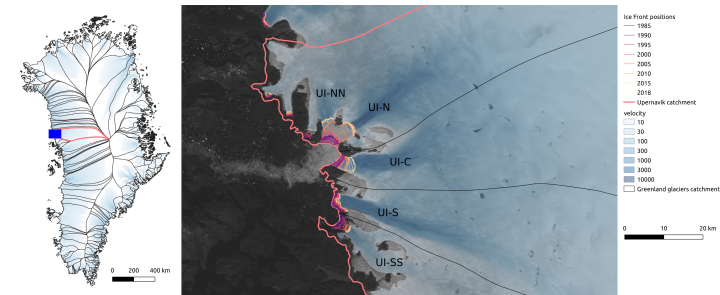


Ensemble simulation for ice sheet model initialisation

Method and application to Upernavik Isstrøm



Eliot Jager (PhD student / 3rd year)
IGE, CNRS, Université Grenoble-Alpes

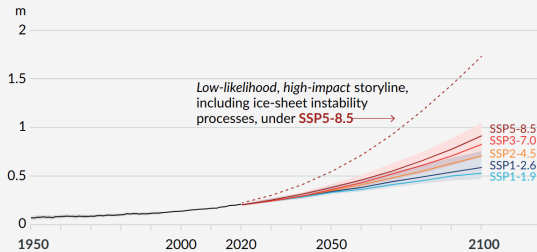
Supervisors : Gillet-Chaulet Fabien, Champollion Nicolas, Mougintot Jérémie

Tuesday April 4th 2023

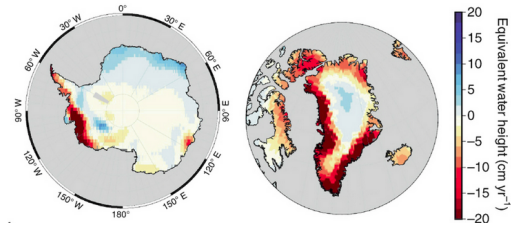


Sea level rise

(d) Global mean sea level change relative to 1900



source : IPCC AR6 Summary for policymakers

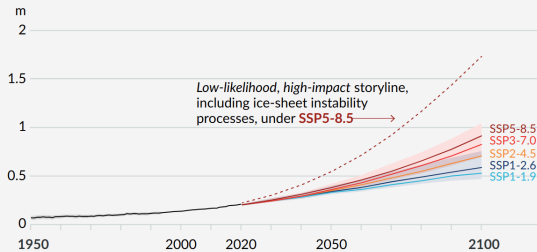


source : mission GRACE

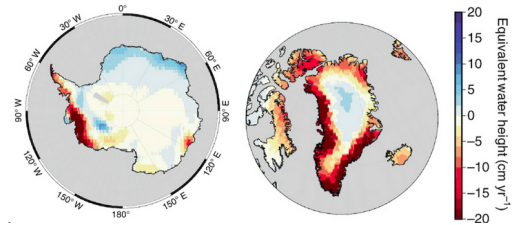
- Contributors: ocean expansion, glaciers, **GrIS** and **AIS**
- High uncertainties of the future of **AIS** and **GrIS**

Sea level rise

(d) Global mean sea level change relative to 1900



source : IPCC AR6 Summary for policymakers

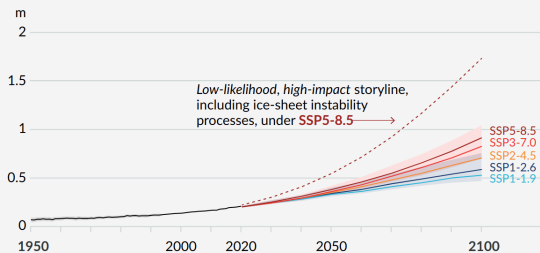


source : mission GRACE

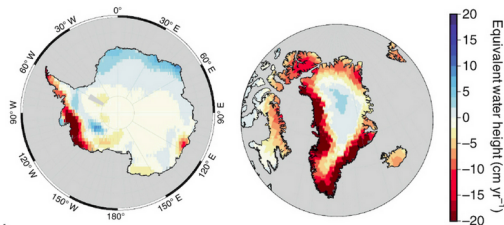
- Contributors: ocean expansion, glaciers, **GrIS** and AIS
- High uncertainties of the future of AIS and **GrIS**

Sea level rise

(d) Global mean sea level change relative to 1900



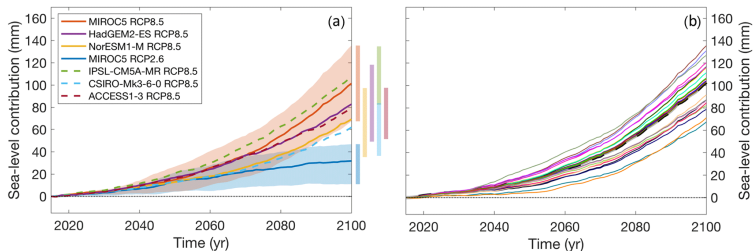
source : IPCC AR6 Summary for policymakers



source : mission GRACE

- Contributors: ocean expansion, glaciers, **GrIS** and AIS
- High uncertainties of the future of AIS and **GrIS**

ISMIP6 Greenland

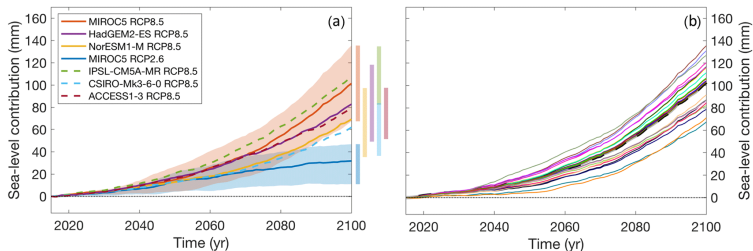


Ensemble sea-level projections. a) Different GCM for all ISM. b) Different ISM results for one forcing (Heiko Goelzer et al.)

State of the art for sea-level projection :

- Great representation of uncertainty related to GCM and different ISM
- Limited number of SSP and RCM, no representation of the intern uncertainty of ISM
- Ability of models to represent past changes?

ISMIP6 Greenland

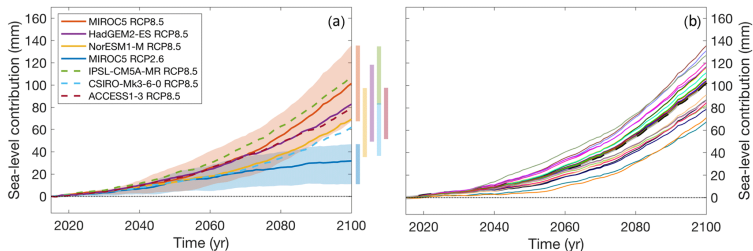


Ensemble sea-level projections. a) Different GCM for all ISM. b) Different ISM results for one forcing (Heiko Goelzer et al.)

State of the art for sea-level projection :

- Great representation of uncertainty related to GCM and different ISM
- Limited number of SSP and RCM, no representation of the intern uncertainty of ISM
- Ability of models to represent past changes?

ISMIP6 Greenland

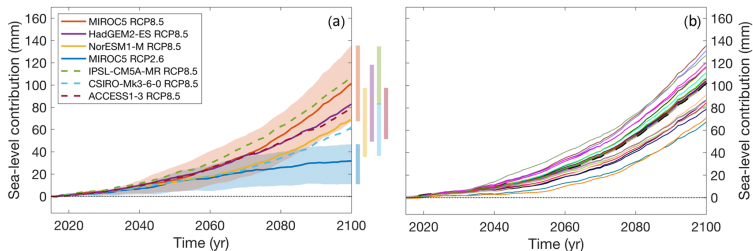


Ensemble sea-level projections. a) Different GCM for all ISM. b) Different ISM results for one forcing (Heiko Goelzer et al.)

State of the art for sea-level projection :

- Great representation of uncertainty related to GCM and different ISM
- Limited number of SSP and RCM, no representation of the intern uncertainty of ISM
- Ability of models to represent past changes?

ISMIP6 Greenland



Ensemble sea-level projections. a) Different GCM for all ISM. b) Different ISM results for one forcing (Heiko Goelzer et al.)

State of the art for sea-level projection :

- Great representation of uncertainty related to GCM and different ISM
- Limited number of SSP and RCM, no representation of the intern uncertainty of ISM
- Ability of models to represent past changes?

