transition of fresh snow to glacier ice. The firn layer thickness typically ranges from 40 m in the coastal zone to and can exceed 100 m in the interior (van den Broeke, 2008), and also fluctuates in time due to changes in accumulation, melt and firn compaction. A detailed understanding of the firn layer and its variability is important for constraining current and future AIS mass change for two three key reasons. Firstly, firn depth and density estimates are required to convert altimetry observed volume-to-mass changes, which remains a major source of uncertainty in mass balance studies (Verians et al., 2021)

25 (Morris and Wingham, 2015; Verjans et al., 2021). Secondly, firn depth and density estimates are required to convert column thickness to ice thickness, to calculate solid ice discharge over the grounding line (Rignot et al., 2019). Thirdly, firn provides pore space in which nearly all (>98-94 %) the surface meltwater refreezes or is retained, thereby buffering surface meltwater and preventing mass loss by runoff of surface meltwater (Medley et al., 2020) (Medley et al., 2022).

From both observational and modelling studies, firn layer depth and density are known to vary greatly across the AIS, resulting from the wide range of (near) surface climatic conditions (e.g., Keenan et al., 2021), i.e. temperature, wind speed, accumulation and melt. Low temperatures in the interior (Fig. 1) cause slow densification rates resulting in a firn layer thickness (defined here as the depth of the $830 \, \mathrm{kg} \, \mathrm{m}^{-3}$ density level) greater than $100 \, \mathrm{m}$. On the other hand, on warm and dry ice shelves, densification and melt is enhanced, resulting in denser firn layers with a depth less than $50 \, \mathrm{m}$ (van den Broeke, 2008).

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Changes in of the AIS surface elevation are used as measures of its dynamics and mass balancean expression of changes in the firn layer and of dynamical change of the underlying ice and bedrock. The difficulty is that these seasonal cycles and decadal changes in elevation are not solely caused by fluctuations in mass, but also for a large part by fluctuations in the firn density (Ligtenberg et al., 2012; Medley et al., 2020)(Arthern and Wingham, 1998; Ligtenberg et al., 2012; Medley et al., 2022). In altimetry studies, it is therefore important to accurately separate these changes in a mass and density mass- and density change component. The mass component can subsequently be separated in an ice-dynamical and a ice-dynamical, bedrock and surface mass balance component (Willen et al., 2021).

Another important characteristics characteristic of the firn layer is its capability to retain meltwater in pore spaces by capillary forces and refreezing. In a warmer future climate, reduced accumulation, enhanced firn compaction, melt, and refreezing can all potentially lead to firn air depletion, thereby limiting the meltwater storage capacity of the firnand promoting. This is especially important over the ice shelves, where meltwater accumulation can lead to hydrofracturing-induced ice shelf collapse (Munneke et al., 2014; Datta et al., 2019), contributing to thereby accelerating future Antarctic mass loss and sea level rise (Gilbert and Kittel, 2021). Accurate information about firn conditions is thus essential both for understanding both the current AIS mass balance, and to predict its future.

Firn models can fill spatial and temporal gaps between the relatively sparse observations from firn cores and snow pits. While remote sensing products provide high temporal and spatial resolution information about surface height changes, and indirectly, firn properties of the top few meters (e.g., Davis and Poznyak, 1993; Schröder et al., 2019; Shepherd et al., 2019) (e.g., Davis and Poznyak, 1993; Medley et al., 2015; Schröder et al., 2019; Shepherd et al., 2019), they do not provide information about the deeper firn column. Firn models are, therefore, also important to obtain vertical variations in density, and can be used to perform future simulations of firn depth and density in response to climate change scenarios (Ligtenberg et al., 2012; Munneke et al., 2014).

Firn models can roughly be divided in two classes: physically based and semi-empirical models. Physics-based models describe densification using a constitutive relationship between stress and strain for snow, while semi-empirical models use physics based densification equations in combination with parameters that are calibrated with observational density profiles. Semi-empirical models neglect some processes such as drifting-snow compaction (e.g., Sommer et al., 2018) and use a larger degree of tuning to are calibrated with observations representative of the past or present climate, which while these conditions

might be not representative under future climatic conditions of a future climate. Nevertheless, semi-empirical models require less poorly known a smaller number of poorly constrained parameters, such as roughness length, snow grain shape and size, and computational demands are lower (Keenan et al., 2021). This is especially advantageous for data-scarce regions and large-scale studies. Therefore, semi-empirical firn models have been used widely for correction of satellite altimetry products (e.g., Schröder et al., 2019; Smith et al., 2020; Willen et al., 2021) (e.g., Adusumilli et al., 2018; Smith et al., 2020; Willen et al., 2021) and to assess the meltwater buffering capacity of the Antarctic firn layer (Ligtenberg et al., 2014; Munneke et al., 2014). Several

and to assess the meltwater buffering capacity of the Antarctic firn layer (Ligtenberg et al., 2014; Munneke et al., 2014). Several studies have found that the performance of semi-empirical models is comparable to that of physically based models (Steger et al., 2017; Vandecrux et al., 2020; Keenan et al., 2021).

Differences in the formulations of, e.g., firn densification and fresh snow density can lead to a substantial spread between in modelled firn thickness and air content (Lundin et al., 2017; Verjans et al., 2021)(not sure?). Using statistical emulation of firn densification models, Verjans et al. (2021) quantified how much different sources of uncertainty contribute to the total uncertainty in modelled firn thickness change of the East Antarctic ice sheet. At This study shows that, at the basin scale, the ensemble uncertainty in firn thickness changes ranges from 0.2 to 1.0 cm yr⁻¹, which amounts to a 15-300 % relative uncertainty. While differences in climate forcing have the largest influence on the spread, in basins with high snowfall and a large spatial variability of climatic conditions, firn densification formulation also has a substantial contribution to the spread (up to 46 %). In basins with recent increases in snowfall rates the contribution of fresh snow density to the total uncertainty is also substantial (up to 28 %).

In this study, we have improved the semi-empirical IMAU Firn Densification model (IMAU-FDM) (Ligtenberg et al., 2011) by updating the (i) the fresh snow density, (ii) the firn densification parameterizations, and (iii) by using updated atmospheric forcing fields. In addition, the model has been retuned against 148 field calibrated and evaluated using 112 in situ firn core density measurements across the AIS. This version 1.2A of IMAU-FDM supersedes the previous version (Ligtenberg et al., 2011). Here, we present the simulated contemporary characteristic of the Antarctic firn layer for the period 1979-2020. We focus on spatial, seasonal and decadal variability in firn thickness and firn air content (FAC). To evaluate the decadal surface elevation changes and seasonal cycles we compare the simulations to remote sensing altimetrya multi-mission remote sensing surface height change product for the period 1992-2015 (Schröder et al., 2019). Finally, we test how sensitive the modelled firn layer is to uncertainties in model variables and spin-up procedure.

2 Materials and methods

2.1 IMAU-FDM

IMAU-FDM is a semi-empirical 1D firn densification model that simulates the time evolution of firn density, temperature, liquid water content and surface height changes due to firn and SMB processes. The model employs up to 3000 layers of 3 to 15 cm thickness, which represent the firn properties in a Lagrangian way. Abbrevations of all versions of IMAU-FDM used in this study are listed in Table 1. The model was originally developed by Helsen et al. (2008), and updated by Ligtenberg et al. (2011) to a version abbreviated here to FDM v1.1. Brils et al. (2021) Brils et al. (2022) recently improved IMAU-FDM applied

Table 1. Abbreviations and characteristics of IMAU-FDM versions used in this study.

Abbreviation	IMAU-FDM version	Forcing	Fresh snow density	MO _{830*} fit
FDM v1.2A	FDM v1.2 Antarctica	RACMO2.3p2, ERA-5	Eq. (2); this study	Power (Eq. 5)
FDM FS-K	FDM v1.2 Antarctica	RACMO2.3p2, ERA-5	Eq. (1); Kaspers et al. (2004)	Power (Eq. 5)
FDM FS-L	FDM v1.2 Antarctica	RACMO2.3p2, ERA-5	Eq. (2); Lenaerts et al. (2012)	Power (Eq. 5)
FDM v1.2A-log	FDM v1.2 Antarctica	RACMO2.3p2, ERA-5	Eq. (2); this study	Log (Eq. 4)
FDM v1.1p1	FDM v1.1	RACMO2.3p1, ERA-Interim	Eq. (1); Kaspers et al. (2004)	Log (Eq. 4)
FDM v1.1p2	FDM v1.1	RACMO2.3p2, ERA-5	Eq. (1); Kaspers et al. (2004)	Log (Eq. 4)

to the Greenland ice sheet by including a refined parameterization of the thermal conductivity to version FDM V1.2G. The various versions of IMAU-FDM have been extensively evaluated against firn density and temperature observations from both

95 Greenland and Antarctica (Ligtenberg et al., 2011; Kuipers Munneke et al., 2015; Brils et al., 2021). The performance of these IMAU-FDM versions in Greenland as well as in Antarctica was found to be comparable to that of the more physically based SNOWPACK model (Steger et al., 2017; Vandecrux et al., 2020; Keenan et al., 2021). (Ligtenberg et al., 2011; Kuipers Munneke et al., 2011; We further improved IMAU-FDM for Antarctica the model to FDM v1.2A by updating the (i) the fresh snow density parameterization and (ii) the firn densification, as discussed below. Abbrevations of all versions of IMAU-FDM used in this study are

100 listed in Table ??.

2.1.1 Fresh snow density

The fresh snow density (p_s ; p_s ; $kg m^{-3}$) is an important, but poorly constrained upper boundary condition of firn models. FDM v1.1 used the a fresh snow density parameterization of Kaspers et al. (2004) based on Kaspers et al. (2004) with a correction by Helsen et al. (2008), which has been calibrated over Antarctica and yields density values that typically represent the first 0.5 m of the snowpack. It varies as a function of annual averages of average surface temperature ($\bar{T}_{\bar{s}_s}$; K), 10-m wind speed (\bar{V}_{10} ; p_s ; p_s) and accumulation (p_s ; p_s) and accumulation (p_s ; p_s).

$$\rho_s = A + B\bar{T}_s + C\bar{V}_{10} + \frac{DE\dot{b}}{DE\dot{b}}$$

$$\tag{1}$$

in which A, B, C and DE are fit coefficients. Snow crystal size and therefore type and degree of riming are indeed temperature

dependent, and previous work shows that fresh snow density indeed increase increases with increasing temperature (Judson and Doesken, 20 . Also, strong winds during snowfall cause crystal breaking, thereby reducing the snow effective grain size (Sato et al., 2008), which results in more efficient packing and increased fresh snow densities. However, at least two concerns can be raised against Eq. (1). First, due to the partly similar spatial patterns of accumulation and temperature in Antarctica, a statistical relationship between fresh snow density and accumulation may exist, however this relationship has no physical grounds. Other studies have therefore not found a dependency between annual accumulation and fresh snow density in Antarctica, Greenland and the Alps (Lehning et al., 2002; Fausto et al., 2018; Lenaerts et al., 2012). Second, Eq. (1) neglects the impact of the meteorological

conditions at the time of deposition, which can be highly variable in time. Therefore, we tested the more recent parameterization of Lenaerts et al. (2012), which has been calibrated over Antarctica, in which $\rho_{\overline{s}-s}$ is a function of instantaneous surface temperature $(T_{\overline{s}s}; K)$ and 10-m wind speed $(V_{\overline{10};}):_{10;} \text{m s}^{-1}$):

$$\rho_s = A + B \underbrace{T}_s + C \underbrace{V}_{10} \tag{2}$$

in which *A*, *B* and *C* are fit coefficients. The resulting density values are representative of the fresh skin layer density, which typically represents the upper few cm of the snowpack. Within days after depositing, the crystal structures structure of freshly deposited snow break breaks down due to wind and redistribution of drifting snow, which increases the surface snow density (Groot Zwaaftink et al., 2013). IMAU-FDM does not calculate this densification by wind and redistribution of drifting snow. The fit coefficients *A*, *B* and *C* in Eq. (2) are retuned in order for the fresh snow density are recalibrated to improve the fit with the of the simulated with observed surface snow densities, defined as the top 0.5 m of the firn columnto match the observations, as this best matches the thickness of the sampled layer. These fit coefficients are than used to simulate the fresh snow density. This recalibration is described in Section 3.1. As we only use 10 additional measurements in the evaluation dataset compared to the calibration dataset, we also include a 10-fold cross evaluation.

