Sensitivity of basal conditions in an inverse model: Vestfonna Ice-Cap, Nordaustlandet/Svalbard

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Abstract

The dynamics of Vestfonna ice-cap (Svalbard) is dominated by fast-flowing outlet glaciers. Its mass balance is poorly known and affected dynamically by these fast flowing outlet glaciers. Hence it is a challenging target for ice-flow modeling. Precise knowledge of the basal conditions and implementation of a good sliding law are crucial for the modeling of this ice-cap. Here we use the Full-Stokes finite element code Elmer/Ice to model the 3-D flow over the whole ice-cap. We use a Robin inverse method to infer the basal friction from the surface velocities observed in 1995. Our results illustrate the importance of the basal friction parameter in reproducing observed velocity fields. We also show the importance of having variable basal friction as given by the inverse method to reproduce the velocity fields of each outlet glacier – a simple parameterization of basal friction cannot give realistic velocities in a forward model. We study the robustness and sensitivity of this method with respect to different parameters (mesh characteristics, ice temperature, errors in topographic and velocity data). The uncertainty in the observational parameters and input data proved to be sufficiently small not to adversely affect the fidelity of the model.

1 Introduction

Svalbard is an Arctic archipelago at around 80° N. Nordaustlandet is the second largest island in this archipelago and has two major ice-caps: Austfonna and Vestfonna. These ice-caps represent one of the largest ice-covered areas in the Eurasian Arctic. The Vestfonna ice-cap is the smaller of the two ice-caps and covers about 2400 km² and ranges in altitude from sea-level to 630 m a.s.l. It is composed of a small, but relatively smooth and flat area in the center (Pohjola et al., 2011b; Dowdeswell, 1986; Dowdeswell and Collin, 1990). A larger fraction is occupied by fast flowing outlet glaciers (Figs. 1 and 2; Pohjola et al., 2011b). Some of these have been suggested to be of surge-type (Hagen et al., 1993). A large part of the total ice-flux towards the margin is through these spatially limited areas of fast-flow units.
The role of glaciers as indicator of climate change, as well as their contribution to sea level rise is widely acknowledged (Moore et al., 2011). The fast-flow units and surge type glaciers are especially important in that context, but are however the most challenging glaciers to model. The high velocities are mainly achieved through basal motion (Clarke, 1987), as ice deformation in regions where the bed is at the melting point limits velocities to a range of 1–10 m yr\(^{-1}\). If unfrozen and wet basal conditions are maintained, stable fastflow of outlet glaciers occurs. In the case of evolving basal conditions, acceleration and deceleration will occur in response to changes in the release of frictional heat and water production (Bougainmont et al., 2011). The basal conditions are thus crucial for the observed flow-regime as well as possible changes or even surges.

Here, we use the Full-Stokes model Elmer/Ice (Gagliardini and Zwinger, 2008; Zwinger and Moore, 2009) to study in detail the conditions at the bottom of Vestfonna (VSF) in 1995. Determining the optimal basal conditions from the glacier topography and the surface velocities is an inverse problem. The focus of this paper is the application of an inverse method and its technical aspects rather than an interpretation or analysis of the obtained distribution. A key motivation of this paper is that the sensitivity of the inverse method to errors in input data or choice of model parameters such as mesh resolution may play a critical role in the derived basal properties. Hence we perform sensitivity tests.

Recently, several methods have been proposed to solve this particular inverse problem, for example Morlighem et al. (2010), Goldberg and Sergienko (2011) and Gudmundsson and Raymond (2008); Raymond and Gudmundsson (2009); Pralong and Gudmundsson (2011). Morlighem et al. (2010) study Pine Island Glacier, West Antarctica with a control method using an adjoint model. The spatial patterns of basal drag inferred are similar to those obtained with simpler models (Vieli and Payne, 2003; Joughin et al., 2009). Goldberg and Sergienko (2011) also use an adjoint model, while Gudmundsson and Raymond (2008); Raymond and Gudmundsson (2009); Pralong and Gudmundsson (2011) rely on a Bayesian method.
Here we use the variational approach with a Robin inverse method proposed by Arthern and Gudmundsson (2010); Gudmundsson (2011) and subsequently used in simulations of Variegated Glacier and the Greenland Ice Sheet (Jay-Allemand et al., 2011; Gillet-Chaulet et al., 2012). This algorithm can be implemented numerically without developing an adjoint model, but rather using the normal forward solution of the Navier-Stokes equations. We present the results obtained with the Robin method and compare them with results of a simpler approach: the use of mutually varying, but constant, sliding coefficients in different predefined areas. This emphasizes the role of the particular basal conditions at every outlet glacier of the ice cap.

In Sect. 2 the different data-sets used in this study will be presented in detail, Sect. 3 describes the ice-flow model as well as the inverse method. In Sect. 4.1 we discuss some previous modeling efforts with simple sliding laws to motivate the necessity of using a more sophisticated sliding law and from Sect. 4.2 on the results obtained for the basal conditions in 1995 are presented and discussed.