Research questions



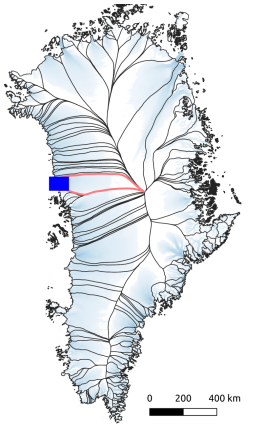
1. Uncertainty quantification for our local ISM : Elmer/Ice
2. Ability of Elmer/Ice to represent past changes : velocity, elevation, ice discharge and ice mass loss

Research questions

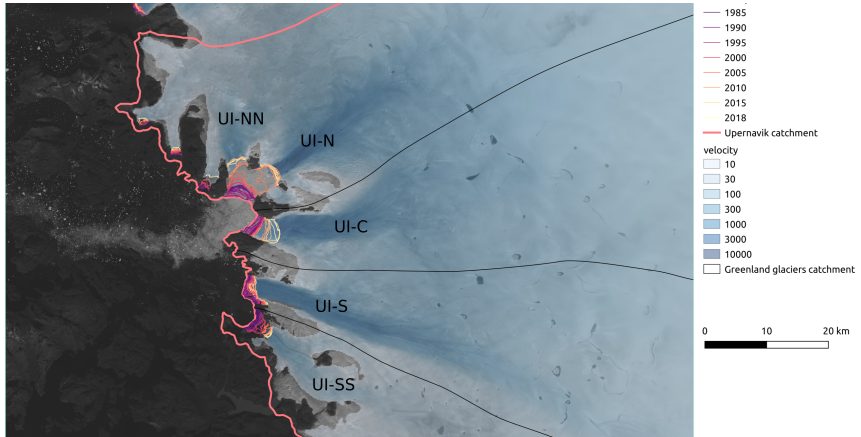
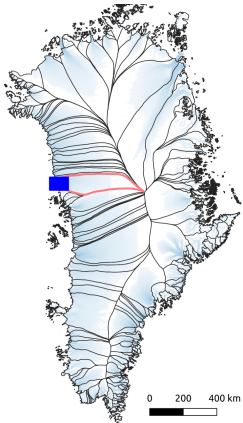


1. Uncertainty quantification for our local ISM : Elmer/Ice
2. Ability of Elmer/Ice to represent past changes : velocity, elevation, ice discharge and ice mass loss

Why Upernavik Isstrøm ?

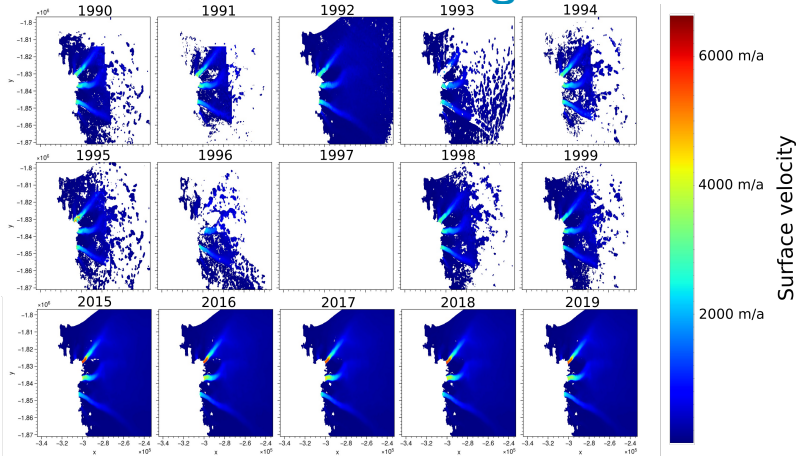


Why Upernavik Isstrøm ?



Upernavik Isstrøm catchment situation and front evolution since 1985

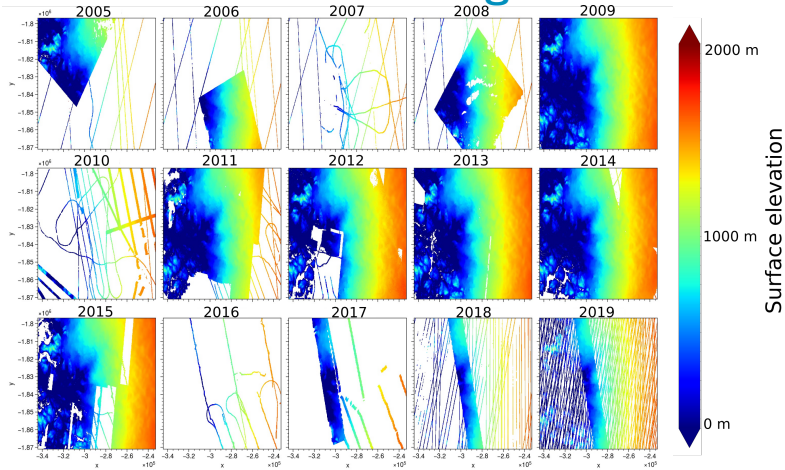
Big data era



- Increase of velocity and elevation observations \Rightarrow How can we use this increasing amount of data to better constrain the uncertain initial parameters?

Velocity maps of Upernavik Isstrøm (90's and 2010's)

Big data era



- Increase of velocity and elevation observations \Rightarrow How can we use this increasing amount of data to better constrain the uncertain initial parameters?

Elevation maps of Upernavik Isstrøm (2005-2019)

Uncertainty quantification

1. Which variables are uncertain?
2. Putting it in the form of a probability law
3. Drawing parameter sets by Latin Hypercube sampling

Uncertainty quantification

1. Which variables are uncertain?
2. Putting it in the form of a probability law
3. Drawing parameter sets by Latin Hypercube sampling

Uncertainty quantification

1. Which variables are uncertain?
2. Putting it in the form of a probability law
3. Drawing parameter sets by Latin Hypercube sampling

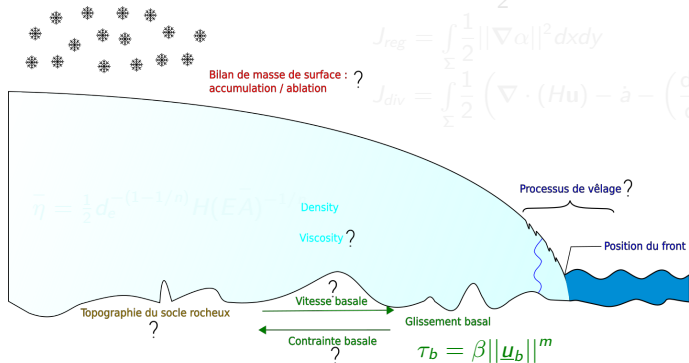
Inverse method uncertainties

$$J = J_0 + \lambda_{reg} J_{reg} + \lambda_{div} J_{div}$$

$$J_0 = \sum_{i=1}^{N_{obs}} \frac{1}{2} (\|\mathbf{u}_i^{mod} - \mathbf{u}_i^{obs}\|)^2$$

$$J_{reg} = \int_{\Sigma} \frac{1}{2} \|\nabla \alpha\|^2 dx dy$$

$$J_{div} = \int_{\Sigma} \frac{1}{2} \left(\nabla \cdot (H\mathbf{u}) - \dot{a} - \left(\frac{dH}{dt} \right)_{obs} \right)^2 dx dy$$



Uncertain parameters :

- friction law (m)
- viscosity (E)
- year of observation velocity (u_{obs})
- weight of cost functions (λ_{reg} et λ_{div})

Processes influencing glacial dynamics and their uncertainties

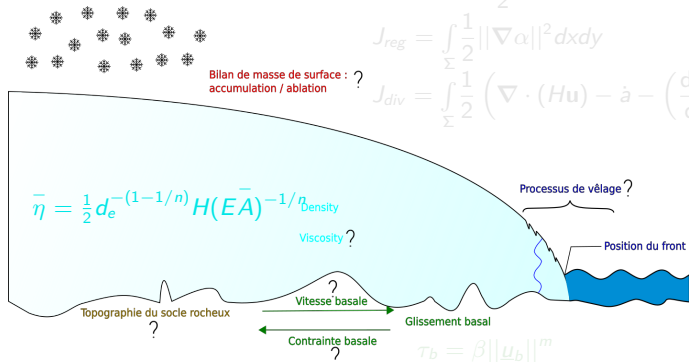
Inverse method uncertainties

$$J = J_0 + \lambda_{reg} J_{reg} + \lambda_{div} J_{div}$$

$$J_0 = \sum_{i=1}^{N_{obs}} \frac{1}{2} (\|\mathbf{u}_i^{mod} - \mathbf{u}_i^{obs}\|)^2$$

$$J_{reg} = \int_{\Sigma} \frac{1}{2} \|\nabla \alpha\|^2 dx dy$$

$$J_{div} = \int_{\Sigma} \frac{1}{2} \left(\nabla \cdot (H\mathbf{u}) - \dot{a} - \left(\frac{dH}{dt} \right)_{obs} \right)^2 dx dy$$



Uncertain parameters :

- friction law (m)
- viscosity (E)
- year of observation velocity (u_{obs})
- weight of cost functions (λ_{reg} et λ_{div})

Processes influencing glacial dynamics and their uncertainties

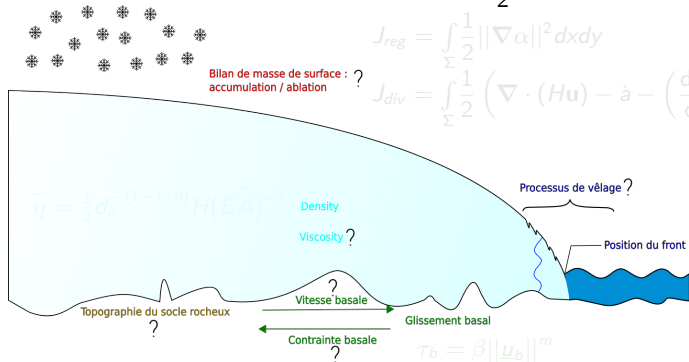
Inverse method uncertainties

$$J = J_0 + \lambda_{reg} J_{reg} + \lambda_{div} J_{div}$$

$$J_0 = \sum_{i=1}^{N_{obs}} \frac{1}{2} (\|\mathbf{u}_i^{mod} - \mathbf{u}_i^{obs}\|)^2$$

$$J_{reg} = \int_{\Sigma} \frac{1}{2} \|\nabla \alpha\|^2 dx dy$$

$$J_{div} = \int_{\Sigma} \frac{1}{2} \left(\nabla \cdot (H\mathbf{u}) - \dot{a} - \left(\frac{dH}{dt} \right)_{obs} \right)^2 dx dy$$