2.1.2 Dry snow densification rate

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Over time, the freshly deposited snow becomes denser. The rate of the dry firn densification $(d\rho/dt)$ is calculated using the semi-empirical equation of Arthern et al. (2010):

$$\frac{\mathrm{d}\rho}{\mathrm{d}t} = D\dot{b}g(\rho_i - \rho)e^{(\frac{E_c}{RT} - \frac{E_g}{RT_{ave}})}$$
(3)

where D is a constant, \dot{b} is the long-term average accumulation rate (kg m⁻² y⁻¹), g is the gravitational acceleration, ρ_{Γ_i} is the density of bubble free ice (917 kg m⁻³), R is the gas constant, ρ is the layer density (kg m⁻³), T is the instantaneous layer temperature (K), T_{ave} is the average surface $_{ave}$ is the long-term average surface skin temperature (K), and $E_{\overline{c}_{c}}$ and $E_{\overline{g}_{g}}$ are the activation energies for creep (60 kJ mol⁻¹) and grain-growth (42.4 kJ mol⁻¹), respectively. \dot{b} is used as a measure of the overburden pressure, thereby assuming that the accumulation rate is constant over time. Using FDM v1.2G for Greenland, Brils et al. (2022) show that assuming a constant \dot{b} introduces a minor error in the load experienced by a layer of firm (e.g. <3.2 % at Summit and <1.9 % at Dye-2). The constant D has different values above (0.03) and below (0.07) the critical density level ρ = 550 kg m⁻³ to represent the distinct densification processes mechanisms. For ρ < 550 kg m⁻³, densification mainly occurs by settling and sliding of grains, and for ρ > 550 kg m⁻³, it mainly occurs by deformation, sublimation and diffusion \dot{c} (Herron and Langway, 1980).

By comparing simulated and observed depths () of the 550 and 830 kg m⁻³ density levels (z_{550} and $z_{830,550}$ and z_{830} , respectively), Ligtenberg et al. (2011) found that Eq. (3) requires a correction function, the so-called MOfits (ratio of Modelled to Observed depths), which turn out to depend on the accumulation rate. These MO fits are defined using the ratio between correction terms MO₅₅₀ for ρ < 550 kg m⁻³ and MO_{830*} for 550 < ρ < 830 kg m⁻³, which are defined as the ratio of modelled and observed values of z_{550} and $z_{830*,550}$ and z_{83

correction terms MO₅₅₀ and MO_{830*} are added as a multiplier to Eq. (3)is used without correction fits. Ligtenberg et al. (2011) and Brils et al. (2021); MO values below one reduce the densification rate, and values above one enhance the densification rate. MO₅₅₀ and MO_{830*} are chosen as functions of the long-term mean accumulation rate. Ligtenberg et al. (2011) and Brils et al. (2022) used logarithmic correction functions, thus:

$$MO = \alpha - \beta \ln(\dot{b}) \tag{4}$$

in which MO is either MO₅₅₀ or MO₈₃₀₅₅₀ or MO₈₃₀₅, and α and β are fit coefficients which differ for MO₅₅₀ and MO₈₃₀₅₅₀ and MO₈₃₀₅₅₀ and MO₈₃₀₅₅₀ and MO₈₃₀₅₅₀ and MO₈₃₀₅₅₀. In FDM v1.1 a minimum value of 0.25 is imposed on MO₅₅₀ and MO₈₃₀₅₅₀ and MO₈₃₀₅. For Greenland and Antarctica, different values for α and β have been used to optimize densification. Here, we also tested a power-law function for MO₈₃₀₅₅₀₅₀₅:

$$MO_{830830*} = \delta \dot{b}^{-\epsilon} + \phi \tag{5}$$

in which δ , and ϵ and ϕ are tuning parameters. fit coefficients. The fit coefficients in Eq. (4) and (5) are calibrated, which is described in Section 3.2. As we only use 10 additional measurements in the evaluation dataset compared to the calibration dataset, we also include a 10-fold cross evaluation.

2.1.3 Heat conduction, meltwater percolation and refreezing

The conduction of heat is simulated by using a one-dimensional heat transfer equation, which couples vertical heat conduction to temperature gradients through the thermal conductivity of firn. In FDM v1.1 the thermal conductivity is a function of firn density only (Anderson, 1976), and in FDM v1.2G this has been extended to a function of density and temperature (Calonne et al., 2019)—, which is also included in FDM v1.2A. The lower boundary condition assumes a constant heat flux across the lowest layer, where the deep temperature is allowed to change along with long-term changes in surface temperature or internal heat release. Meltwater percolation is simulated using the tipping-bucket bucket method, whereby each firn layer has a maximum irreducible water content that decreases with increasing density (Coléou et al., 1999). The meltwater can percolate through all layers in a single timestep. The retained meltwater refreezes when it reaches a layer with sufficient available pore space and if the latent heat can be released a temperature below the freezing point. To capture the percolation and refreezing of meltwater more accurately, the model timestep is reduced from 3600 to 300 s for locations with the significant melt. Further details and evaluation of heat conduction, percolation and refreezing in IMAU-FDM are provided by Brils et al. (2021) Brils et al. (2022).

2.2 Model strategy

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An initial firn layer is obtained by spinning up the model over the a reference period until the firn layer is in equilibrium with its the surface climate. The equilibrium is roughly reached when the entire firn column, defined here as the depth where the density ρ reaches 830 kg m⁻³, is fully refreshed by accumulation. The required spin-up time in years is therefore calculated as the total thickness mass of the modelled firn layer divided by the annual accumulation in w.e. (Kuipers Munneke et al.,

2015). As no obvious strong long-term trends have been detected in Antarctica's surface climate and SMB during the period 1979-2020 (e.g., Favier et al., 2017; Mottram et al., 2021), and no reliable surface climate fields are available prior to 1979, the spin-up forcing is obtained by looping over the 1979-2020 forcing data. Thereby, we make the rather strong assumption that this 42-year period is representative for the past 100-1000 years and that the firn layer is in equilibrium with that climatic period. In the actual simulation after the spin-up, a minor trend in total firn air content remains, due to increasing air content in the modelled ice below the firn column(<0.6 mm yr⁻¹ averaged over the AIS) in total FAC remains, because at the bottom of the column, ice with a density between 830 - 917 kg m⁻³ slowly replaces bubble free ice (ρ = 917 kg m⁻³), which originates from the initialization of the firn column prior to the spin-up. This trend is removed prior to before further analysis of the results.

2.3 Surface height elevation change

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IMAU-FDM quantifies the impact of firn and SMB processes on the depth of the firn layer. The resulting vertical velocity of the firn top surface ($\frac{dhv}{dt_{tot}}$) is defined as the sum of these components:

$$\frac{\mathrm{d}h}{\mathrm{d}t} \underbrace{v_{tot}}_{v_{tot}} = v_{snow} + v_{sub} + v_{snd} + v_{melt} + v_{\underline{ice}} + v_{fc} + v_{by} + v_{\underline{ice}}$$

$$\tag{6}$$

where v_{snow} represents the vertical velocity component as a result of snowfall, v_{sub} of sublimation, v_{incl} of snowdrift erosion/deposition, v_{mell} and of snowmelt, v_{tcc} represents the negative contribution to firm depth due to the upward moving firm-ice transition, v_{tc} f_{c} of firm compaction and v_{by} of buoyancy on floating ice shelves. All components are derived from run-time output of IMAU-FDM. In a steady-state firm layer, v_{tcc} represents the vertical motion associated with the changing ice shelf draft when the mass of the firm layer changes. v_{tcc} represents the downward/upward movement of the surface by the local divergence/convergence of the ice flow, driven by the long-term average vertical mass flux through the lower boundary of the firm column. In a steady-state firm layer, this equals the mass flux at the upper boundary. v_{tcc} v_{tcc} is therefore equal to the mean SMB (2) times ρ_{t} of long-term average annual SMB (kg m⁻² yr⁻¹) over the reference period divided by ρ_{t} , but of opposite sign. All components are derived from run-time output of IMAU-FDM. The minor residual trend in dhH/dt due to increasing air content in the ice below the firm layer is added to v_{tcc} v_{ty} , only relevant over ice shelves, is equal to the negative change in firm mass divided by the density of sea water.

2.4 RACMO2.3p2 forcing

IMAU-FDM is forced at the upper boundary with 3-hourly fields of snowfall, sublimation, snowdrift erosion, 10-m wind speed, surface temperature, snowmelt and rainfall. In this study, FDM v1.2A is driven with ERA-5 reanalysis data (Hersbach et al., 2020), dynamically downscaled with the regional atmospheric climate model RACMO2.3p2 (Van Wessem et al., 2018). RACMO2.3p2 is run with a horizontal resolution of 27x27 km and the simulation covers the period 1979-2020. RACMO2.3p2 has been thoroughly evaluated for the Antarctic and Greenland ice sheets by Van Wessem et al. (2018) and Noël et al. (2018), respectively. In comparison to RACMO2.3p1, RACMO2.3p2 now-employs upper-air relaxation, has updated topography, and

a retuned cloud scheme and snow properties. Upper-air relaxation is the indiscriminate nudging applied to the upper part of the atmosphere only. The most important effects of these changes are (i) increased snowfall in the interior ice sheet interior (ii) reduced snowdrift sublimation and (iii) increased snowmelt (Van Wessem et al., 2018). Several Antarctic studies use output from FDM v1.1 forced with RACMO2.3p2 (e.g., Schröder et al., 2019; Keenan et al., 2021), but the impact of the improved forcing has not yet been described. As snowfall and surface melt are defining drivers of the state of the firn, we also assess the impact of the updated forcing in this paper. In addition to the updates of RACMO2.3p2, previous FDM simulations for Antarctica used dynamically downscaled ERA-Interim reanalysis data, which has been replaced by ERA-5 in 2019. This results in an improved forcing, for For example, surface mass balance anomalies in Dronning Maud Land from 2006 to 2018 are better captured in ERA-5 compared to ERA-Interim (Gossart et al., 2019). We expect that forcing differences in surface mass balance anomalies have a larger impact on the temporal evolution of the surface height compared to adjustments in the fresh snow density and dry snow densification parameterizations (Verjans et al., 2021).