2 Data used in this study

2.1 Research site

Approximately 75% of Nordaustlandet is covered by glaciers (Hagen et al., 1993), most of these are of typical Svalbard polythermal type. Usually, marginal areas are cold and frozen to the bed, while thicker central parts are close to melting point. Varied topography of valleys, peaks, and resulting ice thickness changes produce the outlet glaciers characteristic of Vestfonna. The Vestfonna (VSF) ice-cap was the focus of a recent IPY project (Pohjola et al., 2011b, and references therein) with several glaciological fieldwork programs focused on establishing the mass balance, identifying flow dynamics, remote sensing, snow chemistry and paleo ice-sheet reconstruction. The Nordaustlandet ice-caps were also the target of an expedition in the International Geophysical Year (IGY) during 1957–1958 when ice-cap geometry, temperatures, mass-balance
and ice-thickness were measured (Schytt, 1964). More recently Austfonna has been visited by expeditions led by University of Oslo (Dunse et al., 2009, 2011b,a). The research area is presented in Fig. 1. Here we focus on the smaller ice cap (VSF) which is more topographically variable and has relatively more valleys and outlet glaciers than the neighboring Austfonna (ASF) ice-cap. A variety of fast outlets are concentrated on a small area on VSF (Fig. 2).

2.2 Topographic data

The rough surface topography (Fig. 1) of the ice cap shows clear ridges in its central parts: the main ridge is oriented East-West and another ridge extends northwards from the eastern part of the main ridge. Between the ridges and the coast, there are pronounced fast-flowing outlet glaciers (Fig. 2). Most of the outlet glaciers are orientated north-south. However, Franklinbreen, the largest outlet glacier, is situated on the northwestern side of the ice-cap and some smaller ones drain into northeastern fjords. Some of them have been suggested to be surge-type glaciers (Hagen et al., 1993). For example Braun et al. (2011) and Pohjola et al. (2011a) study the recent acceleration of the northwestern outlet glacier Franklinbreen.

The bedrock data used is presented in Petterson et al. (2011). The data-set consists of combined ground-based pulsed radar data collected in 2008–2009 and airborne radio-echo soundings data acquired in 1983 and 1986 over the ice-cap (Dowdeswell, 1986). The airborne radar did not give good bedrock depths in the thicker central parts of the ice-cap due to scattering and absorption losses in the ice, while the ground based radar was not deployed at the fast flowing outlet glaciers for safety reasons. The radar data were combined, and hence represent the ice-cap at different periods, however crossover analysis of the limited number of places where both surface and airborne radars registered bedrock indicates that the differences in ice thickness change over the period 1986–2008 was estimated as not more than 7 m over the whole ice-cap over the whole period, which is in agreement with a widespread thinning rate on VSF of 23 cm yr$^{-1}$ during the period 1996–2002 (Bamber et al., 2005). The bedrock data is
shown in Fig. 3. From this figure it is obvious that the south-western part of the DEM is poorly supported by actual data (Petterson et al., 2011). There is also a poorly covered region at the north-east coast, but since the bed topography is smoother here, this introduces less error in the data interpolation. Also, the radar profiles on the southern coast are aligned with the direction of flow of the outlet glaciers so that any bedrock ridges between the outlet glaciers may not be captured well in the DEM. The total areal fraction with a good spatial auto-correlation in the ice thickness interpolation was estimated to about 55% (Petterson et al., 2011). The rest of the ice-cap was mapped with lesser accuracy. The sub-glacial landscape undulates between elevations of −160 m and +410 m relative to sea-level at an the average ice-thickness of 186 m.

2.3 Surface velocities

Tandem Phase ERS-1/2 1-day InSAR scenes acquired between December 1995 and January 1996 are used to calculate the ice surface velocity structure of Vestfonna. In the following this data-set is referred to as “1995 velocities”. In Fig. 2 the total horizontal surface velocity is shown, in addition the directional information was also used in this study. In most cases errors are assumed to be smaller than about 2 cm per day (or 7 m per year). The displacement maps derived from satellite SAR data were geocoded to UTM projection at 100 m posting using the NPI DEM from 1990 (Pohjola et al., 2011a). We observe two very different flow regimes: a general pattern of very slow ice flow over the central area of the ice cap, with pronounced high velocity areas in the outlet valley glaciers.

2.4 Thermal regime

Svalbard has a maritime climate characterized by cooler summers and warmer winters than usually expected at this latitude. Mean monthly air temperatures on VSF during the short summer season do not exceed +3 °C and in the winter averages are between −10 and −15 °C reaching minimum values between −25 °C and −40 °C depending
on the year (Möller et al., 2011). At Kinnvika research station in the north-east of VSF winter monthly minimum temperatures were around −25 °C, and summer monthly maximum +4 °C (Pohjola et al., 2011b, and references therein).

