



1 **Glacier thickness estimations of alpine glaciers using data and**
2 **modeling constraints**

3
4 Lisbeth Langhammer¹
5 Melchior Grab^{1,2}
6 Andreas Bauder²
7 Hansruedi Maurer^{1*}

8
9
10 ¹ Institute of Geophysics, ETH Zurich, Switzerland

11 ² Laboratory of Hydraulics, Hydrology and Glaciology (VAW), ETH Zurich, Switzerland

12 * Corresponding author (hansruedi.maurer@erdw.ethz.ch)

13
14
15 **Abstract**

16
17 Advanced knowledge of the ice thickness distribution within glaciers is of fundamental
18 importance for several purposes, such as water resource management and studying the
19 impact of climate change. Ice thicknesses can be modeled using ice surface features, but
20 the resulting models can be prone to considerable uncertainties. Alternatively, it is
21 possible to measure ice thicknesses, for example, with ground-penetrating-radar (GPR).
22 Such measurements are typically restricted to a few profiles, with which it is not possible
23 to obtain spatially unaliased subsurface images. We developed the Glacier Thickness
24 Estimation algorithm (GlaTE), which optimally combines modeling results and measured
25 ice thicknesses in an inversion procedure to obtain overall thickness distributions.
26 Properties and benefits of GlaTE are demonstrated with three case studies performed on
27 different types of alpine glaciers. In all three cases, subsurface models could be found
28 that are consistent with glaciological modeling and GPR data constraints. Since acquiring
29 GPR data on glaciers can be an expensive endeavor, we additionally employed elements
30 of sequential optimized experimental design (SOED) for determining cost-optimized
31 GPR survey layouts. The calculated benefit-cost curves indicate that a relatively large
32 amount of data can be acquired, before redundant information is collected with any
33 additional profiles and it becomes increasingly expensive to obtain further information.
34 Only at one out of the three test sites this level was reached.

35



36 1 Introduction

37

38 Estimating the amount of the glacier ice around the globe is crucial, for example, for sea-
39 level predictions, securing fresh water recourses, designing hydropower facilities in high-
40 alpine environments, and predicting the occurrence of glacier-related natural hazards. For
41 estimating the overall glacier ice mass and its local distribution, (i) knowledge of the
42 glacier outline, (ii) its surface topography and (iii) the underlying bedrock topography is
43 required. The first two quantities can be observed with aerial and satellite imagery, but
44 the bedrock topography is more difficult to determine.

45

46 The conceptually simplest option includes drilling boreholes through the glacier ice (e.g.,
47 Iken, 1988). This approach offers ground-truth information, but only a very sparse
48 observation grid can be obtained with realistic efforts. Therefore, geophysical methods
49 have been employed for obtaining more detailed information. Due to the very high
50 electrical resistivity of glacier ice and the relatively high electromagnetic impedance
51 contrast between ice and bedrock material, ground-penetrating-radar (GPR) techniques,
52 also referred to as radio-echo-sounding (RES), have been the primary choice for such
53 investigations (e.g., Evans, 1963). GPR data can either be acquired ground-based (e.g.,
54 Watts and England, 1976), or, more efficiently, using fixed-wing airplanes (e.g.,
55 Steinhage et al., 1999) or helicopters (e.g., Rutishauser et al., 2016).

56

57 Despite the powerful capabilities of modern GPR acquisition systems, it is still beyond
58 any practical limits to acquire spatially un-aliased 3D data sets. GPR data are therefore
59 collected only along a sparse network of profiles, which leaves considerable uncertainties
60 in the regions between the profiles.

61

62 To address this problem, glaciological modeling techniques have been established to
63 relate observable surface parameters to the thickness distribution of ice. One of the
64 earliest concepts was published by Nye (1952). He established a simple relationship
65 between the surface slope and ice thickness. During the past decades, more sophisticated
66 ice thickness modeling techniques have emerged rapidly. Various glaciological
67 constraints, such as mass conservation and/or the relation between basal shear stress and
68 ice thickness, were considered (e.g., Farinotti et al., 2009; Huss and Farinotti, 2012; Clarke
69 et al., 2013; Linsbauer et al., 2012; Morlighem et al., 2011). See Farinotti et al. (2017) for
70 a more complete review of most of the approaches published to date.