2.5 Observational data

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To tune-calibrate and evaluate the firn model, we collected published firn density profiles from widely varying locations across the AIS (Fig. 1). We used 125-112 density profiles from firn cores and 8 density profiles from neutron density probe measurements, combining multiple datasets (van den Broeke, 2008; Schwanck et al., 2016; Bréant et al., 2017; Fernandov et al., 2010: Montgomery et al., 2018: Fourteau et al., 2019: Olmi et al., 2021; Winstrup et al., 2019). The firn cores and neutron density probe measurements are mostly obtained in summer months between 1980 and 2020. For tuning the MO fits, only dry firn cores are used, as Eq. (3) described dry snow densification only. A firn core is considered dry if its location experiences on average less melt than 5 % of the average annual accumulation. For the MO fits, 104 dry firn cores could be used, which improves upon Ligtenberg et al. (2011), who used data from 48 dry firn cores. To evaluate the firn density profiles from the simulation using the derived MO fits, 122 firn cores could be used. For tuning the calibrating the fresh snow density parameterization, observations of the density of surface snow, defined as the top 0.5 m, from 65-61 firn cores could be used. Eight additional surface snow density values from density profiles of neutron density probe measurements and two from firn cores (Montgomery et al., 2018) were added after the tuning calibration, and are thus only used for evaluation. We used densities of the top 0.5 m, as this depth is representative of many in situ surface snow density measurements, it generally consists of snow from multiple snowfall events, and it captures the densification by wind and redistribution of drifting snow (Brun et al., 1997). For calibrating the MO fits, only dry firn cores are used, as Eq. (3) describes dry snow densification only. A firn core is considered dry if its location experiences on average less melt than 5 % of the average annual accumulation. For the MO fits, 93 dry firn cores could be used, which improves upon Ligtenberg et al. (2011), who used data from 48 dry firn cores. In addition to the 93 dry cores used for calibration of the MO ratios, another 10 wet cores and five additional dry cores were used to evaluate the z₅₅₀ and z₈₃₀ model output. For evaluating FAC output, 31 firn cores could be used. Table S1 in the Supplement lists all measurements that have been used, the corresponding coordinates, method and citation.

In order to verify whether the modelled surface elevation trends and seasonal variability align with observations, we used a multi-mission satellite altimetry product covering the period January 1992 to December 2015 as presented by Schröder et al.

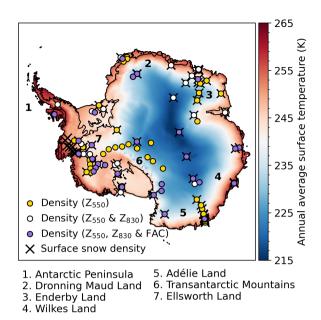


Figure 1. Map of Antarctica showing annual average surface temperature and the locations of the in situ density observations. The yellow circles indicate observational locations of the depths of the critical density level of $\rho = 550 \text{ kg m}^{-3}$ (z_{550550}), the white circles of the depths of the critical density levels of both $\rho = 550 \text{ kg m}^{-3}$ (z_{550550}) and $\rho = 830 \text{ kg m}^{-3}$ (z_{830830}), the purple circles of the depths of the critical density levels of both $\rho = 550 \text{ kg m}^{-3}$ (z_{550}) and $\rho = 830 \text{ kg m}^{-3}$ (z_{830}), and of the firm air content (FAC), and the black crosses of the surface snow densities (top 0.5 m of the firm column). The grey circles indicate ten additional locations that were included in the sensitivity analysis (on top of the yellow and white circles). The numbers indicate regions mentioned throughout the manuscript.

(2019). In this product, data from the ERS-1, ERS-2, Envisat, ICESat, and CryoSat-2 missions are combined into a monthly time series on a 10x10 km grid, with the polar gap, floating ice shelves and steep terrain areas being excluded. Validation of this dataset with in situ and airborne observations showed that the products are successfully combined (Schröder et al., 2019). The altimetry product has been resampled to the FDM grid with the bilinear resamplingmethodusing bilinear resampling. The measurement precision over flat terrain ranges from 5 to 15 cm depending on the satellite. However, as the precision decreases with increasing surface slope, the uncertainty can reach several meters in the coastal regions with steep topography. Offsets due to variations of varying penetration depth of the radar signals (thus relevant for ERS-1, ERS-2, Envisat and CryoSat-2) caused by variations in electromagnetic properties of the ice sheet are not included in this estimateuncertainty estimate. These additional offsets can reach up to several centimeters per year, and can for instance increase the seasonal amplitude at continent-wide scales by up to several centimeters (Nilsson et al., 2022). While corrections are applied to account for variations in penetration depths, these corrections are in many cases unable to fully correct the artificial signals, which can e.g. be seen from the intermission variation in seasonal amplitude (Schröder et al., 2019; Nilsson et al., 2022). On the other hand, signal penetration is negligible for laser measurements (such as ICESat, ICESat-2, Operation IceBridge). However, ICESat and Operation IceBridge datasets have poor spatial and temporal coverage, and we believe that the record of ICESat-2 is

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currently is too short (only two years and two months overlap with FDM v1.2A, and no overlap with FDM v1.1p1) to provide a viable evaluation. Therefore, we used the multi-mission satellite altimetry product as presented by Schröder et al. (2019), which has a good spatial and temporal coverage. Our firn simulations only provide elevation change due to anomalies in firn thickness caused by variations in e.g. snow accumulation, surface temperature, firn densification and melt. The difference between the observed and simulated elevation change therefore represents a combination of elevation changes due to ice-dynamical imbalance, bedrock motion, errors in the firn model and its climate forcing and errors in the altimetry observations. For this comparison, we therefore exclude regions with ice-dynamical imbalance, by using the ice-dynamical imbalance mask of Shepherd et al. (2019), which mainly consists of Pine Island, Thwaites, Totten and Getz glacier basins and the Kamb ice stream. Validation of this dataset with in situ and air borne observationsshowed that the products are successfully combined (Schröder et al., 2019).

2.6 Sensitivity analysis

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To study the model sensitivity to input and parameter uncertainty we performed a sensitivity analysis for the 105 observational 106 locations shown in Figure 1.—S1a. These locations include 95 locations from firn core sites (the firn cores used are located at 95 unique grid points, excluding five sites that were later added for evaluation). To improve the representation of high accumulation and low temperature areas representativeness of the climatic conditions over the AIS, we used 10-11 additional locations in this analysis (Fig. 1). that are located in high-accumulation or low-temperature regions, as these areas are underrepresented in the 95 firn core sites (Figs. S1b,c).

All the performed experiments are listed in Table 2. Verjans et al. (2021) and Lundin et al. (2017) showed that differences in accumulation, temperature, firn densification formulation and fresh snow density can lead to a substantial spread in modelled firn thickness and air content. We performed model sensitivity tests in which these components were separately adjusted by adding and subtracting the estimated uncertainties. The uncertainty in the accumulation forcing was based on the spread of precipitation over the AIS among regional climate models and a re-analysis product (1-8%) (Mottram et al., 2021). The uncertainty of temperature forcing was based on the For each of these components, we separately performed a model sensitivity test, which includes the spin-up as well as the main run. The fresh snow density was varied with the RMSE from the evaluation with in situ measurements (+1.5)(Van Wessem et al., 2018). For the 30 kg m⁻³; Section 3.1). To test the sensitivity to uncertainty of the dry snow densification rate, we perform a simulation where we use the 95% confidence intervals table*interval boundaries of the MO fits (Eqs. 4 and 5; Fig. 3). The uncertainty of the fresh snow density was based on evaluation with in situ measurements (30; Section 3.2). Another important To test the sensitivity to uncertainty in accumulation and temperature forcing, we use the standard deviation of snowfall and temperature among regional climate models and reanalysis products (Carter et al., 2022) (Fig. S1). This spread varies spatially and we vary the accumulation with the ratio of the ensemble standard deviation to the ensemble mean of each location, and vary the temperature with the ensemble standard deviation.

Another source of forcing uncertainty is the climate forcing during the spin-up period. As explained in Section 2.22.2, the spin-up forcing is obtained by looping over the 1979-2020 forcing data. However, firn core (Thomas et al., 2017) (Thomas et al., 2015, 2017) and isotopes (Stenni et al., 2017) studies show that in several Antarctic regions, especially the Antarctic Peninsula the Antarctic

Peninsula and Ellsworth Land, the accumulation and temperature have been increasing by were typically about 10 % and 1 K over the lower during the last centuries. In addition, over the remainder of the AIS, the accumulation and temperature were about typically 5 % and 0.5 K higher or lower during the last centuries. Therefore To investigate the typical effect this has on the IMAU-FDM results, we include schematic sensitivity tests in which we adjusted the accumulation and temperature forcing with these values only during the spin-up period. In the accumulation sensitivity test we reduced the accumulation prior to 1895. To mimic a gradual increase in precipitation over the Antarctic Peninsula and Ellsworth Land, we performed a spin-up experiment in which the precipitation up until the third-to-last loop (the 42 year long reference period) is decreased by 10 %, from 1895 to 1936 in the second-to-last by 6.66 % and from 1937 to 1978, and in the last by 3.33 %. In the temperature sensitivity test we reduced the temperature prior to 1895 with respect to the mean precipitation of 1979-2020. To mimic a gradual increase in temperature over the Antarctic Peninsula and Ellsworth Land, we create a spin-up experiment in which the temperature up until the third-to-last is decreased by 1, from 1895 to 1936. K, in the second-to-last by 0.66 K and from 1937 to 1978, and in the last by 0.33 K. Over the remainder of the AIS accumulation and temperature were both decreased and increased with 5 % and 0.5 K during the spin-up period. As the uncertainties between temperature and accumulation during the spin-up period are not independent dependent, we also performed a test-tests in which we simultaneously adjusted the temperature and precipitation.

For all our experiments, except for the MO fits, we re-ran our model calibration procedure to get the optimal MO fits. We then ran the model with these updated MO fits and compared the outputs to our reference FDM v1.2A run. We investigate the impact of the uncertainties on the FAC and surface elevation change. As the model set up of FDM v1.2A assumes a steady-state firm layer, there is no net surface elevation trend in this simulation over the period 1979-2020, except for sensitivity tests in which the accumulation and temperature during the spin-up period are adjusted. In the remainder of the sensitivity experiments, we therefore compare surface elevation change over the randomly selected period 2015-2020, instead of over 1979-2020. By relating the sensitivities of FAC and surface elevation change of each experiment to climatic variables at the sample locations, we can expand the uncertainty estimates over the entire ice sheet. The empirical relations for each sensitivity experiment are included in Figure S3. The uncertainties of each test were combined quadratically to obtain a value for the total uncertainty, where for the spin-up experiments only the combined temperature and accumulation experiments are included.