Uncertain parameters :

- friction law (m)
- viscosity (E)
- year of observation velocity (u_{obs})
- weight of cost functions (λ_{reg} et λ_{div})

Processes influencing glacial dynamics and their uncertainties

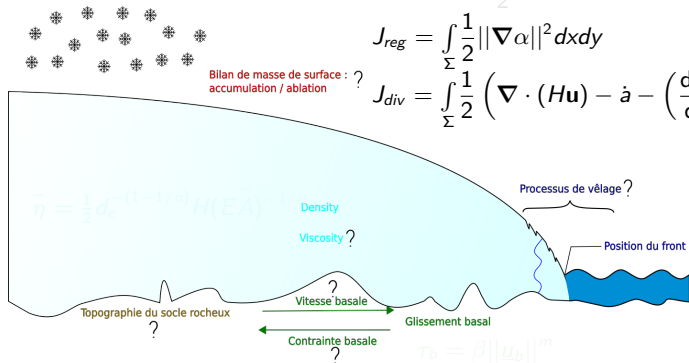
Inverse method uncertainties

$$J = J_0 + \lambda_{reg} J_{reg} + \lambda_{div} J_{div}$$

$$J_0 = \sum_{i=1}^{N_{obs}} \frac{1}{2} (\|u_i^{mod} - u_i^{obs}\|)^2$$

$$J_{reg} = \int_{\Sigma} \frac{1}{2} \|\nabla \alpha\|^2 dx dy$$

$$J_{div} = \int_{\Sigma} \frac{1}{2} \left(\nabla \cdot (Hu) - \dot{a} - \left(\frac{dH}{dt} \right)_{obs} \right)^2 dx dy$$

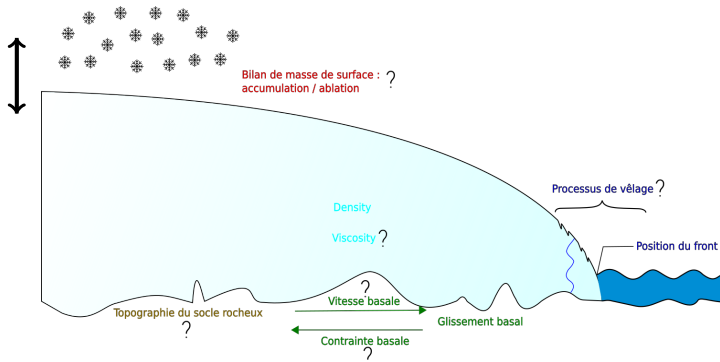


Uncertain parameters :

- friction law (m)
- viscosity (E)
- year of observation velocity (u_{obs})
- weight of cost functions (λ_{reg} et λ_{div})

Processes influencing glacial dynamics and their uncertainties

Initial surface uncertainty

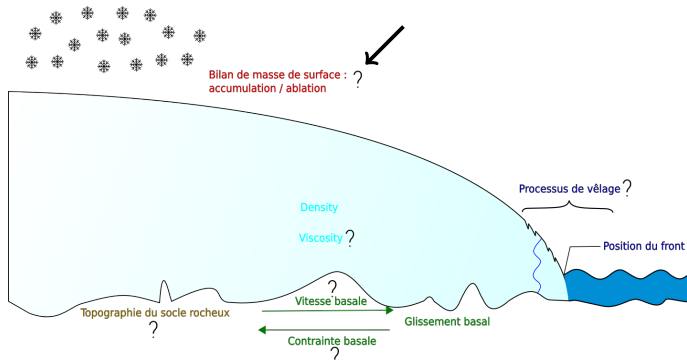


Variable parameters :

- Relaxation time t_{relax}
- Constant surface mass balance SMB_{date}

Processes influencing glacial dynamics and their uncertainties

Forcing uncertainty



Uncertain parameters :

- Two different SMB models (SMB_{model})
- No uncertainty of front position

Processes influencing glacial dynamics and their uncertainties

Experimental design

Results for the ensemble using the regularised Coulomb law:

1. Drawing of uncertain parameter sets
2. Inversions to calibrate β_{RC} with data from the 2010-2020 period
3. Ensemble of surfaces obtained with relaxations (constant forcing)
4. Ensemble simulation between 1985 and 2020
5. Analysis of the sensitivity to different input parameters

Experimental design

Results for the ensemble using the regularised Coulomb law:

1. Drawing of uncertain parameter sets
2. Inversions to calibrate β_{RC} with data from the 2010-2020 period
3. Ensemble of surfaces obtained with relaxations (constant forcing)
4. Ensemble simulation between 1985 and 2020
5. Analysis of the sensitivity to different input parameters

Experimental design

Results for the ensemble using the regularised Coulomb law:

1. Drawing of uncertain parameter sets
2. Inversions to calibrate β_{RC} with data from the 2010-2020 period
3. Ensemble of surfaces obtained with relaxations (constant forcing)
4. Ensemble simulation between 1985 and 2020
5. Analysis of the sensitivity to different input parameters

Experimental design

Results for the ensemble using the regularised Coulomb law:

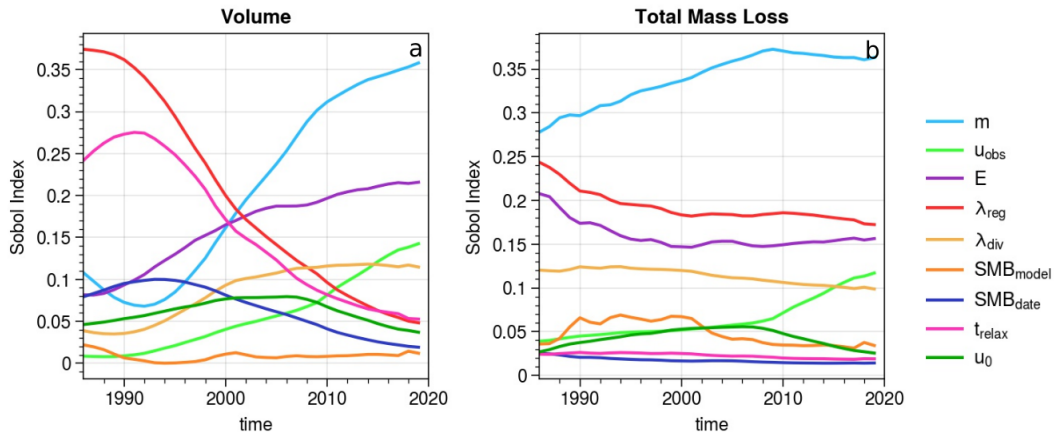
1. Drawing of uncertain parameter sets
2. Inversions to calibrate β_{RC} with data from the 2010-2020 period
3. Ensemble of surfaces obtained with relaxations (constant forcing)
4. Ensemble simulation between 1985 and 2020
5. Analysis of the sensitivity to different input parameters

Experimental design

Results for the ensemble using the regularised Coulomb law:

1. Drawing of uncertain parameter sets
2. Inversions to calibrate β_{RC} with data from the 2010-2020 period
3. Ensemble of surfaces obtained with relaxations (constant forcing)
4. Ensemble simulation between 1985 and 2020
5. Analysis of the sensitivity to different input parameters

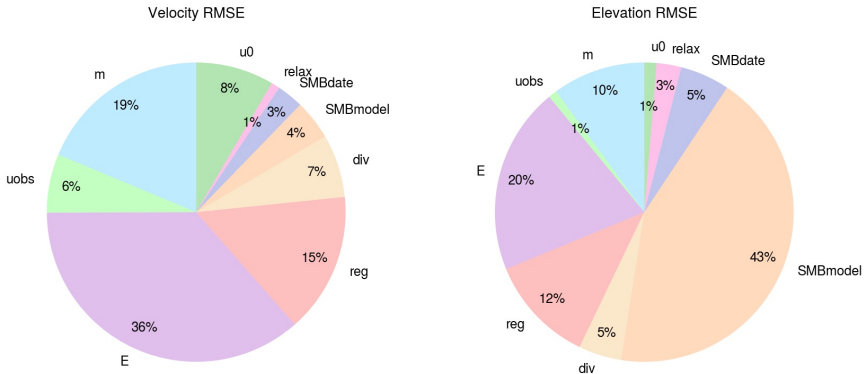
Global variables



Sobol index for volume and mass change for RCE over time

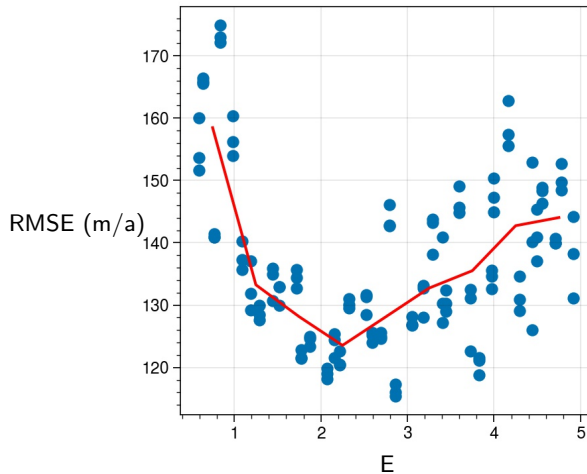
Influence on scores

Sobol Index



Sobol index for RMSE on velocities (left) and elevations (right)