71

72 Due to inaccuracies of the observed data (GPR measurements, surface topography, etc.)
73 and/or inadequacies of the modeling approaches, modeled ice thicknesses cannot be
74 expected to be perfect. This can be considered by formulating ice thickness estimation as
75 an optimization problem, in which the discrepancies between observed and predicted data
76 are minimized (e.g., Morlighem et al., 2014). In this contribution, we follow an approach
77 similar to Morlighem et al. (2014), but with a different implementation. We introduce the
78 general framework of Glacier Thickness Estimation (GlaTE), with which modeling and
79 data constraints can be combined in an appropriate fashion. After introducing the
80 underlying theory, we demonstrate the performance of the GlaTE inversion procedure
81 with three case studies. In the second part of the paper, we employ elements of GlaTE to



82 address the experimental design problem. Here, we seek a measured data set that offers
83 maximum information content at minimal costs. For that purpose, we consider
84 sequentially optimized experimental design (SOED) techniques (e.g., Maurer et al.,
85 2017). The paper concludes with a critical review of potential problems and shortcomings
86 of GlaTE and the associated SOED procedures, and we outline options to address these
87 issues and propose useful extensions of the methodology.

88
89

90 **2 GlaTE inversion algorithm**

91

92 **2.1. Theory**

93

94 The basic idea of GlaTE inversions is to combine observable data with glaciological
95 modeling constraints, whereby it is attempted to consider appropriately the uncertainties
96 associated with both types of information. All constraints are formulated, such that they
97 can be integrated into a single system of equations, which can be solved with an
98 appropriate solver.

99

100 The first type of constraints includes the GPR data. They can be written in the form of

101

$$102 \quad (1) \quad \mathbf{G}\mathbf{h}^{\text{est}} = \mathbf{h}^{\text{GPR}},$$

103

104 where \mathbf{h}^{est} is a vector including the unknown (*estimated*) ice thicknesses at M locations
105 (typically defined on a regular grid R on a glacier), and \mathbf{G} is a $N^{\text{GPR}} \times M$ matrix with
106 ones in its main diagonal and zeros everywhere else (N^{GPR} = number of available GPR
107 data points, M = number of elements in \mathbf{h}^{est}). The vector \mathbf{h}^{GPR} of length N^{GPR} includes the
108 GPR-based thickness estimates. Since the GPR data usually do not coincide with the grid
109 points of R , the values \mathbf{h}^{GPR} are obtained by interpolating or extrapolating the GPR data
110 to the nearest grid points of R .

111

112 Next, we consider glaciological modeling constraints. In principle, any of the algorithms
113 proposed in the literature can be employed. Here, we follow closely the approach
114 described in Clarke et al. (2013). Input data include a digital terrain model (DTM,
115 defined on R) and the glacier outline.

116

117 First, the glacier area is subdivided into so-called flowsheds using the Matlab TOPO-
118 Toolbox (Schwanghart and Kuhn, 2010). The subsequent procedure is applied to each
119 flowshed individually (see comments in Clarke et al. (2013) for more information on the
120 flowshed subdivision).

121

122 Next, the apparent mass-balance, defined as

123

$$124 \quad (2) \quad \tilde{\mathbf{b}} = \dot{\mathbf{b}} - \frac{\partial \mathbf{h}}{\partial t},$$

125



126 with $\dot{\mathbf{h}}$ being the mass balance rate, and $\frac{\partial \mathbf{h}}{\partial t}$ the thickness change rate, is either

127 determined by measuring $\dot{\mathbf{h}}$ and $\frac{\partial \mathbf{h}}{\partial t}$, or computed via the condition

128

129 (3)
$$\int_{\Omega_G} \tilde{\mathbf{b}} = 0 ,$$

130

131 where Ω_G denotes the glacier area (see Farinotti et al. (2009) for more details). In a next
132 step, the flowsheds are partitioned into a prescribed number of elevation zones D_i

133 ($i = 1 \dots \text{number of elevation zones}$), for which the ice discharge Q_i through its lower

134 boundary is computed using

135

136 (4)
$$Q_i = \int_{\Omega_{D_i}} \tilde{\mathbf{b}} ,$$

137

138 where Ω_{D_i} is the area of zone D_i . Following Clarke et al. (2013), the basal shear stress τ