3 Tuning Calibration and model performance

3.1 Fresh snow density

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To tune calibrate and evaluate the fresh snow density parametrization, we performed simulations with FDM v1.2A using parameterization of Kaspers et al. (2004), Lenaerts et al. (2012) and the parameterizations of Kaspers et al. (2004) (Eq. 1). Lenaerts et al. (2012) (Eq. 2), the latter with updated constants, and of FDM v1.1p1. These simulations are abbreviated as FDM FS-K, FDM FS-Land, FDM v1.2A and FDM v1.1p1, respectively. The fit coefficients and statistics of the simulations are listed in Table 1. For all sensitivity tests presented here, we calculated the associated MO fits. However, the impact of the different MO fits on the surface snow densities is negligible. 3. Figure 2a shows the modelled against observed surface snow

Table 2. Overview of the sensitivity experiments. The inputs and parameters are adjusted by subtracting and adding the uncertainties either during the entire run or only during the spin-up. The values used are based on the references provided or on analysis within this work.

Experiment name	Variable	Variation	When	Reference
MO fits	MO fits	+ and -95 % confidence interval	entire run	Figs. 3a,b
$ ho_s$	$ ho_s$	+ and -30 kg m^{-3}	entire run	RMSE evaluation Section 3.1
T_s	T_s	+ and - spatial variable K	entire run	σ from Carter et al. (2022)
b	b	+ and - spatial variable $\%$	entire run	σ from Carter et al. (2022)
spin-up T _s	T_s	+0/+1 K and -1/-0.5 K ^a	spin-up	Stenni et al. (2017)
spin-up b	b	+0/+10% and -10/-5 $\%^a$	spin-up	Thomas et al. (2017)
spin-up T_s , b	b and T_s	+0/+5 %, +0/+0.5 K and -10/-5 %, -1/-0.5 K a	spin-up	Stenni et al. (2017); Thomas et al. (2017)

^a The values on the left side of the slash indicate the variation for the Antarctic Peninsula and Ellsworth Land, and the values on the right side of the slash for the remainder of the AIS. Temperature and accumulation over the Antarctic Peninsula and Ellsworth Land are the second-to-last loop of spin-up reduced by 6.66 % and 0.66 K and in the last loop with 3.33 % and 0.33 K.

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densities for FDM FS-K and FDM v1.2A. FDM FS-K somewhat overestimates the surface snow densities (bias = 18.2-17.6 $kg m^{-3}$), especially for high surface density, and FDM v1.1p1 to a lesser extent (bias = 10.3 $kg m^{-3}$), where the update in the atmospheric forcing results in a poorer performance of FDM FS-K compared to FDM v1.1p1. FDM FS-L (not shown)FS-L substantially underestimates surface snow densities (bias = $-44-45.1 \text{ kg m}^{-3}$), but has a higher correlation with the observations than FDM FS-K. The degree of underestimation increased increases with increasing wind speed, which suggests that a higher value of fit coefficient C is needed to properly capture the impact of wind compaction. This aligns with the fact that IMAU-FDM does not include densification by wind packing. In the optimized fit coefficients (FDM v1.2A, Table 23) the value of C is 2.6 times higher than the value reported by Lenaerts et al. (2012). The wind dependency is also 2.4 times stronger than proposed by Kaspers et al. (2004), which is in agreement with Sugiyama et al. (2012) who found a factor 2.8. The RMSE of the surface snow density of FDM v1.2A is 26 %-respectively 28 % and 23 % lower compared to FDM FS-LFS-K and FDM v1.1p1, due to both minimizing the bias and improving the representation of the variability. In general Compared to FDM FS-K, the surface snow density is reduced with by 18 kg m⁻³, especially in the high accumulation high-accumulation margins, and to a lesser extent in windy escarpment regions (Fig. 2b). The updated parameterization has a similar bias and RMSE to-With the 10-fold cross evaluation for FDM v1.2A, we found that the RMSE for the surface snow density observations compared to the SNOWPACK model and the Community Firn Model (CFM)(Keenan et al., 2021; Medley et al., 2020)slightly increases to 30.4 kg m⁻³ (compared to 28.8 kg m⁻³ for the reference FDM v1.2A). The fresh snow density parameterization for Greenland used in FDM v1.2G, is only a function of vearly temperature (Brils et al. (2021). This is in contrast with Antarctica, where we found a strong dependency with annual average temperature (Brils et al., 2022), owing to a lack of co-located surface snow density, temperature and wind speed observations on the GrIS. This contrasts to the expressions used here, which includes a dependency on instantaneous wind speed and temperature, which is in line with previous work (Keenan et al., 2021; Lenaerts et al., 2012; van Kampenhout et al., 2017), and likely owing to the larger range in temperature

Table 3. Fit coefficients and evaluation of the fresh snow density parameterizations (Eqs. 1,2). The fit coefficients of FDM FS-K and FDM FS-L are taken from Kaspers et al. (2004) and Lenaerts et al. (2012), respectively.

Version	$\begin{array}{c} A \\ (\text{kg m}^{-3}) \end{array}$	B $(kg m^{-3} K^{-1})$	C (kg s m ⁻⁴)	E $(\text{kg yr m}^{-3} \text{ mm w.e.}^{-1})$	\mathbb{R}^2	RMSE (kg m ⁻³)	Bias (kg m ⁻³)
FDM v1.2A ^b	83	0.77	11.67		0.46	28.8	-1.3
FDM FS- K^a	-77	1.5	6.8	0.075	0.39	40.2	17.6
FDM FS-L b	97.5	0.77	4.49		0.41	55.3	-45.1
FDM v1.1p1 ^a	-77	1.5	6.8	0.075	0.37	37.3	10.3

^a using annual average accumulation, temperature and wind speed. ^b using instantaneous temperature and wind speed.

and wind speed conditions during snow deposition in Antarctica. For all sensitivity tests presented here, we calculated the associated MO fits. However, the impact of the different MO fits on the surface snow densities is small (<0.1 %).

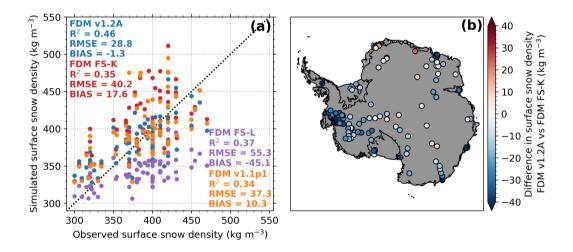


Figure 2. (a) Scatter plot of simulated against observed surface snow density using the fresh snow density expressions of FDM v1.2A and Kaspers et al. (2004) (FDM FS-K) (Kaspers et al., 2004), FDM FS-L (Lenaerts et al., 2012) and FDM v1.1p1. (b) Map of average difference between the simulated surface snow densities of the two tests (FDM v1.2A —and FDM FS-K) for the locations with observations.

3.2 Dry snow densification rate

To tune and evaluate calibrate the dry snow densification rate we first performed a simulation of FDM v1.2A without MO fits corrections. The resulting MO fits and statistics are listed in Table 2-4 and shown in Figures 3a,b. The optimal MO₅₅₀ fit is less steep with respect to accumulation (β = 0.117 vs 0.151) than the one found by Ligtenberg et al. (2011) in FDM v1.1p1. The fresh snow density is now independent of accumulation and generally lower, and hence more densification is

needed at high accumulation sites. The updated parameterization of the fresh snow density leads to a high-accumulation sites to match the density measurements. This might explain the slightly deteriorated correlation for the MO₅₅₀ of MO₅₅₀. For the MO₈₃₀ regressionline_{830*} regression, we found that a power fit (FDM v1.2A) compared to a logarithmic fit (FDM v1.2A-log) improves correlation (R² ² = 0.88 compared to R² = 0.83). Furthermore, a minimum value is now explicitly derived with the power fit, so that a rather arbitrary minimum value no longer needs to be prescribed. When the locations are rerun using the derived MO fitsnew MO values with the power fit for MO_{830*}, the resulting RMSE of the modelled z_{830*}, and z₈₃₀ (firn thickness) with the power fit is respectively 25 and FAC are respectively 39, 36 and 49 % lower compared to the logarithmic fit (Figs. 3c,d). The update in atmospheric forcing without updated MO calibration does not result in an improvement in FAC (shown by the similar RMSE of FDM v1.1p1 and FDM v1.1p2). Also, the RMSE of FAC simulated with FDM v1.2A is respectively 21 and 23 % lower compared to the logarithmic fit FDM v1.1p1 and FDM v1.1p2, which indicates smaller improvement compared to FDM v1.2A-log (49 %). This is caused by the difference in MO fit between FDM v1.2A-log and FDM v1.1 (Fig. 3e). b), which results from a combination of the larger firn core dataset and updated atmospheric forcing.

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Using a power compared to logaritmic fit results in higher $MO_{830-830*}$ values for low (<400 mm yr⁻¹) and high (>1,400 mm yr⁻¹) accumulation regions, and lower $MO_{830-830*}$ values for intermediate accumulation regions (400-1,400 mm yr⁻¹) (Fig. 3b). Therefore, the densification rate is increased in high and low accumulation low-accumulation regions and reduced in intermediate accumulation regions. This results in decreased firn thickness in high and low accumulation low-accumulation regions, and in increased firn thickness in intermediate accumulation regions (Fig. 3d). Especially large differences are found in high accumulation high-accumulation areas where the MO fit curve approaches zero. Differences between the MO fits used for Greenland in decreases asymptomatically towards zero. With the 10-fold cross evaluation for FDM v1.2G might partly be related to the fact that dry firn cores in Antarctica cover a larger range of annual average accumulation conditions, 20 A, we found that the RMSE for z_{550} slightly increases to 2.33 m (compared to 2.26 m for the reference FDM v1.2A), and the RMSE of z_{830} to 960 compared to 80 to 680 in Greenland, 9.14 m (compared to 8.87 m for the reference FDM v1.2A).

In FDM v1.2G, the version for the Greenland ice sheet, a logarithmic MO₈₃₀ fit was applied. Dry firn cores from the GrIS cover a smaller range of average annual accumulation, 80-680 mm yr⁻¹ compared to 20-960 mm yr⁻¹ in Antarctica. If we only include cores from the 80-680 mm yr⁻¹ accumulation range, we find a similar R² value for the logarithmic vs power fit (0.67 vs 68). Another difference is that in FDM v1.2G, an almost constant value of 0.67 for the MO₅₅₀ fit was found, which implies a linear correlation between the densification rate and the accumulation rate. In FDM v1.2A we do find a moderate (R²=0.37, β = 0.12) correlation of MO₅₅₀ with accumulation, i.e. AIS densification rate depends more strongly on accumulation than temperature compared to the GrIS. Again, if we only include cores from the 80-680 mm yr⁻¹ accumulation range this correlation weakens (R²=0.08, β = 0.07).

Table 4. Fit coefficients and statistics of the MO fits (Eqs. 4,5). The fit coefficients and statistics of FDM v1.1p1 are taken from Ligtenberg et al. (2011). The FDM v1.2A-log uses the settings of FDM v1.2A; only Eq. (3) is optimized using a logarithmic MO_{830*} fit.