Influence of E

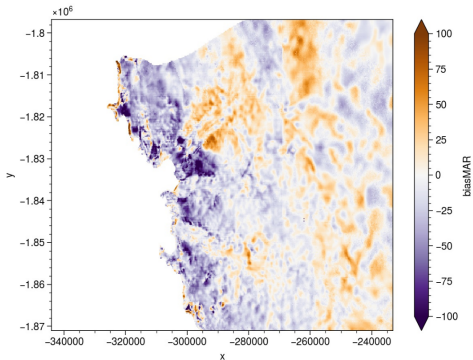


RMSE in function of E

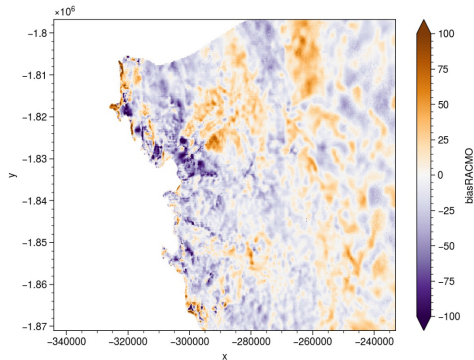
Ensemble simulation for ice sheet model initialisation

Influence of SMB

MAR



RACMO



Bias for the mean member depending on the SMB model used

Take home messages

1. Initial surface has little influence on total mass loss
 2. The calibration parameters have a strong influence on the different quantities
 3. Possibility to refine uncertain parameters (E , SMB_{model})
- Jager et al., 2023, JOG, (submitted)

Take home messages

1. Initial surface has little influence on total mass loss
 2. The calibration parameters have a strong influence on the different quantities
 3. Possibility to refine uncertain parameters (E , SMB_{model})
- Jager et al., 2023, JOG, (submitted)

Take home messages

1. Initial surface has little influence on total mass loss
 2. The calibration parameters have a strong influence on the different quantities
 3. Possibility to refine uncertain parameters (E , SMB_{model})
- Jager et al., 2023, JOG, (submitted)

Conclusion on the method

Our new initialisation method:

1. reproduces the past trend without bias and we arrive at a dynamic state ensemble close to reality \Rightarrow Allows confidence in projections.
2. allows the representation of internal uncertainty of our ISM Elmer/Ice

Jager et al., 2023, JOG, (submitted)

Conclusion on the method

Our new initialisation method:

1. reproduces the past trend without bias and we arrive at a dynamic state ensemble close to reality \Rightarrow Allows confidence in projections.
2. allows the representation of internal uncertainty of our ISM Elmer/Ice

Jager et al., 2023, JOG, (submitted)

Perspectives

1. Upernavik projections : the importance of uncertainties related to dynamics (friction, rheology) compared to those related to forcing (surface mass balance, interaction with the ocean); Jager et al., 2023, (in prep.)
2. Assessment projections : What is the right way to score members to evaluate projections ? Jager et al., 2023, (in prep.)
3. Projections at GrIS scale : Implementation of the friction parameterisation; work in progress

Perspectives

1. Upernavik projections : the importance of uncertainties related to dynamics (friction, rheology) compared to those related to forcing (surface mass balance, interaction with the ocean); Jager et al., 2023, (in prep.)
2. Assessment projections : What is the right way to score members to evaluate projections ? Jager et al., 2023, (in prep.)
3. Projections at GrIS scale : Implementation of the friction parameterisation; work in progress

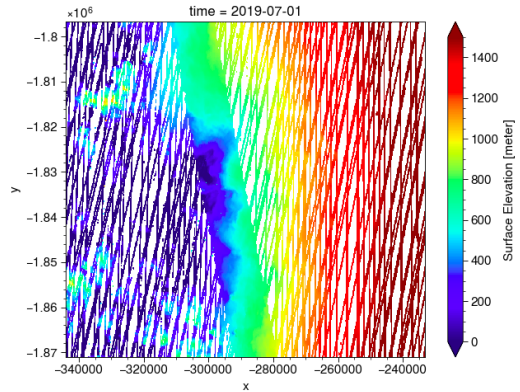
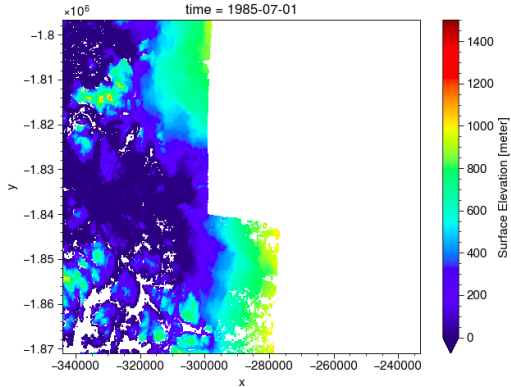
Perspectives

1. Upernavik projections : the importance of uncertainties related to dynamics (friction, rheology) compared to those related to forcing (surface mass balance, interaction with the ocean); Jager et al., 2023, (in prep.)
2. Assessment projections : What is the right way to score members to evaluate projections ? Jager et al., 2023, (in prep.)
3. Projections at GrlS scale : Implementation of the friction parameterisation; work in progress

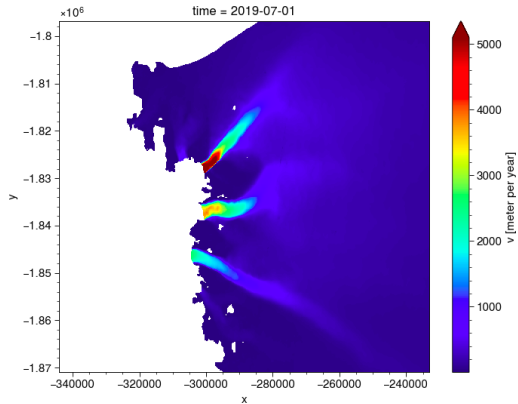
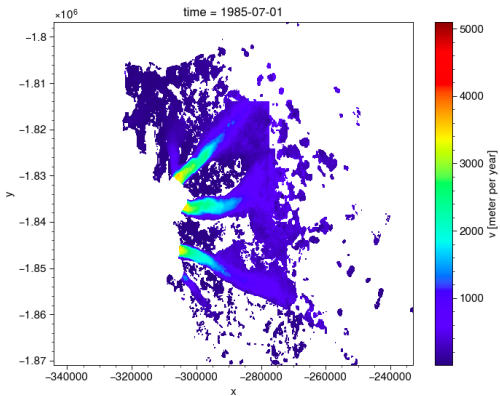
References

- Front positions : Wood et al. 2021
- Velocity : Joughin et al. 2018, Mouginot et al. 2017, Derkacheva et al. 2020
- Elevation : Howat et al. 2017, Howat et al. 2014, Korsgaard 2016, Brunt 2016, Moller 2019
- Bedmachine : Morlighem et al. 2017
- Assimilation : Gillet-Chaulet 2020
- Elmer/Ice : Gagliardini et al. 2013
- Numerical environnement : Most of the computations presented in this study were performed using the GRICAD infrastructure (<https://gricad.univ-grenoble-alpes.fr>), which is supported by Grenoble research communities.

Observations

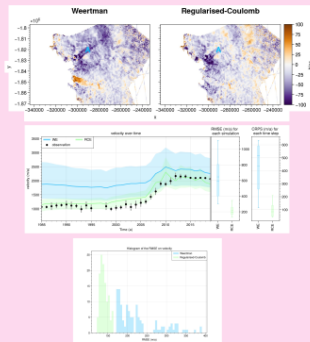


Observations



Analyse

Scores :
bias, RMSE, CRPS



What we do

1. Part of the data used for calibration and initialisation.
2. A large part of it is kept for validation! The scores allow the performance of the model to be quantified.
3. For greater robustness of the results, a set of simulations is carried out: this makes it possible to take into account the uncertainties of the calibration, the initialization, the forcing and the physical assumptions

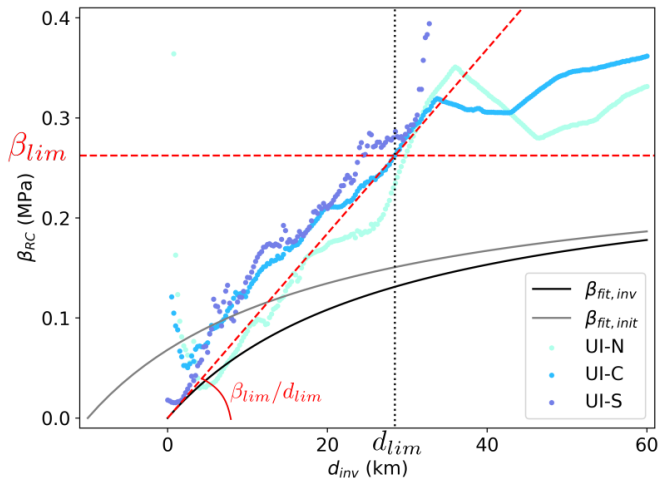
What we do

1. Part of the data used for calibration and initialisation.
2. A large part of it is kept for validation! The scores allow the performance of the model to be quantified.
3. For greater robustness of the results, a set of simulations is carried out: this makes it possible to take into account the uncertainties of the calibration, the initialization, the forcing and the physical assumptions