The mean annual surface temperature $T_{surf}$ is estimated using the mean annual air temperature at sea-level $T_{sea}$ and an elevation lapse-rate $\gamma$

$$T_{surf}(x) = T_{sea} - \gamma \cdot S(x),$$  \hspace{1cm} (1)

where $S(x)$ is the surface elevation at the horizontal position $x$. We use a lapse-rate of $\gamma = 0.004 \text{ K m}^{-1}$ (indicated by Schuler et al., 2007; Wadham et al., 2006; Wadham and Nuttal, 2002). $T_{sea} = -7.7^\circ C$ is estimated from two weather stations on Austfonna during 2004 to 2008 (Schuler, 2007) as well as four weather stations during 2008 and 2009 on Vestfonna (Möller et al., 2011).

### 2.4.1 Geothermal heat-flux

Svalbard belongs to the Barents Sea shelf of the Eurasian plate (Labrousse et al., 2008). The tectonic history of the archipelago is complex and the rock outcrops of the coastal exposures of western Nordaustlandet formed during different geological eras. In southwestern Nordaustlandet Paleozoic strata rests on a Precambrian basement (Harland, 1997). Mesozoic dolerite intrusions dot the coastal exposures of the western part of the island. The basement of Svalbard consists of terraces assembled during the Caledonian orogeny in the Paleozoic (Labrousse et al., 2008). The basal ice temperature depends on the geothermal heat flux, which varies with the bedrock type and age (Waddington, 1987). A heat flow value of 63 mW m$^{-2}$, representative of the Post-Precambrian non-orogenic tectonic region (Lee, 1970) is used for Vestfonna in this paper.

### 2.4.2 Temperature profile in the ice

In Svalbard glaciers are typically polythermal: at higher elevation on the ice cap, the release of latent heat when seasonal meltwater percolates and refreezes in the firn...
pack creates an additional heat source leading typically to temperate ice covered by cold firn. An example of such a typical accumulation area temperature profile under the influence of refreezing meltwater is shown in Fig. 4, gray dots (Watanabe et al., 2001). The ice-core studied by Watanabe et al. (2001) was drilled at the east-end of the summit ridge in June 1995 to a depth of 210 m where the over-all ice-thickness is around 340 m (Petterson et al., 2011, Fig. 3). Theoretical temperature profiles based on the assumption of a steady state will lead to different shapes of the profile due to the differences in net mass balance in accumulation and ablation areas (Fig. 4, red and blue curve) (Paterson, 1994; Suter, 2002). In addition, in the accumulation area the profile is modified to become warmer in the upper part by the release of latent heat from refreezing of percolating melt water (Fig. 4, gray curve). At lower elevations in the ablation zone, the ice is usually below the freezing point, since solid ice prevents any percolation and refreezing of water. Thinner snow cover at lower elevations also allows efficient outgoing long-wave radiative cooling of the glacier during the long Arctic winters. The profile for an equilibrium line is given in the same figure (Fig. 4, pink line).

The temperature profile in VSF is poorly known and in this paper we will use two extreme cases to investigate the sensitivity of the model to variations in the temperature profile. Naturally, the assumption of temperate ice throughout the whole ice body is an upper, even though unrealistic limit for the ice-temperature. As simple estimation of the lower limit we use the steady state equilibrium line temperature profile (pink line in Fig. 4), this scenario will be referred to as “cold scenario” in the rest of the paper. In the ablation area the real temperature profile will be slightly warmer as a consequence of advection of warmer ice from deeper ice-layers. However our approximation will be relatively close to a typical ablation area temperature profile – the area we are most interested in. In the accumulation area this “cold scenario” profile may be strictly speaking a colder limit only for the upper part of the ice-column and slightly too warm in the lower part depending on the amount and penetration depth of latent heat. This temperature profile is applied in a de-coupled way from the velocity field.
By neglecting the advective heat transport, the equilibrium line temperature profile is determined by the surface temperature and the geothermal heat flux:

\[ T(x) = T_{\text{surf}} + \frac{\phi}{\kappa} \cdot D(x), \]  

with \( \phi \) the geothermal heat flux, \( \kappa \) the heat conductivity and \( D(x) \) the ice-depth (distance to the surface at a given location \( x \) in the ice-body). We use \( \kappa = 2.072 \text{ W K}^{-1} \text{ m}^{-1} \) (Ritz, 1987), \( \phi = 63 \text{ mW m}^{-2} \), \( \gamma = 0.004 \text{ K m}^{-1} \) and \( T_{\text{sea}} = -7.7^\circ\text{C} \) (see above) as parameters determining \( T_{\text{surf}} \). Temperatures in Eq. (2) are limited to values below pressure melting \( T = 0^\circ\text{C} \). We obtain bedrock temperatures between \(-8^\circ\text{C}\) near the coast and \(0^\circ\text{C}\) in the thick central parts of the ice-cap.

In summary we assume that our idealized temperature profile is a lower limit for temperatures and will furthermore show that the results do not differ significantly from what we obtain with an isothermal ice-body that corresponds to an (unrealistic) upper limit of possible temperatures.