Version	Fit	α	β	σ	ϵ	φ	\mathbb{R}^2	RMSE (m)
MO ₅₅₀ FDM v1.1p1	Log (Eq. 4)	1.435	0.151				0.43	
MO ₅₅₀ FDM v1.2A	Log (Eq. 4)	1.288	0.117				0.37	
MO _{830*} FDM v1.1p1	Log (Eq. 4)	2.366	0.292				0.71	
MO _{830*} FDM v1.2A-log	Log (Eq. 4)	2.504	0.334				0.83	0.13
MO _{830*} FDM v1.2A	Power (Eq. 5)			6.387	0.477	0.195	0.88	0.11

4 Spatial and temporal variability of AIS firn depth and density

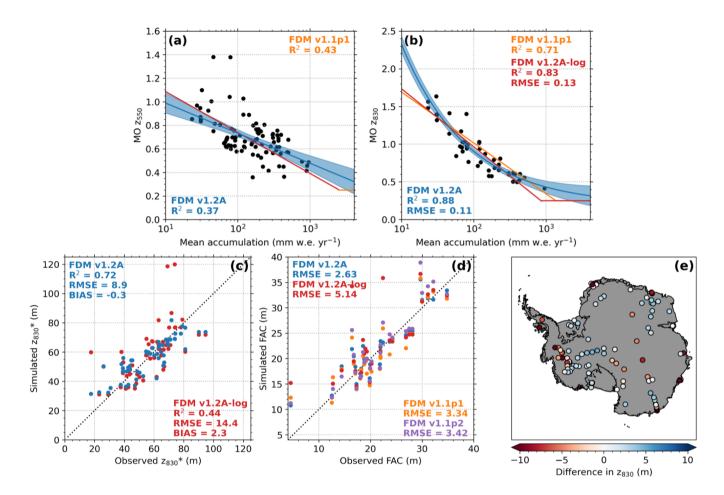
4.1 Spatial density variability

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Using the improved firn model, we simulate the spatial and temporal patterns of the AIS firn characteristics at 27 km horizontal resolution. The spatial patterns of the densities, including values found by field measurements observed values, are shown in Figure 4. Figure 4a shows the spatial patterns of the average surface snow density of the (top 0.5 m), which is on average 376 kg m⁻³. The patterns are primarily a reflection of the variation in fresh snow density (eq. 2), with the lowest values on the very cold and calm East Antarctic East Antarctic plateau and higher values in the windy escarpment regions and along the warm coastal margins. On top of In addition to this, the surface snow density further increases over the ice shelves and, where the average surface snow density amounts to 414 kg m⁻³, and over the coastal margins, as where intermittent melting and refreezing of retained meltwater increases the near-surface density.

In line with observations (Winther et al., 2001), nearly all (>99 %) of the AIS is covered by a layer of firm. This represents the ice sheet fraction where average SMB is positive. The spatial pattern of the depths of the critical density levels z_{550} and z_{830} are shown in Figures 4b and c, and which are on average 14.1 and 73.8 m. The patterns are roughly the inversed pattern of the surface snow density, which constitutes the upper boundary of the depth-density profile. For simplicity, henceforth we define the firn thickness as z_{830830} . The patterns vary spatially across climatic regions with temperature as a primary driver and accumulation as a secondary driverand surface melt as secondary drivers. The highest z_{550} and z_{830} values z_{550} and z_{830} values (30.6 and 114.2 m, respectively) are found in the interior, where the densification rate is low due to low temperatures, and the surface snow density is low. Generally, the z_{550} and z_{830} values are low z_{550} and z_{830} are small along the warm coast coastal margins due to faster densification, higher surface snow densities and melt. However, especially the z_{830} salues can be relatively high in high accumulation high accumulation regions along the coast, particularly in West Antarctica and the Antarctic Peninsula. In these regions, the accumulation is so high that the fast burial of snow leads to a thick firn layer despite the mild The high accumulation rates in these regions result in less densified snow with depth and thus thick firn layers despite relatively high temperatures. The z_{550} and z_{830} values are also low z_{550} and z_{830} values are lowest in relatively warm



and dry regions on and around Ross, Filchner Ronne and Amery ice shelf, where fresh snow is buried more slowly, and the densification rates are high due to relatively high temperatures. The measurements show that the model captures the spatial variations in z_{550} and z_{830} well.

Figure 4d shows the age of firn at the pore close off depth, here assumed to equal z_{830,830}, which is on average 754 yr. The combination of strongly increasing firn depth and decreasing accumulation towards the interior leads to firn ages at close-off depth that exceed 2000 years of up to 3240 yr in central East Antarctica, whereas the pore close-off firn age in warm and wet coastal margins can be as low as 20 years. These data are useful for the interpretation of the difference between the age of air in the bubbles and the ice that encloses the bubbles (delta age; Blunier and Schwander, 2000)yr. If we compare this to

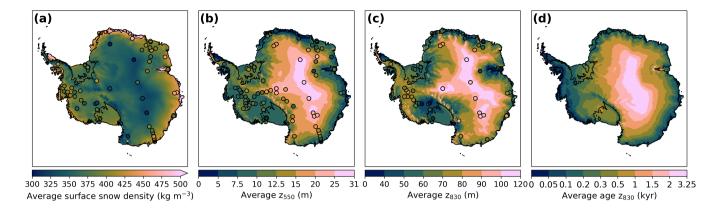


Figure 4. Maps from FDM v1.2A of the average (**a**) surface snow density (defined as the top 0.5 m), depths of critical density levels of (**b**) ρ = 550 kg m⁻³ (z_{550550}) and of (**c**) ρ = 830 kg m⁻³ (z_{530830}) (the firn thickness), and (**d**) firn age of at the critical density level of (c) ρ = 830 kg m⁻³. Circles indicate the values found by field measurements.

observations, we find an RMSE = 231 yr, R² = 0.99. On average, the RMSE is 25 % of the observed values. In comparison, 415 RMSE of z₈₃₀ is 15 % of the observed values and R² = 0.73. These simulations add to our understanding of the large spread in timescales of the firn.

4.2 Firn air content

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Firn air content (FAC), a measure of the total (vertically integrated) pore space, is an important output of firn models as it ean be is used to convert observed volume changes to mass changes in mass balance studies, column thickness to ice thickness, and is an indicator of the meltwater buffering capacity of the firn. The firn air content FAC is expressed in m and represents the change in depth that would occur if the firn layer would be compressed to the density of glacier ice. The average firn air content FAC of the AIS according to IMAU-FDM is 22.8 m, and it varies spatially between 0 to 40 m across climatic regions (Fig. 5a). The pattern resembles the z₈₃₀₋₈₃₀ pattern (Fig. 4c); thick firn layers contain the most air. High values are found in the cold interior and along parts of the coast, particularly in West-Antarctica West Antarctica, and in parts with very areas with high accumulation. Low values are found along the warmer coastal margins, with ice shelves having an average FAC of 15.9 m. These average FAC estimates are somewhat 5 % and 7 % lower compared to values simulated with the CFM forced with MERRA-2 climate data of 24.9 and 19.4-24.0 and 17.0 m respectively (Medley et al., 2020), which might be explained by higher accumulation rates in their forcing data(Medley et al., 2022).

The average FAC over the entire AIS (including the ice shelves) simulated with FDM v1.2A is 0.6 m higher than simulated with FDM v1.1. The significant spatial FAC differences between FDM v1.2A and FDM v1.1 are shown Figure 5b. In the highly elevated and low accumulation low-accumulation regions the new FAC is lower, whereas in the intermediate accumulation regions the new FAC is higher. This implies that the range of FAC is smaller and closer to the mean less spatial variable than previously modelled. The differences in FAC pattern can be explained by differences in the (i) MO₈₃₀ the MO₈₃₀ fit, (ii) the

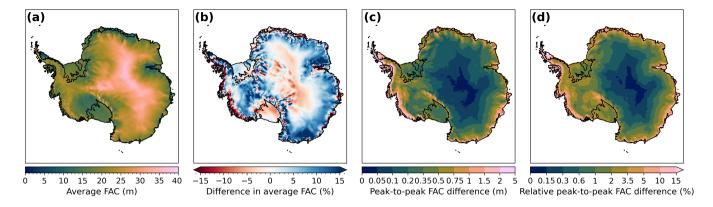


Figure 5. Maps from FDM v1.2A of average (a) firm air content (FAC), (b) the difference in FAC (FDM v1.2A - FDM v1.1p1), (c) peak-to-peak FAC difference, (d) relative peak-to-peak FAC difference compared to the average FAC from (a). The peak-to-peak is defined as the difference between the highest and the lowest FAC value occurring during the study period (1979-2020).

fresh snow density, (iii) snowfall and (iv) snowmelt. FDM v1.2A has higher MO₈₃₀ values for low and high accumulation regions compared to FDM v1.2A, which results in a faster densification and therefore lower FAC values. In intermediate accumulation regions the MO_{830,830} values are lower, resulting in a slower densification and higher FAC values. The fresh snow densities of FDM v1.2A are on average lower (19.1 kg m⁻³, Fig. 2), which contributes to FAC values being higher on average. On top of that, increased snowfall accumulation in the interior and increased snowmelt along the coastal margins in RACMO2.3p2 compared to RACMO2.3p1 (Van Wessem et al., 2018), also contributes to the general pattern of higher FAC in the interior and lower values along the coastal margins. However, in the highly elevated and lowest accumulation regions, we find a reduction in FAC, which implies that the MO₈₃₀ update outweighs the effects of higher accumulation and lower fresh snow density in this region.

Figures 5c and d focus on temporal variability in FAC. Figure 5c shows the peak-to-peak FAC difference, that which represents the difference between the highest and the lowest FAC value occurring during the study period. Large values indicate that seasonal and interannual climate variability cause large temporal variations in FAC. (1979-2020). The peak-to-peak FAC difference is low in the interior, less than 0.1 m, despite the high FAC. Along the coastal margins, the peak-to-peak FAC difference is higher, up to 5 m, even though the FAC is generally lower. Therefore, the relative peak-to-peak FAC difference is low in the interior (<1 %) and high near the ice sheet margins (>5 %) (Fig. 5d). The average relative peak-to-peak FAC difference for the ice shelves, excluding the two largest ones, the Ross and Ronne-Filchner ice shelves, amounts to 13 %. The This high peak-to-peak variability is caused by a combination of significant substantial temporal variability in accumulation and melt. The temporal firm FAC variability and the driving processes are discussed further in Sections 4.3 and 4.4 below.

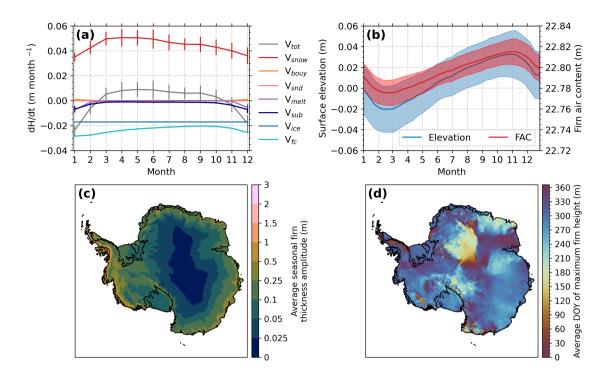


Figure 6. Ice sheet averaged seasonal cycle from FDM v1.2A of the (a) vertical firn surface velocity including its separate components from Eq. 6, and of (b) surface elevation ehange and firn air content (FAC). The bars and colored areas indicate the interannual standard deviations. Maps from FDM v1.2A (c) of the average seasonal firn thickness peak-to-peak difference and (d) of timing average day of the year of maximum firn heightphase maximum.