139 can then be obtained via the relationship

140

141 (5)
$$\tau = \left[\frac{(n+2) \rho g \sin(\phi)^2 \xi \mathbf{q}}{2A} \right]^{1/(n+2)}$$

142

143 The parameters n , ρ , g and A denote the exponent of Glen's flow law, ice density, gravity
144 acceleration and creep rate factor, respectively (e.g., Cuffey and Patterson, 2010). The
145 factor ξ denotes the creeping contribution (relative to basal sliding) to the ice flux

146 ($0 < \xi < 1$), and \mathbf{q} is the specific ice discharge $q_i = \bar{Q}_i / l_i$, where l_i is the length of the

147 lower boundary of D_i , and \bar{Q}_i is the average of Q_i within D_i . Likewise, the angle ϕ

148 represents the surface slope averaged along the lower boundary of D_i .

149

150 As outlined in Kamb and Echelmeyer (1986), the physics of ice flow can be incorporated
151 into the modeling procedure by applying "longitudinal averaging" of the shear stress (i.e.,
152 along the flow direction). We apply this procedure to the results obtained with

153 Equation (5). Finally, the ice thicknesses $\hat{\mathbf{h}}^{\text{glac}}$ (*glac* stands for glaciological modeling
154 constraints) are obtained using

155

156 (6)
$$\hat{\mathbf{h}}^{\text{glac}} = \frac{\tau^*}{\rho g \sin(\theta)} ,$$

157

158 where τ^* denotes the basal shear stress after longitudinal averaging.

159



160 Some of the parameters in Equation (5) may be subject to considerable uncertainties. For
 161 example, the parameter ξ is often poorly known, and it is not guaranteed that the values
 162 of the parameters A and n , usually taken from the literature, are accurate. Typically, n is
 163 reasonable well constrained, but A can vary over orders of magnitudes. Therefore, the
 164 overall magnitudes of $\hat{\mathbf{h}}^{\text{glac}}$ may be significantly over- or under-estimated. This can be
 165 considered with an additional factor α_{GPR} , yielding

$$166 \quad (7) \quad \mathbf{h}^{\text{glac}} = \alpha_{GPR} \hat{\mathbf{h}}^{\text{glac}} .$$

168 α_{GPR} can be computed with an optimization procedure that minimizes

$$170 \quad \left| \text{mean} \left(\mathbf{h}^{\text{GPR}} - \alpha_{GPR} \hat{\mathbf{h}}^{\text{glac}} \right) \right| .$$

171 The correction factor α_{GPR} accounts for some inadequacies of Equation (5), but it is still
 172 possible that there are systematic differences between \mathbf{h}^{GPR} and \mathbf{h}^{glac} . To avoid the
 173 resulting inconsistencies, we consider not the absolute values \mathbf{h}^{glac} , but the spatial
 174 gradients $\nabla \mathbf{h}^{\text{glac}}$ as glaciological constraints, resulting in

$$177 \quad (8) \quad \mathbf{L} \mathbf{h}^{\text{est}} = \nabla \mathbf{h}^{\text{glac}} ,$$

178 where \mathbf{L} is a difference operator of dimension $M \times M$.

181 Further constraints can be imposed via the glacier boundaries that can be determined
 182 from aerial or satellite images or ground observations. They are considered in the form of
 183 the equation

$$185 \quad (9) \quad \mathbf{B} \mathbf{h}^{\text{est}} = 0 ,$$

186 where \mathbf{B} is a $M \times M$ matrix with ones at appropriate places in its main diagonal.

188 Depending on the discretization of the glacier models (i.e., the discretization of R), the
 189 constraints described above, may allow the resulting system of equations to be solved
 190 unambiguously. However, in most cases, there will be still a significant underdetermined
 191 component, that is, there will be many solutions that explain the data equally well. This
 192 requires regularization constraints to be applied (e.g., Menke, 2012). A common strategy
 193 for regularizing such problems is to follow the Occam's principle, which identifies the
 194 "simplest" solution out of the many possible solutions (Constable et al., 1987). Here, we
 195 define "simplicity" in terms of structural complexity, that is, we seek a smooth model.
 196 This can be achieved via a set of smoothing equations of the form

$$199 \quad (10) \quad \mathbf{S} \mathbf{h}^{\text{est}} = 0 ,$$

200 where \mathbf{S} is a $M \times M$ smoothing matrix.
 201



202
 203
 204

All the constraints can now be merged into a single system of equations

$$(11) \quad \begin{pmatrix} \lambda_1 \mathbf{G} \\ \lambda_2 \mathbf{L} \\ \lambda_3 \mathbf{B} \\ \lambda_4 \mathbf{S} \end{pmatrix} \mathbf{h}^{\text{est}} = \begin{pmatrix} \lambda_1 \mathbf{h}^{\text{GPR}} \\ \lambda_2 \nabla \mathbf{h}^{\text{glac}} \\ 0 \\ 0 \end{pmatrix},$$