3 Ice-flow model

The modeling of glaciers and ice sheets has a large range of applications, such as ice core dating or simulating the role of large ice masses on the climatic system. Whatever the objectives, numerical flow models seem to be the best way to capture the complexity of glacier evolution. The ice flow model solves the momentum and mass conservation equations for given initial and boundary conditions. The evolution of a glacier’s surface geometry, is then fully determined by the ice flow and the climatic forcing. For this it is crucial to understand the basal conditions and to use a suitable sliding law.
3.1 Forward ice-flow model

A right-handed \((O,x,y,z)\) Cartesian coordinate system is used. The surface and bedrock are defined as \(z = S(x,y,t)\) and \(z = B(x,y,t)\), respectively, but the bedrock will be assumed to be a fixed boundary so that \(B(x,y,t) = B(x,y)\). Also, in this work only diagnostically simulations are performed, so that it is \(S(x,y,t) = S(x,y)\). The ice-thickness is given by \(H = S - B\). \(S\) and \(B\) denote the whole surface and bedrock layer, respectively.

In what follows, the ice is considered as non-linear viscous incompressible material.

Due to ice incompressibility, the mass conservation equation reduces to:

\[
\frac{\partial v_x}{\partial x} + \frac{\partial v_y}{\partial y} + \frac{\partial v_z}{\partial z} = 0. \tag{3}
\]

Here, \(v_x, v_y, v_z\) denote the components along the three directions of space of the velocity vector \(\mathbf{v}\). The stress tensor \(\boldsymbol{\tau}\) splits into a deviatoric part \(\tau'\) and the isotropic pressure \(p\),

\[
\tau_{ij} = \tau'_{ij} + p\delta_{ij}, \tag{4}
\]

where \(\delta_{ij}\) is the Kronecker symbol. The constitutive relation, that links deviatoric stresses to strain rates \(\dot{\epsilon}\), is a power-law, referred to as the Glen’s flow law in glaciology:

\[
\dot{\epsilon}_{ij} = A(T)\tau_*^2 \tau'_{ij}, \tag{5}
\]

where \(A(T)\) is the deformation rate factor (see Table 1 for adopted values), \(T\) the ice temperature (in this work variations of the pressure melting point are neglected and it is assumed to be at 0°C) and \(\tau_*\) is the second invariant of the deviatoric stress tensor which is defined as:

\[
\tau_*^2 = \frac{1}{2} \tau'_{ij} \tau'_{ij}. \tag{6}
\]

The strain-rate components \(\dot{\epsilon}_{ij}\) are defined from the velocity components as:

\[
\dot{\epsilon}_{ij} = \frac{1}{2} \left( \frac{\partial v_i}{\partial x_j} + \frac{\partial v_j}{\partial x_i} \right), \quad i,j = x,y,z. \tag{7}
\]
The force balance (quasi static equilibrium) in the three directions of space leads to the Stokes equations:

\[
\begin{align*}
\frac{\partial \sigma_x}{\partial x} + \frac{\partial \tau_{xy}}{\partial y} + \frac{\partial \tau_{xz}}{\partial z} &= 0, \\
\frac{\partial \tau_{xy}}{\partial x} + \frac{\partial \sigma_y}{\partial y} + \frac{\partial \tau_{yz}}{\partial z} &= 0, \\
\frac{\partial \tau_{xz}}{\partial x} + \frac{\partial \tau_{yz}}{\partial y} + \frac{\partial \sigma_z}{\partial z} &= \rho g,
\end{align*}
\]

where \(\sigma_i\) stands for \(\tau_{ii}\), \(g\) is the norm of the gravitational acceleration vector and \(\rho\) is the glacier ice density (see Table 1).

### 3.1.1 Boundary conditions

At the upper boundary a stress free boundary is assumed, the corresponding Neumann condition reads:

\[
\tau \cdot n = 0, \text{ at } S.
\]

At the lower boundary we assume zero basal melting \((v \cdot n = 0)\) as well as a linear friction law (Wertman type sliding law, Robin type boundary condition (Greve and Blatter, 2009)). A Robin boundary condition is a linear combination of a Dirichlet (in our case velocities) and Neumann (in our case stress) boundary condition (Arthern and Gudmundsson, 2010).

\[
t \cdot (\tau \cdot n) + \beta v \cdot t = 0, \text{ at } B, \text{ leading to } v = \frac{1}{\beta} \tau,
\]

where \(n\) and \(t\) are normal and tangential unit vectors, \(\beta\) is the basal friction parameter and \(v\) and \(\tau\) are the basal velocity and stress components. This basal friction parameter will be inferred in this study by an inverse method from measured surface
velocities. We emphasize that $\beta$ is the basal friction or drag parameter and not the commonly used sliding coefficient, i.e. higher values of $\beta$ correspond to increased friction and less sliding.

### 3.1.2 Numerics

In the current study this system of four partial differential equations with four independent unknowns (three components of the velocity vector and one for isotropic pressure) are solved numerical using the Finite Element Method based on the code Elmer (Gagliardini and Zwinger, 2008; Zwinger and Moore, 2009).