4.3 Seasonal firm air content and firm height-depth variability

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In order to better understand the FAC temporal variability and its drivers, we decompose the signal into seasonal and interannual contributions. Figure 6a shows the average seasonal cycle of the vertical firn surface velocity ($V_{tot}v_{tot}$) including its components from Eq. (6). Four main processes drive the seasonality of the firn thickness: snowfall ($V_{snow}v_{snow}$), firn densification ($V_{fe}v_{fc}$), melt ($V_{mell}v_{mell}$) and sublimation (V_{sub}). Snowfall v_{sub}). Averaged over the ice sheet, snowfall is highest in winter, while firn densification, melt and sublimation peak in summer. The net result is that the firn thickness and firn air content FAC steadily increase over the winter, due to enhanced accumulation and reduced firn densification, melt and sublimation (Fig. 6b). This is followed by a sharp decrease over the summer due to reduced accumulation and increased firn densification and melt. The firn height change by ice flux divergence ($V_{tce}v_{tce}$) is constant by definition, while the continental average impact AIS average impacts of snowdrift ($V_{snd}v_{snd}$) and buoyancy (V_{buoy}) is negligible v_{by}) are small (<1 %). The vertical bars in Figure 6a indicate the interannual variability, which is driven solely by snowfall except for the two summer months, which also have contributions from sublimation and melt.

The average seasonal amplitudes of the ice-sheet-wide firn thickness and FAC, defined as half the peak-to-peak values of the average seasonal cycle, are 3.5 and 2.4 are 3.1 and 2.1 cm, which amounts to Antarctic wide integrated volumes of 463 and 304 428 and 307 km³, respectively. This agrees with the modelled firn thickness amplitude of 3.5 3 cm by Medley et al. (2020) found by Medley et al. (2022). The average seasonal peak-to-peak values of firn thickness and FAC, defined as the average difference between the highest and the lowest value of each year, of all local seasonal amplitudes are considerably higher, 13.5 and 8.5 6.8 and 4.2 cm, amounting to volume changes of 1784 and 1123-892 and 562 km³, respectively (Fig. 6c). The difference between the seasonal amplitudes and peak-to-peak values is due to interannual variations in the reason is that there is a spatial variation in timing of the seasonal maximum and minimum. (Fig. 6d). The difference between the firn thickness and FAC seasonal cycles suggests that 63 to 68 (6.8 vs 4.2 cm) suggests that around 62 % of the seasonal surface elevations elevation fluctuations are caused by a change in air content rather than actual mass change. It This result implies that not including the seasonal cycles of FAC in altimetry studies can generate significant biases, which emphasizes the importance of using firn densification models, errors, especially when investigating shorter periods and changes in subannual periods and in regions with large seasonal SMB cycles.

Figure 6c shows that the seasonal peak-to-peak firn thickness change firn thickness amplitude can be up to 3 m in the western Antarctic Peninsula, caused by a rare combination of substantial snowfall and high snowfall and melt. In regions with little melt, such as coastal West Antarctica, the values are typically 1 m. In the interior interior East Antarctica, values of less than 2.5 cm are found, as the annual average accumulation is low and melt absent. Figure 6d shows the average phase of the firn thickness maximum day of the year of the maximum firn height, which generally occurs during late winter or spring, before the summer decrease in accumulation and increase in temperature. In some regions accumulation peaks and firn depth peak in late summer or early winter, e.g., along the Weddell Sea coast, in Enderby Land, Adélie Land and parts of the interior. Here, the the maximum firn depth occurs in late summer or early winter.

485 4.4 Interannual firm air content and firm height depth variability

Figure 7 shows time series of AIS and above/below 2,000 m a.s.l. integrated FAC and the cumulative surface temperature and vertical firn surface velocity anomalies (against the long-term mean) including its components (Eq. 6). The figure confirms that the seasonal firn thickness and FAC variability is are driven by snowfall, densification, melt and sublimation resulting in a firn thickness peak in late winter. The figure also reveals decadal firn thickness and FAC variability mainly resulting from fluctuations in snowfall (46 % of the total cumulative dH anomaly), and that firn densification, despite the long time scale, reduces these snowfall-induced fluctuations by about 15 % . Generally, we observed that periods of significant snowfall and firn thickness fluctuations last 5 to 10 years. For example, from 2005 to 2015, decreased snowfall resulted in decreased firn thickness and FAC20 % (calculated from difference in standard deviation (sd) between the cumulative anomalies of v_{snow} and $v_{snow+vfc}$. We also find significant-interannual variability in melt, but snowfall variability dominates. Below 2,000 m a.s.l. the variability in firn thickness and FAC are substantially larger, which can be attributed to the higher accumulation (70 % of the total accumulation falls on only 40 % of the ice sheet), in combination with variability of larger temporal variability in snowmelt and firn densification. Above 2,000 m a.s.l. the variability relative contribution of firn densification is also relatively

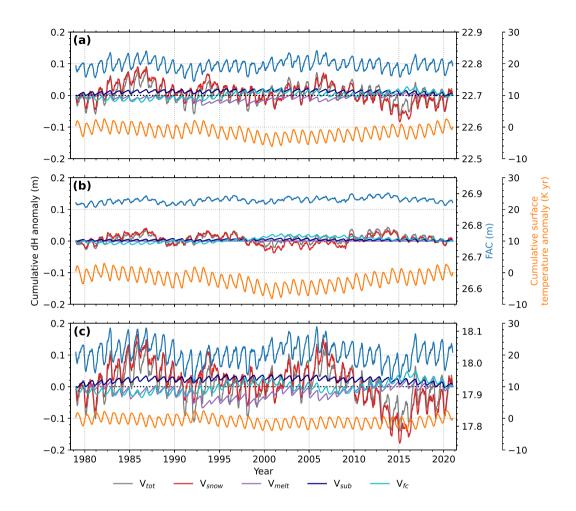


Figure 7. Time series from FDM v1.2A of FACand, of the cumulative anomalies (<u>against long-term mean</u>) of surface temperatureand, of the vertical firm surface velocityincluding, and of its separate components from Eq. (6). Time series are shown for (a) the entire ice sheet, (b) the part of the ice sheet situated above 2,000 m a.s.l. and (c) the part of the ice sheet situated below 2,000 m a.s.l.

high, which to the total cumulative dH anomaly is larger (32 %) than below 2,000 m a.s.l., where this is only 14 %. This difference can partly be explained by the larger temperature variability in this area interannual temperature variability above 2,000 m a.s.l (sd in annual means = 0.78 K compared to 0.48 K) and the absence of melt. The interannual variability in snowfall and temperature are related to low-low and mid-latitude climate oscillations such as the El Nino-Southern Oscillation and the Southern Hemisphere Annular Mode (Kwok and Comiso, 2002; Marshall et al., 2017). Figure 7 shows that the magnitude of the decadal FAC variability (sd = 1.8 cm) constitutes a large fraction (67 %) of the total surface elevation change (sd = 2.7 cm). This reiterates the necessity of the common practice to remove decadal variability in FAC from altimetry signals to avoid errors in mass change estimates.

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5 Comparison with altimetry and FDM v1.1 surface elevation change

In sections 3.1, 3.2 and 4 we have shown that the model captures the strong presented and evaluated the spatial variation in firn thickness and density. To also evaluate the temporal variation, we compare our simulated elevation trends and seasonal variability to multi-mission satellite altimetry observations reported by Schröder et al. (2019) for the period 1992-2015. Our firn simulations only provide elevation change due to anomalies in firn thickness caused by variations in e.g. snow accumulation, surface temperature, firn densification and melt. The difference between the observed and simulated elevation change therefore represents a combination of elevation changes due to ice-dynamical imbalance, errors in the firn model and its climate forcing and errors in the altimetry observations. For this comparison, we therefore exclude regions with ice-dynamical imbalance, by using the ice-dynamical imbalance mask of Shepherd et al. (2019), which mainly consists of Pine Island, Thwaites, Totten and Getz glacier basins and the Kamb ice stream.

5.1 Seasonal amplitude

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Ligtenberg et al. (2012) compared satellite altimetry and FDM v1.1p1 surface height seasonal variability. They found that FDM v1.1p1 patterns qualitatively agree with low variability in the interior and high variability near the margins (Fig. 6e), and that the significantly lower seasonal variability of FDM v1.1p1 compared to satellite altimetry can likely be explained by altimetry errors. The altimetry observations prior to 2003 exhibit a larger short-term variability as a result of the lower measurement precision of the satellites during that period (Schröder et al., 2019; Nilsson et al., 2022), and are excluded from the analysis seasonal amplitude comparison here. To enable comparison with the FDM v1.1p1 simulation, we focus on the period 2003-2015. To reduce the impact of the long-term trends we remove a linear trend from all datasets. After doing so, we find that the seasonal amplitude averaged over the ice sheet of the altimetry observations amounts to 5.2 cm, of the FDM v1.2A simulations to 3.1 cm (60 % of the observed amplitude) and of the FDM v1.1p1 simulations to 2.5 cm (48 % of the observed amplitude). The performance of the new model thus appears to represent an improvement, which higher seasonal amplitude of FDM v1.2A results from increased snowmelt along the coastal margins of in RACMO2.3p2 compared to RACMO2.3p1 in combination with a lower fresh snow density (Fig. 2). Using updated corrections and a normalization of seasonal amplitudes from different missions, Nilsson et al. (2022) found that seasonal amplitudes from 1992-2016 are 47 % smaller than the ones from Schröder et al. (2019), which compares well with the FDM seasonal amplitudes.

5.2 Decadal variability

In Figure 8 we compare observed (Fig. 8a) and simulated 2003-2015 average surface elevation changesurface elevation trends. To obtain these numbers, a linear regression is used fitted to monthly values for the period 2003-2015. The FDM v1.2A surface elevation change trend over this period is positive in Dronning Maud Land, Enderby Land, and negative in most other regions (Fig. 8b). FDM v1.1p1 (Fig. 8c) has roughly similar patterns as FDM v1.2A, but the regional amplitudes magnitudes differ. Figure 8d shows the residual surface elevation change trend of FDM v1.2A, which is calculated as the difference between the altimetry and by subtracting the FDM v1.2A elevation change trend from the altimetry trend, and thus combines unresolved

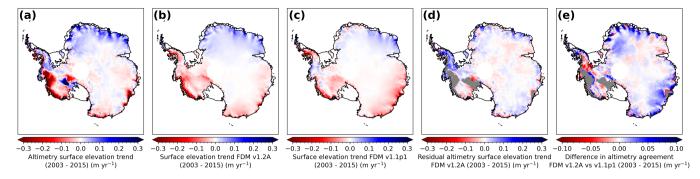


Figure 8. Maps of average trends in surface elevation change over the period 2003-2015 from (a) multi-mission altimetry (Schröder et al., 2019), (b) FDM v1.2A and (c) FDM v1.1p1 obtained using linear regression. (d) Residual trend of altimetry minus FDM v1.2Aminus altimetry. (e) Improvement of the Difference in altimetry agreement between FDM v1.2A trend compared to and FDM v1.1p1. (absolute residual of FDM v1.1p1 compared to altimetry minus absolute residual of FDM v1.2A compared to altimetry). Regions within the ice-dynamical imbalance mask reported by Shepherd et al. (2019) are excluded in (d) and (e) and indicated by the grey shaded area.