206

207 where the parameters λ_1 to λ_4 allow a weighting according to the confidence into
 208 individual contributions. Parameter λ_3 is not critical and can be fixed to an appropriate
 209 value (e.g., 1.0). The magnitudes of the remaining three parameters must be chosen, such
 210 that the system of equations in (11) is solvable. However, it also needs to be considered
 211 that all the constraints related to λ_1 , λ_2 and λ_4 may be subject to significant inaccuracies.
 212 It is difficult to predict the accuracy of the modeling constraints and to judge the
 213 appropriateness of the smoothing constraints, but the accuracy of the GPR data
 214 constraints, subsequently denoted as $\boldsymbol{\varepsilon}^{\text{GPR}}$, can usually be quantified. Therefore, λ_1 , λ_2
 215 and λ_4 have to be chosen, such that the discrepancy of the GPR data ($\|\mathbf{G}\mathbf{h}^{\text{est}} - \mathbf{h}^{\text{GPR}}\|$) is
 216 of the order of $\boldsymbol{\varepsilon}^{\text{GPR}}$, and the GPR data are thus neither under- nor over-fitted. We have
 217 implemented this by choosing the magnitudes of λ_1 , λ_2 and λ_4 , such that a prescribed
 218 percentage of the GPR data (e.g., 95%) satisfies $\|\mathbf{G}\mathbf{h}^{\text{est}} - \mathbf{h}^{\text{GPR}}\| < \boldsymbol{\varepsilon}^{\text{GPR}}$.

219

220 This can be achieved with different strategies. One option is to fix λ_2 and λ_4 , and to
 221 vary λ_1 until the condition, mentioned above, is met. Alternatively, it is possible to fix
 222 the pairs λ_1/λ_4 or λ_1/λ_2 and to vary λ_2 or λ_4 . Choice of the most appropriate strategy

223 depends on the uncertainties associated with the individual contributions in Equation (11)

224

225

226 The dimension of the system of equations in (11) can be very large, but the matrices \mathbf{G} ,
 227 \mathbf{L} , \mathbf{B} and \mathbf{S} are all extremely sparse. Therefore, sparse matrix solvers, such as LSQR
 228 (Paige and Saunders, 1982) can solve such systems efficiently for \mathbf{h}^{est} .

229

230

231 2.2 Performance tests

232

233 For testing the GlaTE inversion algorithm, we investigated glacier ice thickness at three
 234 sites in the Swiss Alps (Figure 1). The first site is Morteratschgletscher (Figure 1a).
 235 Lying at altitudes between 2050 and 4000 m a.s.l. (Zekollari et al., 2013), the glacier has
 236 a typical valley-glacier shape and is located in the Engadin region of Switzerland. In
 237 2015, the tributary glacier Vadret Pers in the east detached from the main trunk of
 238 Morteratschgletscher, but we continue to treat both glaciers as a connected system, since



239 the last available outline of the glaciers in 2015 shows the remnant of the former
240 connection. In 2010, the glacier system covered an area of $\approx 15 \text{ km}^2$, and it had a length
241 of $\approx 7.4 \text{ km}$.