The mass conservation Eq. (3) and the Stokes Eqs. (8) are written in their variational form

$$\int_V \nabla v \Phi dV = 0, \quad (11)$$

$$-\int_V (1p - \tau') \cdot \nabla \Phi dV = -\oint_{\partial V} (1p - \tau') \cdot n \cdot \Phi dA + \rho \int_V g \Phi dV. \quad (12)$$

A vector-like test function $\Phi$ is introduced, the left-hand side term in the momentum equation has been integrated by parts and reformulated by applying Green's theorem.

The deviatoric stress tensor $\tau'$ is expressed in terms of the strain-rate tensor $\dot{\varepsilon}$ by inversion of the power law (Eq. 5). The non-Newtonian stress-strain relation introduces non-linearities into the system implying the application of an iteration scheme. The system is solved applying the standard Galerkin method. Stabilization is obtained applying either the Stabilized Finite Elements (Franca and Frey, 1992; Brezzi et al., 2004) or the Residual Free Bubble method (Baiocchi et al., 1993). Further discussion on the methods mentioned above is to be found in Gagliardini and Zwinger (2008).

### 3.2 Inverse model

Despite the available observations of surface properties, an important source of uncertainty in ice dynamics remains with the determination of basal drag coefficients. Here
we use a method to determine this basal drag coefficient \( \beta \) from the more readily obtainable data-set on the glacier surface (geometry and velocities) targeting the best fit between observed and modeled surface velocities.

The method is a generalization of the the inverse method proposed by MacAyeal (1992, 1993) with the practical advantage that an already existing forward Stokes model is sufficient. It is a variational method, which relies on the minimization of a cost function measuring the mismatch between the modeled and measured surface velocities. Unlike earlier methods, the method used can be applied to a wider class of models like higher-order models or Full Stokes model similar to Maxwell et al. (2008). It can be applied to any spatially variable Robin coefficient.

3.2.1 The inverse method

The Stokes Eqs. (8) are solved iteratively with two different sets of boundary conditions. A cost-function is calculated which determines when to stop the iteration. The Neumann-type problem defined by Eqs. (3), (8) and the surface boundary conditions (9), as well as the associated Dirichlet-type problem defined by the same equations except that the Neumann upper-surface condition (9) is replaced by a Dirichlet condition (13), are solved iteratively. The Dirichlet condition is defined as

\[
\mathbf{v}(x) = \mathbf{v}_{\text{obs}}(x), \quad \forall x \in S, \quad (13)
\]

where \( \mathbf{v}(x) \) and \( \mathbf{v}_{\text{obs}}(x) \) stand respectively for the modeled and observed horizontal surface velocities. This condition is enforced for each location where surface velocity data is available. The natural Neumann condition is imposed in the vertical direction and where no observations are available.

The cost function, which expresses the mismatch between the two solutions for the velocity field with different boundary conditions on the upper surface \( S \), is given by

\[
J(\beta) = \int_{S} \beta \left( \mathbf{v}^N - \mathbf{v}^D \right) \cdot \left( \boldsymbol{\tau}^N - \boldsymbol{\tau}^D \right) n dS, \quad (14)
\]
where the superscripts $N$ and $D$ refer to the solutions of the Neumann and Dirichlet problems, respectively. As shown by Arthern and Gudmundsson (2010), the inverse problem can be solved by minimizing this cost function. Furthermore, the gradient of this function with respect to $\beta$ can be expressed as a Gâteaux derivative

$$d_\beta J = \int_B \beta' \left( |v^N|^2 - |v^D|^2 \right) ds.$$ 

(15)

The expression for the gradient in Eq. (15) is strictly valid only for linear rheologies, while the gradient for the non-linear rheology of ice is unknown. However, Arthern and Gudmundsson (2010) showed Eq. (15) also minimizes the cost function in the non-linear case of Glen’s flow law, so we can use it here. Between iteration $t$ and $t + 1$, $\beta$ should be updated as

$$\beta^{t+1} = \beta^t - \alpha_\beta \cdot \left( |v^N|^2 - |v^D|^2 \right),$$

with a step-size parameter $\alpha_\beta$. However, in this study an empirical log-type law was used instead of this mathematical correct law since it proved to work well and improved the convergence significantly:

$$\beta^{t+1} = \beta^t \cdot \left( \frac{|v^N|}{v^D} \right)^\nu$$

with a step-size parameter $\nu$.

The minimization of the cost-function $J$ is obtained by the following iterative descent algorithm:

1. initial estimate of $\beta$,
2. solution of the Neumann problem,
3. solution of the Dirichlet problem,
4. update of $\beta$ using an approximation for gradient of the cost function $d_\beta J$,
5. return to step 2 unless convergence is achieved.

### 3.2.2 Technical aspects

The 1995 data-set is sufficiently complete, so that small data gaps could be filled by interpolation and the Dirichlet condition could be enforced everywhere.
Iterations between the Neumann and the Dirichlet problem from an initial value $\beta = 0.001 \text{ MPa yr m}^{-1}$ over the whole domain are conducted. A $\beta_{\text{max}}$ threshold is introduced to prevent the model from producing unrealistically large values at underdetermined locations. That means that very small values and those close to $\beta_{\text{max}}$ do not have a physical meaning, but indicate that the model cannot adapt to fit the data. $\beta_{\text{max}}$ was set to 1 MPa yr m$^{-1}$ after testing that the obtained solution did not change compared to runs with 10 MPa yr m$^{-1}$ or even higher values. The lower boundary is naturally zero.