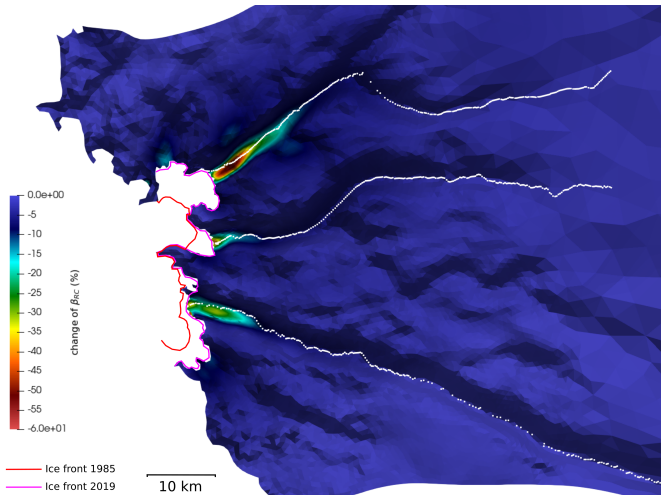
What we do

1. Part of the data used for calibration and initialisation.
2. A large part of it is kept for validation! The scores allow the performance of the model to be quantified.
3. For greater robustness of the results, a set of simulations is carried out: this makes it possible to take into account the uncertainties of the calibration, the initialization, the forcing and the physical assumptions

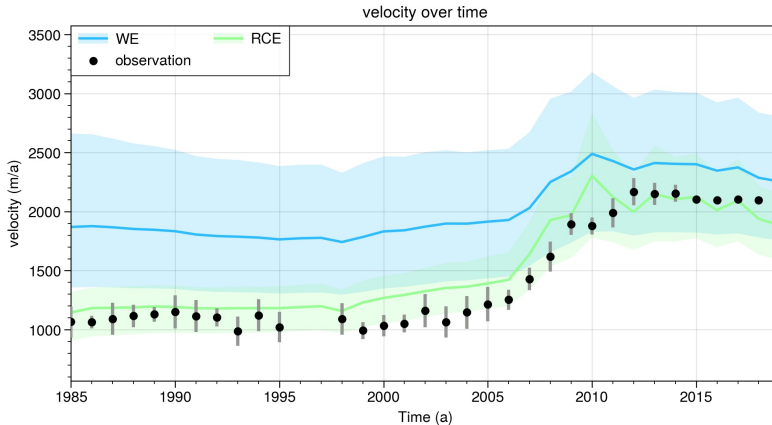
Friction parametrisation



Friction parametrisation



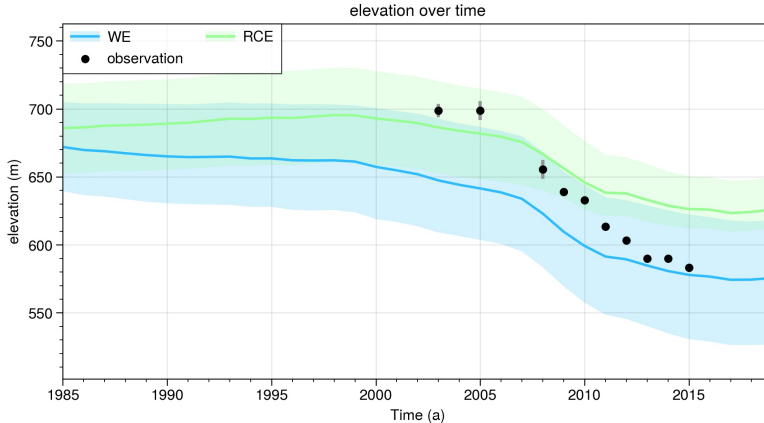
Transient: velocity



Changes in WE and RCE velocities over time at point A (UI-N)

Ensemble simulation for ice sheet model initialisation

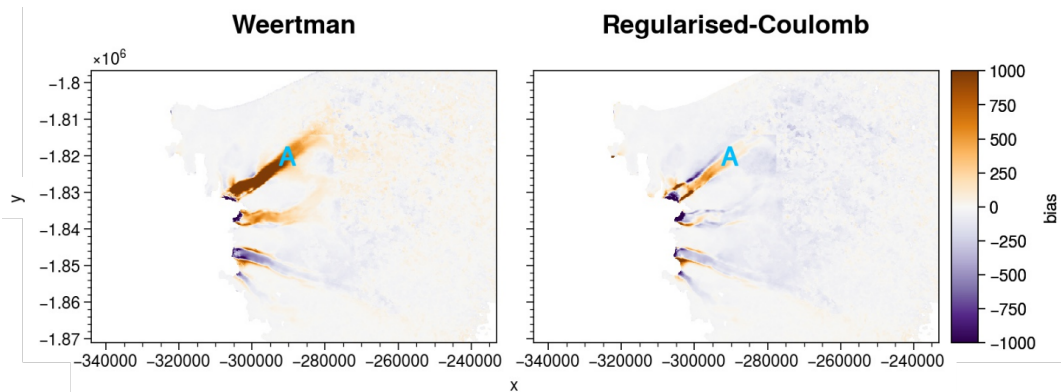
Transient: elevation



Changes in WE and RCE elevations over time at point A (UI-N)

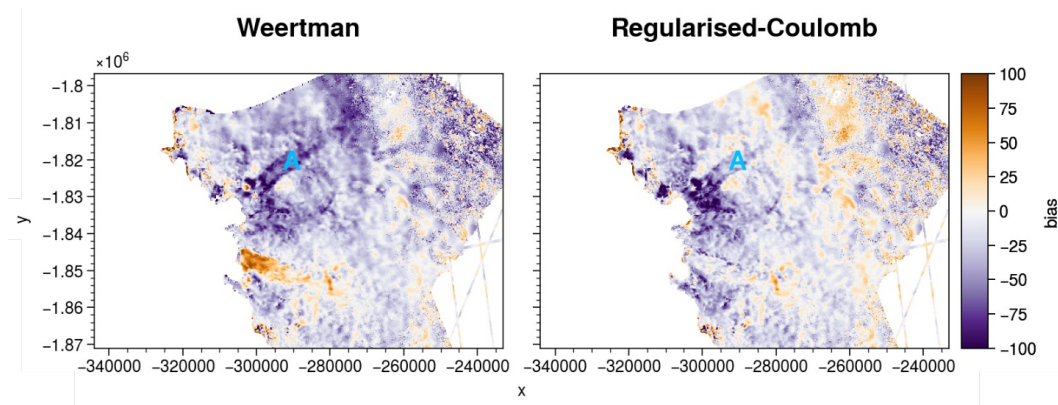
Ensemble simulation for ice sheet model initialisation

Initial period : velocities

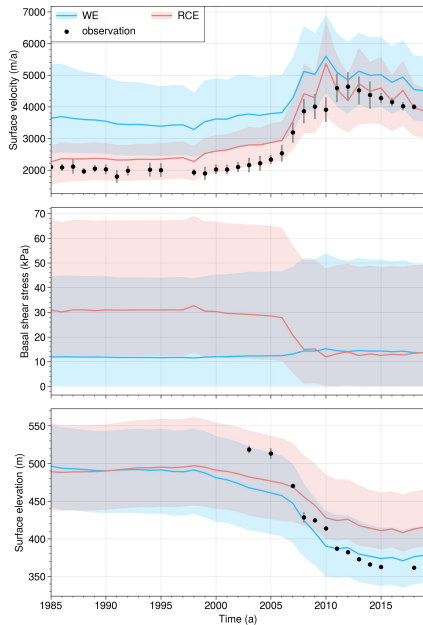


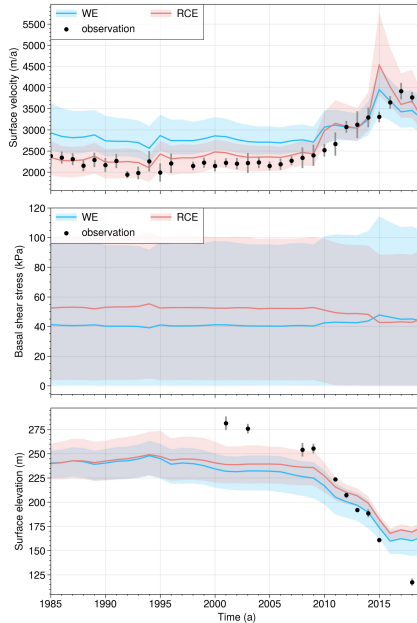
Surface velocity bias of the **ensemble mean** during the period 1985-2005