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trends and the errors in the firm model and altimetry. In Figure 8e we subtracted the absolute residual surface elevation change of show the difference in altimetry agreement between FDM v1.2A and of FDM v1.1p1from each other, which yields the improvement, which is calculated by subtracting the absolute residual surface elevation trend compared to altimetry (Fig. 8d) of FDM v1.1p1 from the similary dervied absolute residual of FDM v1.2A. The blue areas indicate an improvement in altimetry agreement of FDM v1.2A compared to FDM v1.1p1. In Figures 8d and e, regions with ice-dynamical imbalance are excluded. The surface elevation change The average absolute residual trend of FDM v1.2A has improved in most regions, especially compared to altimetry has been reduced by 17 % compared to FDM v1.1p1 (from 2.6 to 2.1 cm yr⁻¹). The improvement is most notable in Dronning Maud Land, Wilkins-Wilkes Land and Adélie Land, where. In Dronning Maud Land the FDM v1.2A trend has become either more positive rless negative. The residual absolute trend of FDM v1.2A has been reduced by 17 % compared to FDM v1.1p1 (from 2.6 to 2.1 more positive, and in Wilkes Land and Adélie Land the trend has become less negative (Fig. 8b,c, for region names see Fig. 1). In the Antarctic Peninsula and Ellsworth Land the altimetry agreement of FDM v1.2A with altimetry has reduced compared to FDM v1.1p1. In these regions, a substantial residual positive trends remains (>10 cm yr⁻¹) (Fig. 8d), which might could be due to a long-term increase in accumulation and/or ice-dynamical thickening. Smith et al. (2020) used ICESat data to show that the mass gain is mainly located along the steep slopes of the Antarctic Peninsula and decreases with distance from the ocean, which is indicative of an increase in accumulation snow accumulation. The increase in accumulation on centennial time scales in these regions (e.g., Thomas et al., 2015, 2017). This (e.g., Thomas et al., 2015, 2017; Medley and Thomas, 2019) implies that the actual firn column is not in balance with the 1979-2020 climate, and that the accumulation used for our as assumed in the FDM spin-up period, which we assumed to be equal to 1979-2020 average accumulation, is overestimated in these regions procedure. As a result, the vertical downward ice flow (Eq. 6) is overestimated, which results in underestimated surface elevation change. This is further discussed in the next paragraph.

In Figure 9 we compare time series of altimetry observed elevation change over the period 1992-2015 to simulated elevation change for eleven representative locations across the AIS. These locations were selected as they cover the main distinct patterns from Figure 8 and have continuous observations. The time series have been smoothed using a 6-monthly moving average 560 window. At most locations, FDM, in order to reduce the impact of seasonal variations in radar penetration depth. The red shaded area is the aggregated uncertainty range from the sensitivity analysis, which is primarily determined by uncertainty in the climate forcing during the spin-up period (see next section for details of the sensitivity analysis). The agreement of FDM v1.2A shows comparable patterns to the observations, and the agreement seems to have been improved. A compared to FDM v1.1p1. This improvement has improved ($R^2 = 0.58$ compared to 0.36), which is mainly due to changes in the RACMO2.3p2 565 forcing, given the strong similarity between elevation change of FDM v1.2A and FDM v1.1p2. As in Figure 8, we find a substantial residual altimetry trend (>10) at location 5 (Antarctic Peninsula). At locations 1 (Dronning Maud Land) and 6 (Ellsworth Land) we find a minor residual altimetry trend (~5). The altimetry observations prior to 2003 exhibit relatively stronger short-term variability (+9 % sd), likely related to the measurement imprecision compared to after 2003), in line with 570 the smaller measurement precision (Schröder et al., 2019). The agreement with FDM v1.2A also appears to increase increases after 2003 $-(R^2 = 0.93)$ compared to 0.64, RMSE recued by 49 %), however the altimetry variability remains higher than the simulated variability (+13 % sd)-compared to FDM v1.2A), as these more recent observations still contain noise (<11 cm over flat terrain) and have variations in radar penetration depth (Schröder et al., 2019), while on the other hand, it can also partly be related to errors in the firn model. The uncertainty bands in Figure 9 show that a large part (73 %) of the difference between FDM v1.2A and altimetry falls within the uncertainty range. As in Figure 8d, we find a substantial positive residual altimetry 575 trend (~ 14 and 5 cm yr⁻¹) at location 5 (Antarctic Peninsula) and 6 (Ellsworth Land). The upper bound of the uncertainty for those two locations is determined by the sensitivity tests in which the climate during the spin-up period was colder and drier (spin-up T_c, b experiment), indicated by the red dotted lines in Figure 9. With this presumption, the trend difference for location (5) greatly reduces and largely disappears for location (6).

6 Uncertainty Sensitivity analysis

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We investigated the impacts of uncertainties in the elimate forcing, model densification expressions and model densification rate expression, fresh snow densityon our final FAC, z₅₅₀, and z₈₃₀ results, climate forcing, and climate forcing during the spin-up period, on the final FAC and surface elevation change results for (i) the 106 sample locations and (Table 3). Note that, if we initially would have had a different forcing and fresh snow density, we would also have obtained different MO fits (Eq. 4 and 5). Therefore, we performed additional model runs in which we also adjusted the MO fits. With the original MO fits, ii) expanded to the entire ice sheet (Table 5). For the sample locations, the FAC responds rather linearly to the imposed changes, with the highest sensitivity to the temperature uncertainties sensitivities with updated MO fits are lower than without updated MO fits, especially the FAC uncertainties, which are 41 % lower. The results for the sample locations with updated MO fits are then expanded to the entire ice sheet. Some differences between the ice-sheet-wide and sample locations exist, which can be explained by the sample locations being biased towards high accumulation, which is needed to obtain optimal empirical

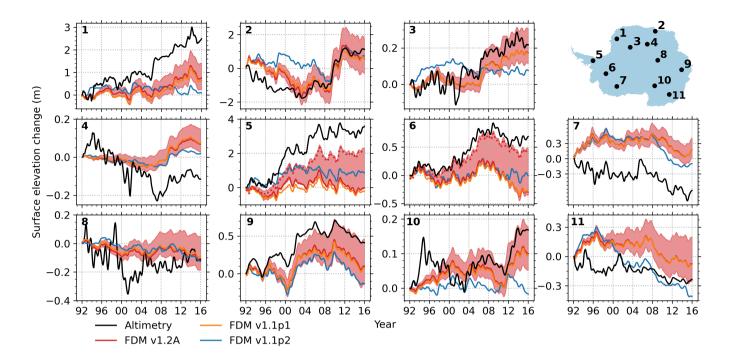


Figure 9. Time series of the altimetry observed observations and modelled elevation change for eleven locations over the period 1992-2015. The time series have been smoothed using a 6-monthly moving average window. The modelled elevation change includes FDM v1.2A, FDM v1.1p1 and FDM v1.1p2. The number of each panel corresponds to the numbered locations on the maps. The red shaded areas indicate the total uncertainty from the sensitivity analysis. The red dotted lines in panels (5) and (6) indicate the surface elevation change from the spin-up T_s , b experiment.

relations for expanding the sample locations uncertainties. The average simulated FAC, for the sample locations and entire ice sheet (both with updated MO fits), is most sensitive to the uncertainties in the MO fits (5.1 and 5.2 %), temperature forcing (1.8 and 2.3 %) and accumulation forcing (4 and 3.7 %), and least sensitive to uncertainties in fresh snow density (0.7 and 1.2 %), and the spin-up forcing (ranging from 0.5 to 0.7 %). The simulated surface elevation change, for the sample locations and the entire ice sheet, is most sensitive to uncertainties in the accumulation during the spin-up (+/-6 %)22.4 and the lowest sensitivity to the accumulation uncertainties 19.0 mm yr⁻¹), followed by the accumulation forcing (8.0 and 5.6 mm yr⁻¹), and temperature during the spin-up forcing (5.2 and 4.4 mm yr⁻¹). As the uncertainties in temperature and accumulation during the spin-up period have opposite effects, the tests in which accumulation and temperature were adjusted simultaneously yield an surface elevation trends of of 17.4 and 14.9 mm yr⁻¹. The simulated surface elevation change is least sensitive to uncertainties in the MO fits (+0.7/-2 %)0.8 and 0.5 mm yr⁻¹), fresh snow density (2.1 and 1.5 mm yr⁻¹) and temperature forcing (1.5 and 0.7 mm yr⁻¹). The RMSE of the z_{550} and z_{830} somewhat increase. The FAC and RMSE of z_{550} and z_{830} sensitivity with the updated MO fits are substantially lower. These results suggest that our FAC, z_{550} and z_{830} estimates are robust.

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Table 5. Overview of the relative change in annual average firn air content (FAC) and surface elevation change (dH/dt) of the ice-sheet-wide estimates and sample locations experiments. The sample location experiments also include the experiments without adjusted MO fits. An overview of the experiments including the prescribed uncertainties are listed in Table 2.

Experiment name	ice-sheet-wide e	stimates	Sample locations (w	Sample locations (without adjusted MO fits) ^a		
	FAC (%)	$dH/dt (mm yr^{-1})$	FAC (%)	$dH/dt (mm yr^{-1})$		
MO fits	5.1	0.5	5.2	0.7		
$ ho_s$	1.2	1.5	0.7 (5.8)	2.1 (2.4)		
T_s	1.8	0.7	2.3 (4.9)	0.7 (1.5)		
b	4.0	5.6	3.7 (6.8)	7.8 (8.0)		
spin-up T _s	0.5	4.4	0.5 (1.5)	5.2 (5.2)		
spin-up b	0.7	19.0	0.7 (1.1)	22.4 (22.4)		
spin-up T_s , b	0.7	14.9	0.6 (0.8)	17.4 (17.4)		
$total^b$	6.9	16.0	6.8 (11.5)	19.2 (19.4)		

^aThe values between brackets indicate the uncertainties of the experiments without updated MO fits. ^bFor the spin-up experiments only the combined temperature and accumulation experiments are included.

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As the model set up of our reference simulation assumes a steady-state firn layer, there is no net surface elevation trend in this simulation over the period 1979-2020. In the sensitivity tests in which the The spatial patterns of the total ice-sheet-wide FAC and surface elevation change uncertainties are shown in Figure 10, and for all experiments individually in Figures S4 and S5. The total FAC uncertainty, dominated by the MO fits, accumulation and temperature during the uncertainties, is high (>10 %) along the high-accumulation coastal margins and in part of the interior, especially near the Transantarctic Mountains. The surface elevation change uncertainty is dominated by the spin-up period are adjusted, the firn layer is not in steady state anymore, which results in surface elevation trends. The average surface elevation trend over the period 1979-2020 of the accumulation test amounts to +and accumulation uncertainty, and mainly follows the accumulation pattern, with high values (>25 mm yr⁻¹) along the high-accumulation coastal margins, especially in the Antarctic Peninsula, and in the Transantarctic Mountains.