The relative step power coefficient $\nu$ in the update of $\beta$ was set to 3. In the simulations presented in this work, these iterations could be stopped after 15 iterations since the convergence of the cost-function stagnated very quickly.

### 3.3 Finite Element Mesh

A critical factor in FEM is to discretize the 3-D glacier into a suitable mesh. Too large an element size results in missing crucial effects, while too small sizes result in unrealistically large computing resources (memory as well as computing time). We investigate in Sect. 4.3 effects of varying the horizontal and vertical resolutions. An ideal mesh should reflect the local physical variability scale. Our finite element mesh has been established using a fully automatic adaptive isotropic surface remeshing procedure (Yam's Frey, 2001). The mesh-size is adjusted accordingly to the 2nd derivations of the measured horizontal surface velocities. The obtained mesh (see Fig. 5) has a resolution as fine as 250 m in the fast-flow area and as coarse as 2.5 km in the slow-flow areas. In the vertical direction we use 10 equidistant layers.

### 4 Application and sensitivity tests of the inversion of the basal drag force

To illustrate the importance of the inverse method solution, we firstly present a simpler approach. We then show the results for the basal drag coefficient using our best model
setup and observational data followed by discussion of the sensitivity and robustness of the method as a function of various parameters: the mesh resolution, the surface geometry, the velocity data and temperature within the ice-body.

### 4.1 Forward model using estimated sliding parameters

Studying the observed measured surface velocities, it is clear that sliding occurs on the VSF ice-cap and features the fast flowing outlet glaciers. Attempts to model the ice-cap with one and the same sliding law and parameter everywhere failed, and the model exhibited numerical instabilities. We do not show this elementary case here. Instead we attempted to differentiate “by hand” areas with different coefficients in the sliding law. These two areas were characterized by a slow-flow region (observed velocities below 25 m yr\(^{-1}\)) and a fast-flow region (velocities above this threshold). This leads to a stable model which reflects the existence of the outlet glaciers, but yet fails to reproduce the correct velocity pattern of those glaciers. In Fig. 8 (right) an example of this approach is shown. The poor velocity match was produced despite using sliding coefficients tuned to representative values in each outlet obtained from the inverse method (Sect. 4.2). The values are given in Table 2. There is ambiguity in arbitrarily deciding on the boundary between the two sliding areas and choosing the sliding coefficients. In contrast the application of the inverse method requires no need for defining slow and fast-flow areas by hand or tuning of coefficients. Furthermore, this will lead to a correct flow-pattern even inside the outlet glaciers.

### 4.2 Inverse model using 1995 data

The inverse method as described in Sect. 3 was applied to the VSF ice-cap using the 1995 surface velocity data. The convergence was surprisingly fast and after only 15 iterations the algorithm was stopped. The trend of the cost function (normalized to the value of the first iteration) throughout the minimization algorithm is presented in Fig. 6 on a logarithmic scale. The cost-function does not decrease to zero since we
are dealing with real data and the absolute value has – if any – meaning only in terms of how well it is possible to fit the data with the model.

The obtained basal drag coefficient $\beta$ is presented in Fig. 7 (left). The blueish colors correspond to areas with little sliding, the glacier is nearly frozen to the bed while the reddish areas correspond to areas with very fast sliding. The overall pattern corresponds very well to that which may be expected from the surface topography and measured surface velocities. $\beta$ is limited (naturally) by 0 and (artificially) by 1 MPa yr m$^{-1}$. We examined (not shown) changes in $\beta$ very close to zero (fast flowing grid-points). Introducing a lower threshold of $\beta = 10^{-6}$ MPa yr m$^{-1}$ did not alter the resulting velocity field, showing that very low values of $\beta$ do not affect the velocity field.

The velocities obtained with the model (see Fig. 8, left) are very close to the observed surface velocities. The histogram Fig. 10 quantifies the error between measured and modeled velocities and shows that for the most grid points the error is below 20 m per year, we recall that the error in the observation is about 7 m yr$^{-1}$. Figure 9 provides, in addition to the cost-function and the histogram, the spatial distribution of the relative error between the velocities obtained during the solution of the Neumann and respectively the Dirichlet models. It is important to consider this spatial distribution because of the two distinct flow regimes. Over the most of the glacier the difference is less than 30%. There is no noticeable difference in the quality of the results between fast and slow flow regions. Most of the grid points having higher errors are located in the fast-flowing outlets or are isolated points close to the margins plausibly due to observational errors. The correlation plot between the measured and observed surface velocities is shown in Fig. 11 and illustrates once more the agreement over the whole range of velocities.

We observe some spurious values in some isolated points in the center of the icecap where we expect little or no sliding. There we obtain very low basal drag coefficients. These points are situated in the slow-flow areas right on the top of the ridges (Fig. 8, inset), where we expect the data to be erroneous since small errors in position may cause large errors in velocity vectors. Nevertheless, since the velocities are small
here, the big error in $\beta$ has little impact on the resulting velocity distribution and hence can be ignored.