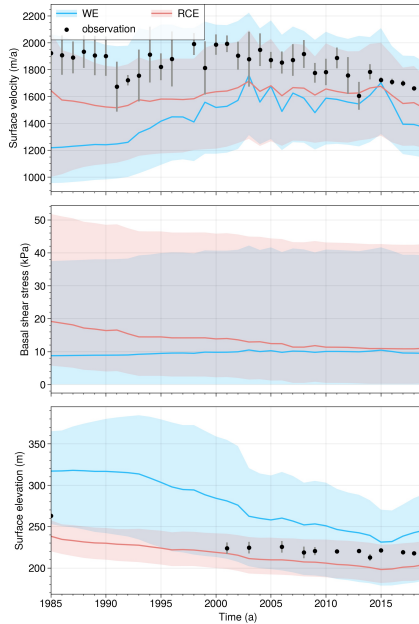
Initial period : elevations



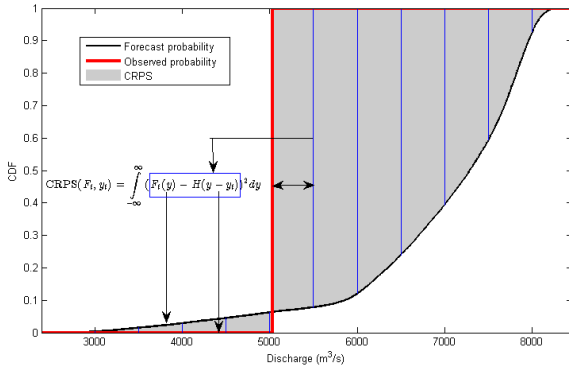
Surface elevation bias of the **ensemble mean** during the period 1985-2005







Scores

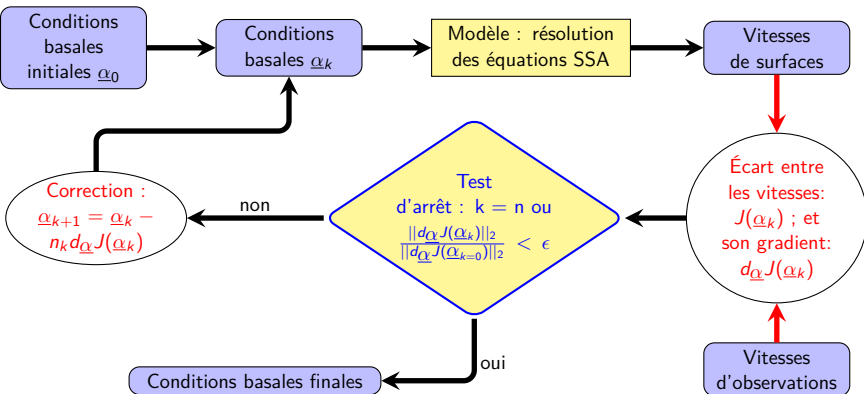


Calculation of the CRPS (source : Matlab site, by Durga Lal Shrestha)

Two different scores for simulation :

- The Root Mean Square Error allow us to compare simulations one by one
- The Continuous Rank Probability Score allow us to compare different ensemble of simulation : if we are overconfident, we will have a bad score if the observation is outside of the spread. However, if we catch the observation, a thinner spread will have a better score.

Inverse method



Add a regularisation function to smooth the obtained solution and a flow divergence function to minimise the difference between the observed and modelled one. Weighting by 2 coefficients: λ_{reg} and λ_{div} .

Diagram of how the inverse method works in glaciology, inspired by <https://interstices.info/>

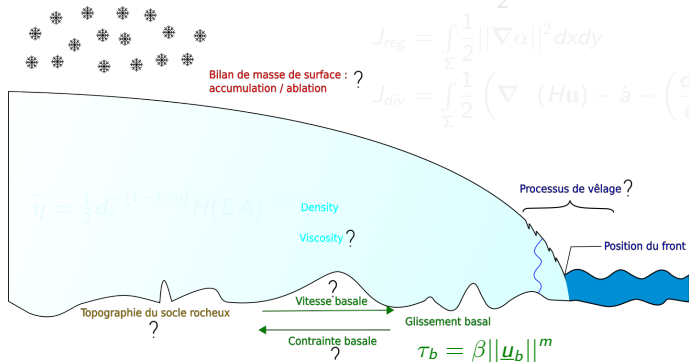
Inverse method uncertainties

$$J = J_0 + \lambda_{reg} J_{reg} + \lambda_{div} J_{div}$$

$$J_0 = \sum_{i=1}^{N_{obs}} \frac{1}{2} (\|\mathbf{u}_i^{mod} - \mathbf{u}_i^{obs}\|)^2$$

$$J_{reg} = \int_{\Sigma} \frac{1}{2} \|\nabla \alpha\|^2 dx dy$$

$$J_{div} = \int_{\Sigma} \frac{1}{2} \left(\nabla \cdot (H\mathbf{u}) - \dot{a} - \left(\frac{dH}{dt} \right)_{obs} \right)^2 dx dy$$



Uncertain parameters :

- friction law (m)
- viscosity (E)
- year of observation velocity (u_{obs})
- weight of cost functions (λ_{reg} et λ_{div})

Processes influencing glacial dynamics and their uncertainties

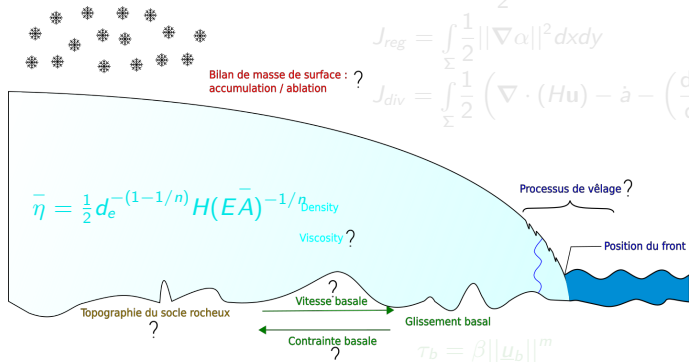
Inverse method uncertainties

$$J = J_0 + \lambda_{reg} J_{reg} + \lambda_{div} J_{div}$$

$$J_0 = \sum_{i=1}^{N_{obs}} \frac{1}{2} (\|\mathbf{u}_i^{mod} - \mathbf{u}_i^{obs}\|)^2$$

$$J_{reg} = \int_{\Sigma} \frac{1}{2} \|\nabla \alpha\|^2 dx dy$$

$$J_{div} = \int_{\Sigma} \frac{1}{2} \left(\nabla \cdot (H\mathbf{u}) - \dot{a} - \left(\frac{dH}{dt} \right)_{obs} \right)^2 dx dy$$



Uncertain parameters :

- friction law (m)
- viscosity (E)
- year of observation velocity (u_{obs})
- weight of cost functions (λ_{reg} et λ_{div})

Processes influencing glacial dynamics and their uncertainties

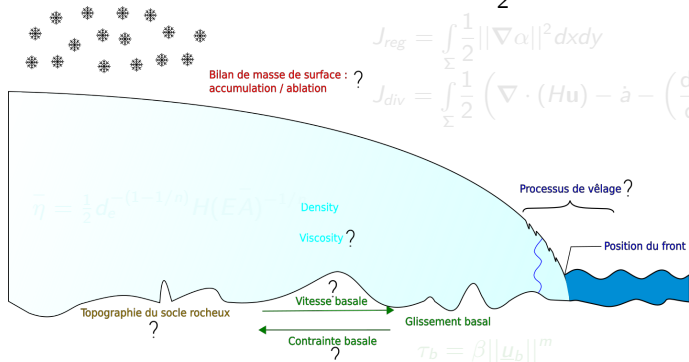
Inverse method uncertainties

$$J = J_0 + \lambda_{reg} J_{reg} + \lambda_{div} J_{div}$$

$$J_0 = \sum_{i=1}^{N_{obs}} \frac{1}{2} (\|\mathbf{u}_i^{mod} - \mathbf{u}_i^{obs}\|)^2$$

$$J_{reg} = \int_{\Sigma} \frac{1}{2} \|\nabla \alpha\|^2 dx dy$$

$$J_{div} = \int_{\Sigma} \frac{1}{2} \left(\nabla \cdot (H\mathbf{u}) - \dot{a} - \left(\frac{dH}{dt} \right)_{obs} \right)^2 dx dy$$



Uncertain parameters :

- friction law (m)
- viscosity (E)
- year of observation velocity (u_{obs})
- weight of cost functions (λ_{reg} et λ_{div})

Processes influencing glacial dynamics and their uncertainties

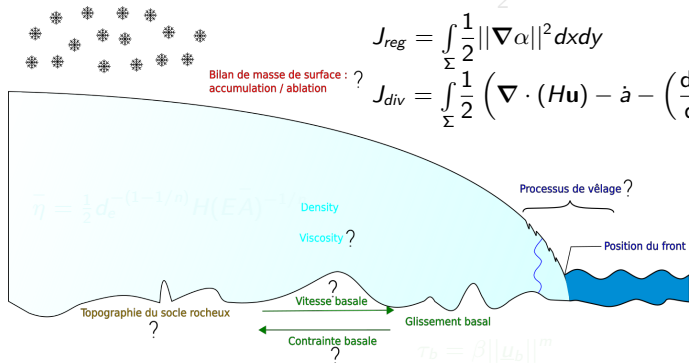
Inverse method uncertainties

$$J = J_0 + \lambda_{reg} J_{reg} + \lambda_{div} J_{div}$$

$$J_0 = \sum_{i=1}^{N_{obs}} \frac{1}{2} (\|u_i^{mod} - u_i^{obs}\|)^2$$

$$J_{reg} = \int_{\Sigma} \frac{1}{2} \|\nabla \alpha\|^2 dx dy$$

$$J_{div} = \int_{\Sigma} \frac{1}{2} \left(\nabla \cdot (Hu) - \dot{a} - \left(\frac{dH}{dt} \right)_{obs} \right)^2 dx dy$$