We show time series of the uncertainty in surface elevation change and FAC at the sample locations in Figure S7a for all experiments, and in Figure S7b for the non-spin-up experiments. First, we discuss uncertainties in surface elevation change for non-spin-up and spin-up experiments. Next, we look at uncertainties in FAC. In non-spin-up experiments, the uncertainty is only 4 (1.72 in total) and of the temperature to -0.7 (-0.3 in total). As the temperature and accumulation have opposite effects, the test in which accumulation and temperature were adjusted simultaneously yields an average surface elevation trend of +3.3 (1.36 in total). From linear regression between the % of the total surface elevation change. This uncertainty is smaller than the 40 % that can be derived from Table 3. This is because the imposed uncertainties have an amplifying effect. For example, lowering the fresh-snow density increases surface elevation change rates of both signs. However, in Table 3 the uncertainty in surface elevation change is calculated over a short 5-year period for each location separately. Spatial and temporal variability

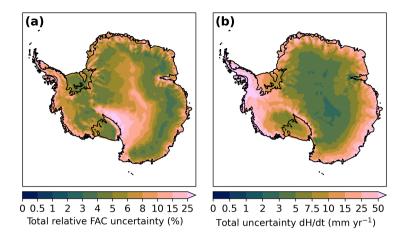


Figure 10. Maps from FDM v1.2A of the estimated ice-sheet-wide total uncertainty in (a) firm air content (FAC) and (b) surface elevation change (dH/dt) from the sensitivity analysis.

in sign of elevation change tend to cancel out the uncertainties. In contrast, the uncertainty in the spin-up experiments gives a spatially and temporally uniform response. This paves the way to expand ice-sheet-wide estimates of uncertainty into time series, like in Figure S7c.

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For FAC, its ice-sheet-wide change is only 19% of the uncertainty in surface elevation trend of the combined accumulation and temperature test and the annual average accumulation, we obtain a relation of 0.05 total surface elevation changeper 100 accumulation (R² = 0.94). This means that in locations with, similar to the 23% for the suite of sample locations. It indicates that imposing an uncertainty in the experiments mostly has an impact on the high-density, low-FAC part of the firm column. The spatial pattern in FAC change uncertainty for the spin-up experiments is different from that of surface elevation uncertainty (Fig. S6), with a decreasing uncertainty for high-accumulation regions (>1000 annual average accumulation the elevation increases by >5 over the period 1979-2020. If we focus on the period that we used for 600 mm w.e. yr⁻¹). We attribute this to the fact that in high-accumulation regions, the imposed uncertainty has an effect on the entire firm layer much more quickly. In contrast, in the non-spin-up experiments, the magnitude of the FAC uncertainty is similar (~85%) to the uncertainty of surface elevation change. The finding that imposing an uncertainty has its main impact on deeper, higher-density firm, means that the impact on a volume to mass conversion is limited. On the other hand, the large uncertainties in the remote sensing comparison in Figure 9 (1992-2015) we find that the surface elevation trends increase by +3.9 (+0.93 in total) for the accumulation test, decrease by -0.8 (-0.18 in total) for the temperature test and increase by +3.2 (0.77 in total) for the combined surface elevation change compromise our ability to separate ice-dynamical from SMB and firm-related surface elevation changes.

In addition, in most places except the Antarctic Peninsula and Ellsworth Land, the long-term trends in accumulation and temperature test. We conclude that FDM is most sensitive to the uncertainty in are often weak, not significant, and variable in time and space. It suggests that the accumulation during the ice-sheet-wide uncertainty due to uncertainty in the spin-up period. Long-term increases in accumulation are mainly found in climate is lower than suggested in Figure S7c. For the

Antarctic Peninsula, Ellsworth Landand to some extent in Dronning Maud Land, and for some regionsover some periods even reduced accumulation has been found (Thomas et al., 2015, 2017). The long-term increases in accumulation partly explains the discrepancy between FDM and altimetry observations for these regions as is discussed in Section 5.2. and Ellsworth Land, accumulation and temperature has increased strongly during the 20th century. To evaluate our results from the spin-up sensitivity tests in these two regions, we used surface elevation trends from altimetry observations (Schröder et al., 2019) (Fig. S8). We found that absolute residual trend altimetry minus the spin-up sensitivity test was reduced by 25 % compared to the reference FDM v1.2A (from 6.1 to 4.5 cm yr⁻¹). When we leave out pixels with a z_{830} age < 42 years (6% of the pixels), where we assume the underlying ice to have responded to the changing climatic conditions, the residual trend was reduced by 38 % (from 5.6 to 3.5 cm yr⁻¹).

7 Remaining limitations and outlook

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The comparison with in situ and remote sensing measurements and the sensitivity analysis showed that FDM in general performs well and that our results are robust. A difficulty of the data to model comparison is that

7 Implications, remaining limitations, and outlook

In this study, the fresh snow density and firn compaction parameterizations of IMAU-FDM have been improved, and an updated atmospheric forcing has been used. The resulting difference in firn changes over time of FDM v1.2A compared to FDM v1.1p1 has an impact on conversion of volume-to-mass changes in altimetry studies. Over the period 2003-2019, (Smith et al., 2020) found that the total mass change of East Antarctica amounts to 195 Gt yr⁻¹, and of West Antarctica to -245 Gt yr⁻¹. When FDM v1.2A is used instead of FDM v1.1p1, the mass change over the period 2003-2015 for East Antarctica is 7.3 Gt yr⁻¹ lower, and for West Antarctica this is 1.2 Gt yr⁻¹ higher. These temporal firn differences are predominantly caused by the updated forcing as is shown by Figure 9. In addition, Figures 2 and 3 show that updates in the fresh snow density and firn densification parameterizations have substantial impact on the time averaged firn density profiles. These outputs are required to assess the meltwater buffering capacity of the firn layer, and to convert column thickness to ice thickness, to calculate solid ice discharge over the grounding line (Rignot et al., 2019)

The comparison with satellite altimetry measurements are used to asses the performance of FDM. However, especially in the interior, both the long-term trend and seasonal variability are small compared to the error range of the altimetry observations, which explains part of the remaining discrepancy that exist exists there, but also hampers the comparison (Verjans et al., 2021). More accurate Accurate in situ measurements of surface height evolution and firn densification are sparse. Future evaluation of firn models will therefore likely continue dependent to depend on satellite altimetry comparison. The ICESat-2 ATL15 (Smith et al., 2021) is a precise Antarctic-wide 3-month height change laser product that is available from May 2018 onwards, thereby providing an opportunity to assess seasonal and decadal variations, especially in the coming years when more data will become

available. Smith et al. (2022) already used ICESat-2 measurements to evaluate regional climate and firn densification models in Greenland.

An important limitation of IMAU-FDM, The sensitivity analysis shows that updating the MO fits reduces the average FAC sensitivity to uncertainties in fresh snow density and climatic forcing by 41 % (Table 5), which implies that the MO factors, partly correct for biases in these boundary conditions. Thus, application of FDM v1.2A calibrated MO fits when using a different fresh snow density or climate forcing is not recommended. The fresh snow density parameterization is also dependent on the climate conditions derived from RACMO2.3p2, so any biases will impact the updates of the coefficients. In addition, the simulated fresh snow density accounts for the wind-driven compaction, and the updated coefficients are thus not recommend to use in combination with a firn model that includes wind-driven densification of the upper firn layers.

Another limitation, which is evident from both the remote sensing comparison and the sensitivity analysis, is the assumption of a steady-state firn layer, which may be a poor assumption in e.g., the Antarctic Peninsula and Ellsworth Land. The steady increase in accumulation over the last centuries in these regions results in a positive surface elevation trend in these regions which can not be captured by the model. Correcting for this accumulation trend by estimating it over the period with reliable climate forcing (1979-present). In our sensitivity analysis we show that correcting for this by estimating previous climatic conditions from ice core data, could be a next step to improve this aspect of firn models reduces the difference between simulated and altimetry observed surface elevation trends by 38 % over the Antarctic Peninsula and Ellsworth Land. Future work could focus on further accounting for pre-1979 climatic conditions.

8 Summary and conclusions

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In this study, we present and evaluate a simulation of the contemporary characteristics of the Antarctic firn layer using the updated IMAU-FDM (Firn Densification Model), version v1.2A, for the period 1979-2020. In this new version, the fresh snow density and firn compaction parameterizations have been improved, and an updated atmospheric forcing has been used. The RMSE when compared to observations of the fresh snow density and firn thickness have been reduced by 26 and 23-28 and 36 %, respectively, compared to the previous parameterizations (FDM FS-K and FDM v1.2A-log). The RMSE of FAC is 21 % lower compared to FDM v1.1p1. In general, the modelled fresh snow densities have decreased, especially in high-accumulation high-accumulation areas and to a lesser extent in windy escarpment regions. Firn thickness and FAC have decreased in lowand high-accumulation high-accumulation areas, whereas it has they have increased in intermediate accumulation areas.

In accordance with observations, the firn thickness and density patterns vary spatially across climatic regions with temperature as a primary and accumulation and surface melt as a secondary driverdrivers. Along the coastal margins, we found that the firn thickness and firn air content FAC vary strongly in time, whereas in the interior the variations are small. The temporal variations can be split into a rather stable seasonal cycle driven by snowfall, compaction, summer melt and sublimation, and more irregular decadal variations mainly driven by slow snowfall anomalies variations. As variations in firn air content FAC and firn thickness alignhaye a similar phase, 63 to 68 % of the seasonal and decadal surface height variability is due to variations

in firn air content FAC rather than firn mass, which emphasizes the importance of correcting for seasonal and decadal firn air content FAC variations in altimetry studies.

Comparison of simulated surface elevation change with altimetry shows that in most regions the modelled and observed trends agree reasonably well. The confirms that the performance of the updated model has improved (a 17 % reduction in error), notably in Dronning Maud Land and Wilkins-Wilkes Land, which is predominantly due to updated climate forcing. However, a significant trend difference substantial trend difference (>10 cm yr⁻¹) remains in the Antarctic Peninsula and Ellsworth Land, which is likely caused by increasing accumulation over the past centuries, that which violates the model assumption of a steady state over the simulated period (1979-2020). The altimetry comparison also suggests that the amplitude of the modelled seasonal cycle has somewhat improved compared to the previous model version. The sensitivity analysis shows that our model in generalis robust, and that it is most sensitive to Correcting for this over the Antarctic Peninsula and Ellsworth Land, by estimating previous climatic conditions from ice core data, substantially improves this (an error reduction of 38 %). In general, uncertainties in the accumulation model formulation, forcing or forcing during the spin-up period. We conclude that the improved model captures the strong spatial and temporal variation in firn conditions in Antarctica that have been previously observed in firn cores and satellite altimetry cause rather small changes in simulated FAC (<5.2 %), while uncertainties in the accumulation forcing, and in the accumulation and temperature forcing during the spin-up period can lead to substantial differences in simulated surface elevation change (4 to 19 mm yr⁻¹).

Code availability. The code of IMAU-FDM v1.2A is available on GitHub at https://github.com/brils001/IMAU-FDM (last access: 23 November 2022).

Data availability. A list of all firn cores used and corresponding references can be found in the Table S1 in the Supplement. IMAU-FDM data can be obtained from the authors without conditions.

Author contributions. SV, WJvdB, MB, PKM and MRvdB defined the research goals and designed the study. SV improved the model, performed the simulations and analyzed the results. All authors contributed to discussions on the manuscript.

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