4.3 Mesh resolution

A test-run with 15 vertical layers led to identical results (not shown), as the one produced by 10 layers meaning that in this case 10 are sufficient. A similar mesh as the previously used was established, but with coarser resolution in the fast-flow areas (minimum mesh-size of 500 m compared with 250 m). The results are presented in Fig. 12. Looking at the example of Frazerbreen it is clear that some features remain unresolved with a too coarse resolution. Also the cost-function is increased by 20% with the coarser resolution showing that finer resolution allows for a better reproduction of the measured velocities. This illustrates the importance of the choice of an adaptive mesh and emphasizes the necessity of a fine resolution in the regions of interest. The changes in the results when switching from a homogeneous mesh to an adaptive mesh are even more striking (Fig. 12, right). The reason for this is probably not only the increase in resolution, but the fact that the method is sensitive to local grid resolution.

4.4 Sensitivity of the method to the DEMs

To test the sensitivity of the method to errors in the surface DEM and also to verify if the difference in dates between surface DEM and velocity data epochs matter we made test-runs with slightly modified surface DEMs. The method seems to be rather stable with respect to such changes. Test-runs for a surface DEM which was modified by 15 times the current mean annual surface mass balance (Möller et al., 2011) were performed and no significant changes in the basal friction pattern could be observed. This modification of the DEM is much greater than what was observed on VSF between the 1950s (Schytt, 1964) and today. This modification was chosen since the change during this period cannot be quantified exactly over the whole ice-cap. We chose it as very rough approximation of what could happen to the ice-cap during a 15 year period.
of extreme non-steady-state surface mass-balance. Modifications on the bedrock topology were also made, showing that obtained basal pattern is dominated by the surface velocities rather than changes in the bedrock DEM since the obtained basal distribution still reflects strongly the surface velocity pattern.

### 4.5 Influence of errors in the velocity data

The error on the measured velocities is estimated to about 7 m per year by Pohjola et al. (2011a). A simulation with a random error between −10 m and +10 m per year on both components of the horizontal velocities was performed. The obtained distribution of the basal drag coefficient as well as the modeled surface velocities was not affected by this perturbation (not shown).

### 4.6 Influence of the ice temperature profile

The inverse method simulation was conducted for an isothermal temperate ice-body. This represents an upper physical bound to ice deformation, though it is certainly unrealistic over much of the ice cap. Results presented in Fig. 7 (right) are rather close to those we obtained with our preferred temperature profile. The correlation between modeled surface velocities (Fig. 13) with the cold and the warm temperature scenario confirms the good agreement of the results obtained with the different assumptions for the temperature. In addition in the same figure the correlation is also shown for the basal velocities. In fact, it is not surprising that the surface velocities are quite similar in both scenarios since the inverse method minimized the difference between observed and modeled surface velocities. But one could imagine that different assumptions for the temperature distribution lead to differences in the viscosity and consequently in differences in the distributions of basal velocities. But in our case the total velocity is dominated by the sliding, the deformation plays hardly any role and we observe nearly the same correlation between velocity fields with different temperature profiles on the surface and at the basis. We can conclude that our earlier presented assumptions are good enough for the study of the basal conditions.
5 Discussion and conclusions

We have presented a 3-D application of the inverse method proposed by Arthern and Gudmundsson (2010). The distribution of the basal friction parameter underneath the VSF ice-cap was inferred from the 1995 winter surface velocity distribution. We obtain a good representation of the observed velocities and can reproduce the fast-flowing outlet glaciers as well as the acceleration pattern within these glaciers. The average value of the error between measured and observed surface velocities is 18 m yr\(^{-1}\) compared to an error of 7 m yr\(^{-1}\) in the observed velocities. The improvement compared to the simpler models with constant sliding coefficients as shown in Sect. 4.1 is clearly visible and shows the necessity of a good knowledge of the spatial distribution of the basal friction parameter \(\beta\).

The robustness of the method with respect to spatial resolution and errors in the input data has been studied and other technical aspects have been discussed. We have also shown that the obtained \(\beta\) distribution does not depend strongly on the temperature, justifying our approximation on the temperature profile in the ice-body. The correlation between modeled surface velocities obtained with different assumption for the ice temperature is satisfactory. Too small values of \(\beta\) have little physical meaning, leading to the conclusion that even though the spatial distribution of \(\beta\) is important, very small local variations in \(\beta\) are not visible in the velocity distribution. The knowledge of the basal conditions will be important in prognostic simulations, but also to improve our understanding of the mechanism of surging.

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Table 1. Glen’s law parameter, Arrhenius factor in SI units (Pa$^{-3}$ s$^{-1}$), Paterson (1994) and other parameters used.