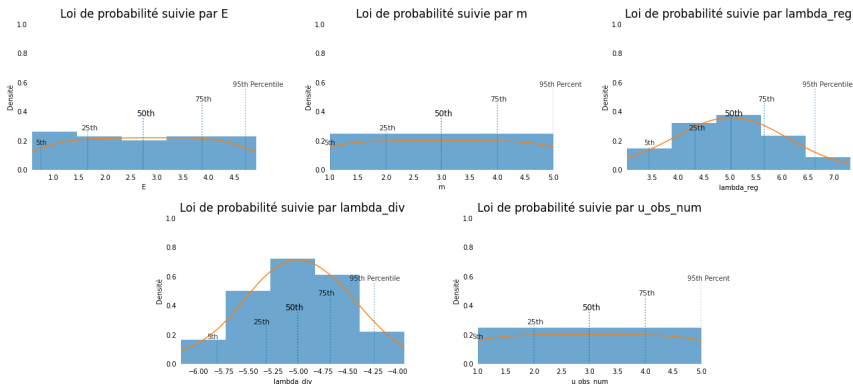


Uncertain parameters :

- friction law (m)
- viscosity (E)
- year of observation velocity (u_{obs})
- weight of cost functions (λ_{reg} et λ_{div})

Processes influencing glacial dynamics and their uncertainties

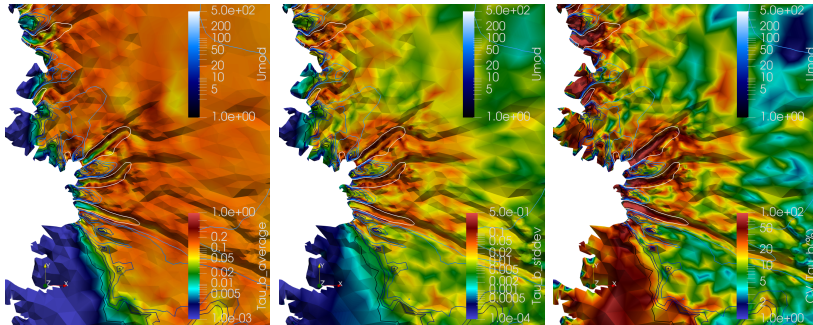
Probability laws



Les lois de probabilités de chaque paramètre

2. Putting uncertainties into the form of a probability law
3. Random draw (Latin Hypercube actually) for each of the 40 parameter sets (40 simulations)
4. For each set of parameters, an inversion is performed to obtain a friction field

Final results

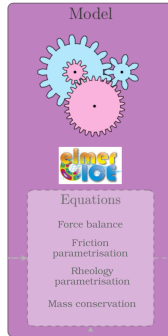


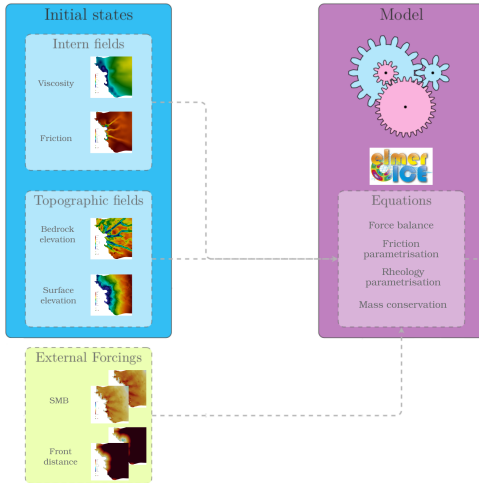
5. We obtain a set of friction fields and basal stresses that represent our uncertainty about these parameters.

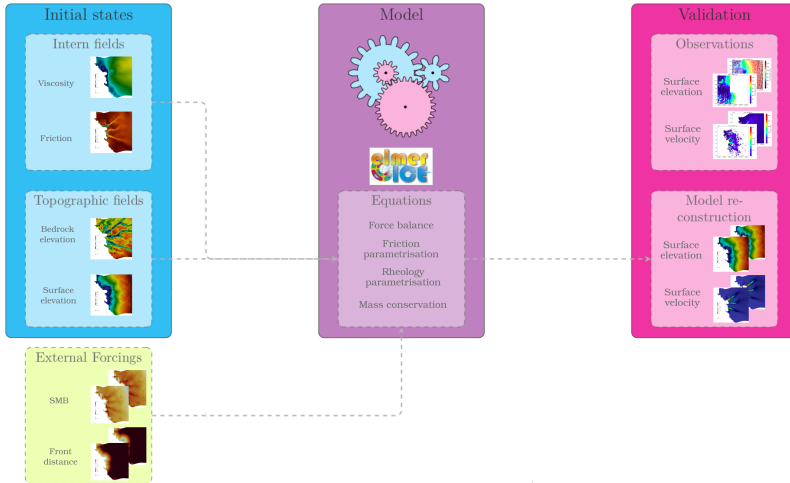
Mean, standard deviation (MPa) and coefficient of variation (in %) of basal stresses for the set of inversions in the Upernavik area.

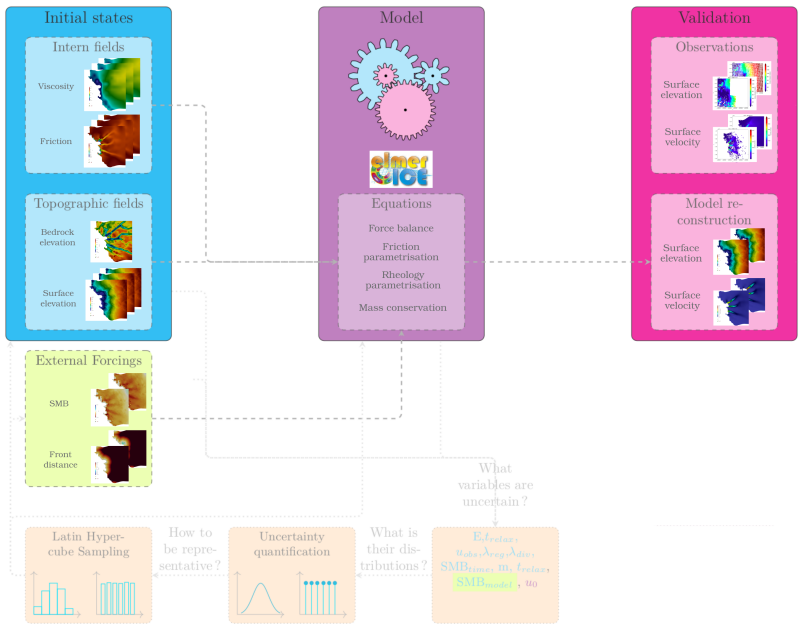
Take home messages

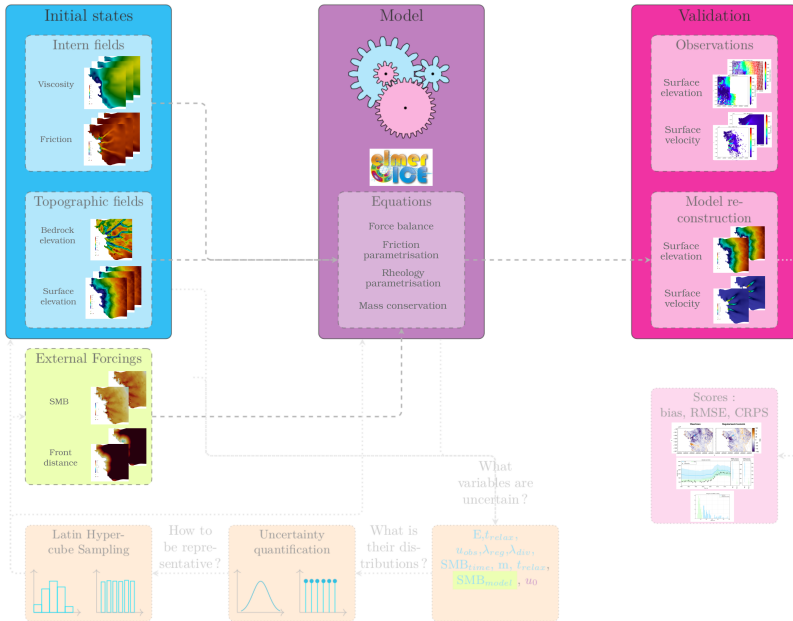
1. Creation of an ensemble of friction fields representative of the uncertainty on this parameter (low confidence due to not taking into account topography changes and modelling uncertainties).
2. Reusable ensemble for simulations/projections with Elmer/Ice (or others ISM ;)).