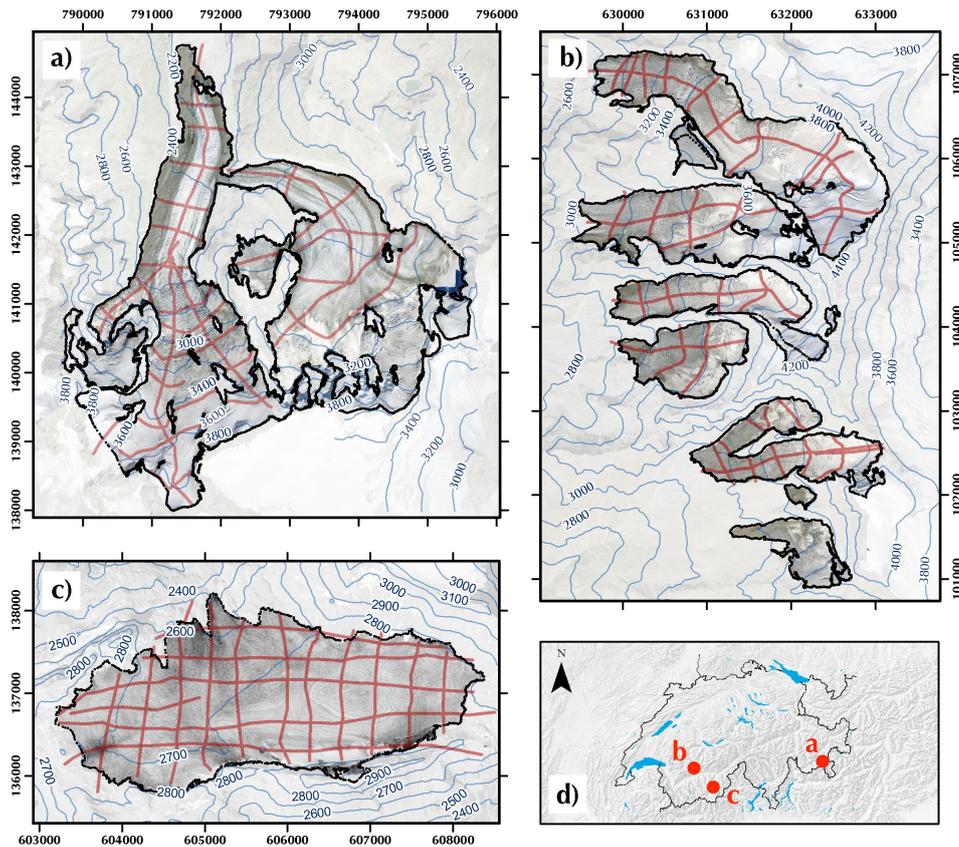
242
243 The second site, Glacier Plaine Morte (2400-3000 m a.s.l., (Figure 1b), is the largest
244 plateau glacier in the European Alps (Huss et al., 2013). The surface slope is shallow
245 with slope angles less than 4° and a short glacier tongue draining towards the North.

246
247 The third site is a cluster of small valley flank and cirque-type glaciers on the eastern
248 flank of the Matter valley (Figure 1c) below the Dom peak. From North to South, the
249 glaciers are named Hohbärggletscher, Festigletscher, Kingletscher and
250 Weingartengletscher. The Hohbärggletscher is the largest (2800-4500 m a.s.l.) and
251 longest of the group.

252
253 For all sites, the recorded GPR profiles are shown in Figure 1. The GPR data are a
254 composite of several campaigns. Most of the data were recorded with the dual
255 polarization system AIR-ETH (Langhammer et al., 2018). On the Glacier Plaine Morte, a
256 grid of profiles was acquired in 2016, and on the Morteratschgletscher and in the Dom
257 Region in 2017. The data were processed as described in Grab et al. (2018), and the
258 bedrock depths and the corresponding ice thicknesses were obtained from the migrated
259 GPR images.

260
261 As input data for the glacier models, surface topography and an outline of the individual
262 glaciers was required. As surface topography, we used the swissALTID3D (DTM, Digital
263 Terrain Model Release 2017 © swisstopo (JD100042)). The most recent version covering
264 the individual glaciers was extracted and down-sampled to 10 m resolution. The outline
265 represents the extension of the glacier in 2015-2016. DTM and glacier outlines are
266 displayed in Figure 1.