<table>
<thead>
<tr>
<th>$A$</th>
<th>$T$</th>
</tr>
</thead>
<tbody>
<tr>
<td>$3.985 \times 10^{-13}$</td>
<td>$T &lt; -10$ °C</td>
</tr>
<tr>
<td>$1.916 \times 10^{-13}$</td>
<td>$-10$ °C $\leq T \leq 0$ °C</td>
</tr>
<tr>
<td>$908$ kg m$^3$</td>
<td>$9.1376 \times 10^{-19}$ MPa a$^2$ m$^2$ density</td>
</tr>
<tr>
<td>$9.81$ m s$^2$</td>
<td>$9.77 \times 10^{15}$ m a$^2$ gravity</td>
</tr>
</tbody>
</table>
Table 2. Values used for $\beta$ in the simpler approach.

<table>
<thead>
<tr>
<th>Outlet glacier</th>
<th>$\beta$ (MPa yr m$^{-1}$)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Franklinbreen, Donckerlobe</td>
<td>0.001002</td>
</tr>
<tr>
<td>Sabinenbreen</td>
<td>0.004706</td>
</tr>
<tr>
<td>Maudbreen</td>
<td>0.000113</td>
</tr>
<tr>
<td>Rijpbreen</td>
<td>0.000429</td>
</tr>
<tr>
<td>Bodleybreen</td>
<td>0.004946</td>
</tr>
<tr>
<td>Aldousbreen</td>
<td>0.000980</td>
</tr>
<tr>
<td>Frazerbreen</td>
<td>0.001243</td>
</tr>
<tr>
<td>Idunbreen</td>
<td>0.000367</td>
</tr>
<tr>
<td>Gimlebreen</td>
<td>0.000401</td>
</tr>
</tbody>
</table>
Fig. 1. Presentation of the research area: VSF ice-cap in Svalbard and digital elevation model of the surface elevation. It is based on the topographic map from the Norwegian Polar Institute (NPI) (1:100 000, 1990) and completed with the International Bathymetric Chart of the Arctic Ocean (IBCAO) (Jakobsson et al., 2008) in the sea.

The Ahlmann summit is located at the position A and the ice-core mentioned in the temperature discussion was drilled at the location C. The names of the different outlet glaciers are detailed in Fig. 2. The glacier outline is also based on the NPI maps. All data is used in the coordinate system WGS 1994, UTM zone 33 N.
Fig. 2. Horizontal surface velocity in 1995 from ERS-1/2 1-day InSAR scenes.
Fig. 3. Digital elevation model for the bedrock elevation, overlaid are the ground-based (black) and airborne radar profiles (blue, gray: lack of bed return). In the land areas this map is completed with the NPI map and in sea-areas with the IBCAO data, however these data was not used to establish the bedrock map from the individual profiles.
Fig. 4. Typical temperature profiles in a polythermal ice-body at a location with ice-thickness 340 m. The red, pink and blue curves are typical steady state curves for accumulation area, equilibrium line and ablation area respectively without influence of refreezing of percolating seasonal meltwater. The gray dots represent measurements of an ice-core drilled at the east-end of the summit ridge which are different from the steady state curve due latent heat created through superimposed ice.
Fig. 5. Unstructured mesh established with Yams (Frey, 2001) accordingly to measured surface velocities for the 1995 velocity data set.
Fig. 6. Cost-function of the inverse method, normalized to the value of the first iteration. The iteration was stopped after 15 iteration.
Fig. 7. Obtained distribution for the basal drag coefficient $\beta$ in a non-homogeneous logarithmic scale, all values below 1 Pa yr$^{-1}$ are grouped in one color. On the left side the distribution for our cold temperature scenario is shown, while on the right side the distribution for a temperate ice-body is opposed. In the small inlet the aspect ratio of the surface DEM is presented. For better visibility the figures are here not overlaid. However, superposing the figures shows that the strikingly small values in the center of the ice-cap occur on topographic highs.
Fig. 8. Velocities obtained in a forward simulation by the model, on the left when using the basal distribution obtained with the inverse method, on the right by using a simpler approach: In each outlet glacier a different, but constant sliding coefficient is used. The values here are in each outlet the mean-values of the values obtained with the inverse method (Table 2).
Fig. 9. Spatial distribution of the relative missfit between Neumann and Dirichlet velocities over the whole glacier surface. Comparison with Fig. 8 indicates that relative errors are similar in the fast-flow and slow-flow regime areas. The points with relative errors higher than 145% are highlighted in red to demonstrate that these are single points, mostly close to the margins.
Fig. 10. Histogram of the difference between modeled and observed surface velocities. Note, that the error on the measured surface velocities is $7 \text{ m yr}^{-1}$ and that the last bin contains all entries with values higher than $280 \text{ m yr}^{-1}$.
Fig. 11. Correlation between the modeled and observed surface velocities.
Fig. 12. Comparison of the distribution of the basal drag coefficient with the usual (left), a coarser mesh (middle) and a regular mesh (right). The figure shows a zoom on two of the southern outlet glaciers.
Fig. 13. Correlation between the modeled velocities for the two different temperature scenarios – for the surface velocities and respectively the basal velocities. Since the movement is dominated by the basal motion, there is hardly any difference between both velocity fields.