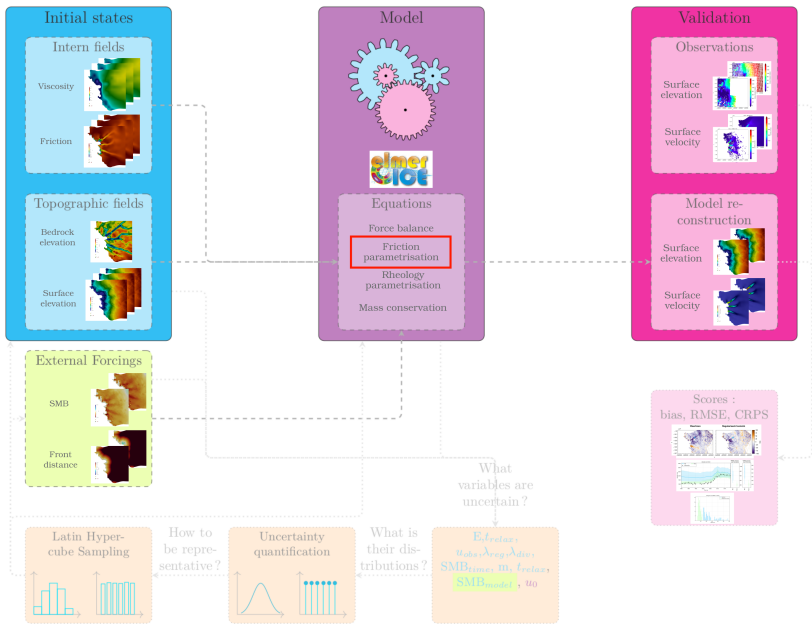


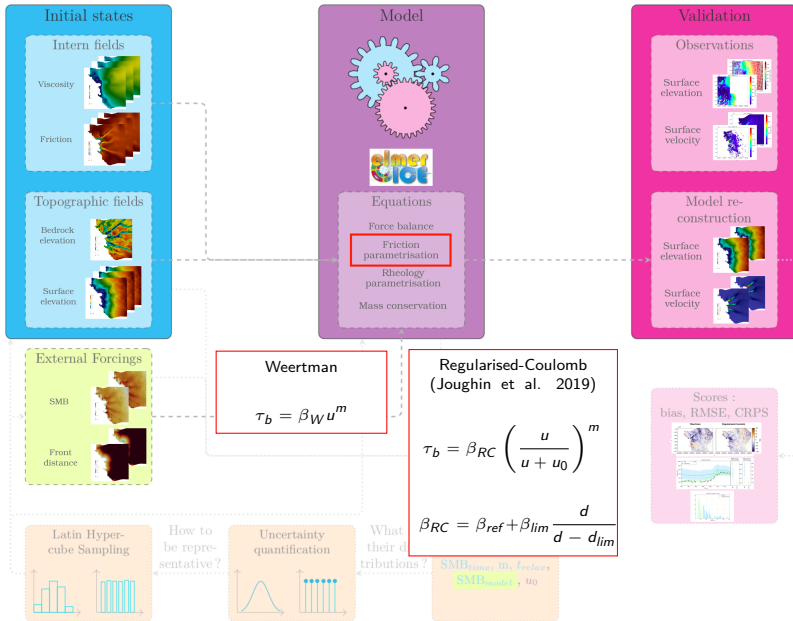












Experimental design

1/ for a given friction law, the parameters that vary in the ensemble :

- Initial friction field obtain by **inverse method**, calibrated with data of the **2010-2020 period**
- Initial surface elevation obtain after a **relaxation**
- Forcing (SMB)
- Dynamic laws (friction, rheology)

Experimental design

1/ for a given friction law, the parameters that vary in the ensemble :

- Initial friction field obtain by **inverse method**, calibrated with data of the **2010-2020 period**
- Initial surface elevation obtain after a **relaxation**
- Forcing (SMB)
- Dynamic laws (friction, rheology)

2/ **2 ensemble** simulations with **different friction parameterisations** where the position of the front and the surface mass balance are forced **between 1985 and 2019**. Each ensemble has **120 members**.

Experimental design

1/ for a given friction law, the parameters that vary in the ensemble :

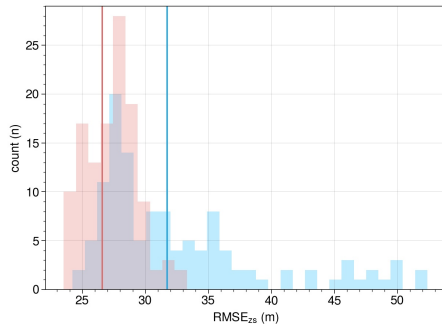
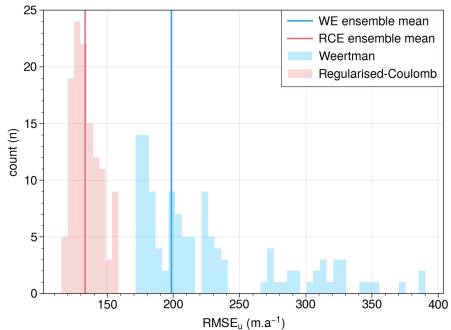
- Initial friction field obtain by **inverse method**, calibrated with data of the **2010-2020 period**
- Initial surface elevation obtain after a **relaxation**
- Forcing (SMB)
- Dynamic laws (friction, rheology)

2/ **2 ensemble** simulations with **different friction parameterisations** where the position of the front and the surface mass balance are forced **between 1985 and 2019**. Each ensemble has **120 members**.

3/ Validation with data :

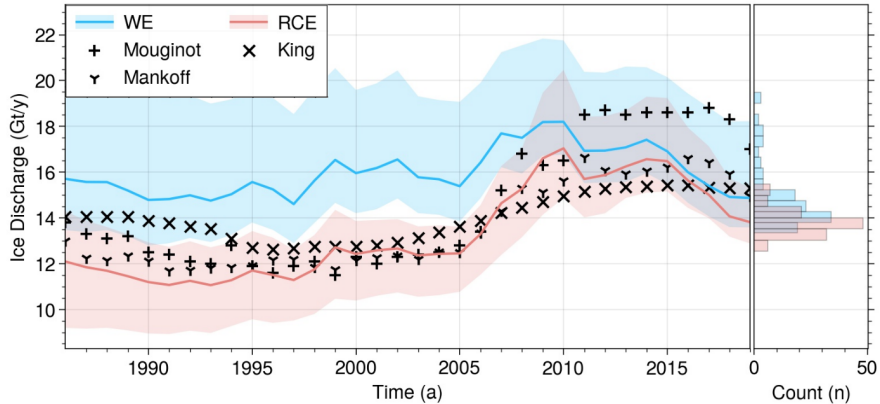
- Surface elevation
- Surface velocity
- Ice discharge
- Volume changes

In a nutshell



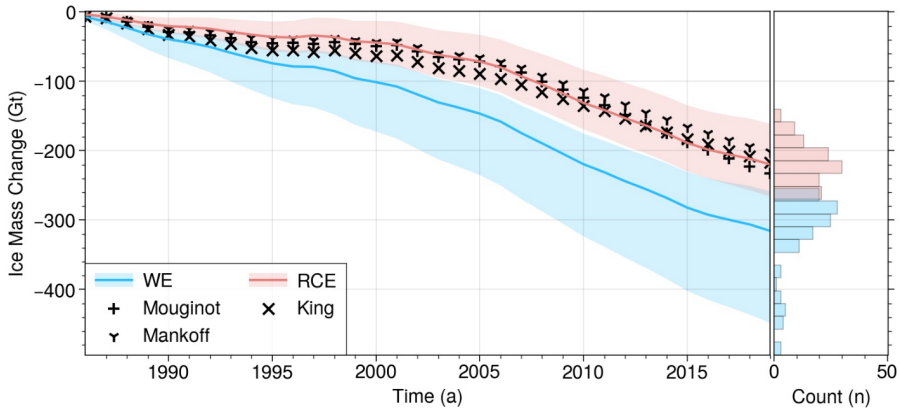
Distribution of WE and RCE RMSE for the surface velocity (left) and the surface elevation (right) in the validation area over the full period (1985-2019)

Global variables: Ice discharge



Ice discharge evolution for **Weertman Ensemble (WE)** and **Regularised-Coulomb Ensemble (RCE)** between 1985 and 2019

Global variables: Ice mass loss



Ice mass change for Weertman Ensemble (WE) and Regularised-Coulomb Ensemble (RCE) between 1985 and 2019

Take-home messages

1. Weertman : unable to reproduce observations under geometric conditions different from those of the calibration + unable to reproduce the acceleration.
2. Regularised-Coulomb with a parameterisation that reduces the friction near the front addresses these 2 issues.
3. Choices of extrapolation can have huge impact for historical reconstruction.
4. A good reproduction of velocities and partly of elevations allows to reproduce the observations of ice discharges and ice volume changes.

Jager et al., 2023, JOG, (submitted)

Take-home messages

1. Weertman : unable to reproduce observations under geometric conditions different from those of the calibration + unable to reproduce the acceleration.
2. Regularised-Coulomb with a parameterisation that reduces the friction near the front addresses these 2 issues.
3. Choices of extrapolation can have huge impact for historical reconstruction.
4. A good reproduction of velocities and partly of elevations allows to reproduce the observations of ice discharges and ice volume changes.

Jager et al., 2023, JOG, (submitted)

Take-home messages

1. Weertman : unable to reproduce observations under geometric conditions different from those of the calibration + unable to reproduce the acceleration.
2. Regularised-Coulomb with a parameterisation that reduces the friction near the front addresses these 2 issues.
3. Choices of extrapolation can have huge impact for historical reconstruction.
4. A good reproduction of velocities and partly of elevations allows to reproduce the observations of ice discharges and ice volume changes.

Jager et al., 2023, JOG, (submitted)

Take-home messages

1. Weertman : unable to reproduce observations under geometric conditions different from those of the calibration + unable to reproduce the acceleration.
 2. Regularised-Coulomb with a parameterisation that reduces the friction near the front addresses these 2 issues.
 3. Choices of extrapolation can have huge impact for historical reconstruction.
 4. A good reproduction of velocities and partly of elevations allows to reproduce the observations of ice discharges and ice volume changes.
- Jager et al., 2023, JOG, (submitted)