267
268



269
270 *Figure 1: Satellite images and surface topography isolines of the glaciers investigated.*
271 *(a) Morteratschgletscher, (b) Glacier Plaine Morte and (c) Dom region. The Swiss map*
272 *in the bottom right panel indicates the locations of the glaciers. GPR profiles acquired*
273 *are shown in red. Orthophotos © 2017 swisstopo (JD100042). Coordinate system:*
274 *CH1903.*

275
276 Before applying GlaTE inversions to all field sites, we tested the different options for
277 determining λ_1 , λ_2 and λ_4 , using the data from Morteratschgletscher. Figure 2 shows
278 the ice thicknesses distributions, (i) when only glaciological constraints are applied (\mathbf{h}^{glac} ,
279 Figure 2a), and (ii) when only GPR constraints are considered (\mathbf{h}^{GPR} , Figure 2b). In the
280 latter case, the thicknesses are obtained by natural neighbor interpolation from the GPR
281 data. Since no extrapolation was performed, not all glacierized regions have an ice
282 thickness estimate. Both images exhibit increased thicknesses in the western glacier, but
283 only the glaciological constraints indicate an overdeepening in the eastern one, thereby
284 indicating that the two models are inconsistent.
285



286 Figure 3 shows the results of the GlaTE inversions using either prescribed λ_1/λ_4 (Figure
 287 3a), λ_1/λ_2 (Figure 3c) or λ_2/λ_4 (Figure 3e) pairs. The corresponding difference plots
 288 (Figure 3b, d and f) refer to the deviation of the obtained thickness results compared with
 289 the thickness calculated with the glaciological approach. We varied the λ_2 and λ_4
 290 parameters by starting with very high values of 50, and by decreasing them successively
 291 until 95% of the GPR data met the condition $\|\mathbf{Gh}^{\text{est}} - \mathbf{h}^{\text{GPR}}\| < \boldsymbol{\varepsilon}^{\text{GPR}}$, where $\boldsymbol{\varepsilon}^{\text{GPR}}$ was
 292 estimated to be 5 m. In contrast, we started with a low value of 0.02 for variable λ_1 , and
 293 increased it successively until 95% of the data were fitted within the error $\boldsymbol{\varepsilon}^{\text{GPR}}$. Table 1
 294 summarizes the prescribed and estimated λ values.

295
 296 All three inversion strategies (i.e., either varying λ_2 , λ_4 or λ_1) yielded comparable
 297 results. Although the difference plots with respect to the glaciological model exhibit
 298 considerable differences (Figures 3b, 3d and 3f), the general shapes obtained with the
 299 glaciological constraints were well preserved in regions where the GPR data coverage
 300 was poor. From this first test, we conclude that (i) the GlaTE inversion approach works
 301 well, and (ii) that the strategy by which the values of λ are chosen is not critical.
 302

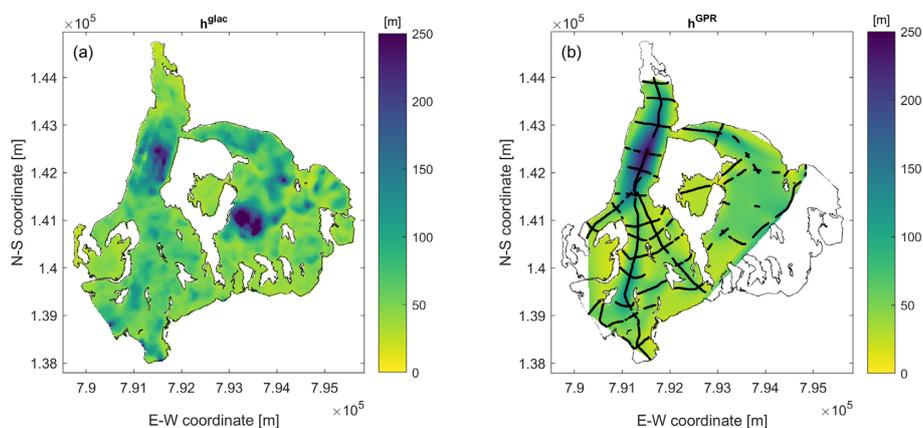
Inversion type	λ_1	λ_2	λ_3	λ_4	Figures
λ_1/λ_4 fixed	1	0.78	1	10	3a, 3b 4c, 4d
λ_1/λ_2 fixed	1	1	1	7.8	3c, 3d
λ_2/λ_4 fixed	1.28	1	1	10	3e, 3f
λ_1/λ_4 fixed	1	1.56	1	2	4a, 4b
λ_1/λ_4 fixed	1	0.00	1	50	4e, 4f

303 *Table 1: Weighting parameters λ employed for the GlaTE inversions shown in Figures 3*
 304 *and 4. Numbers marked red indicate varying parameters.*

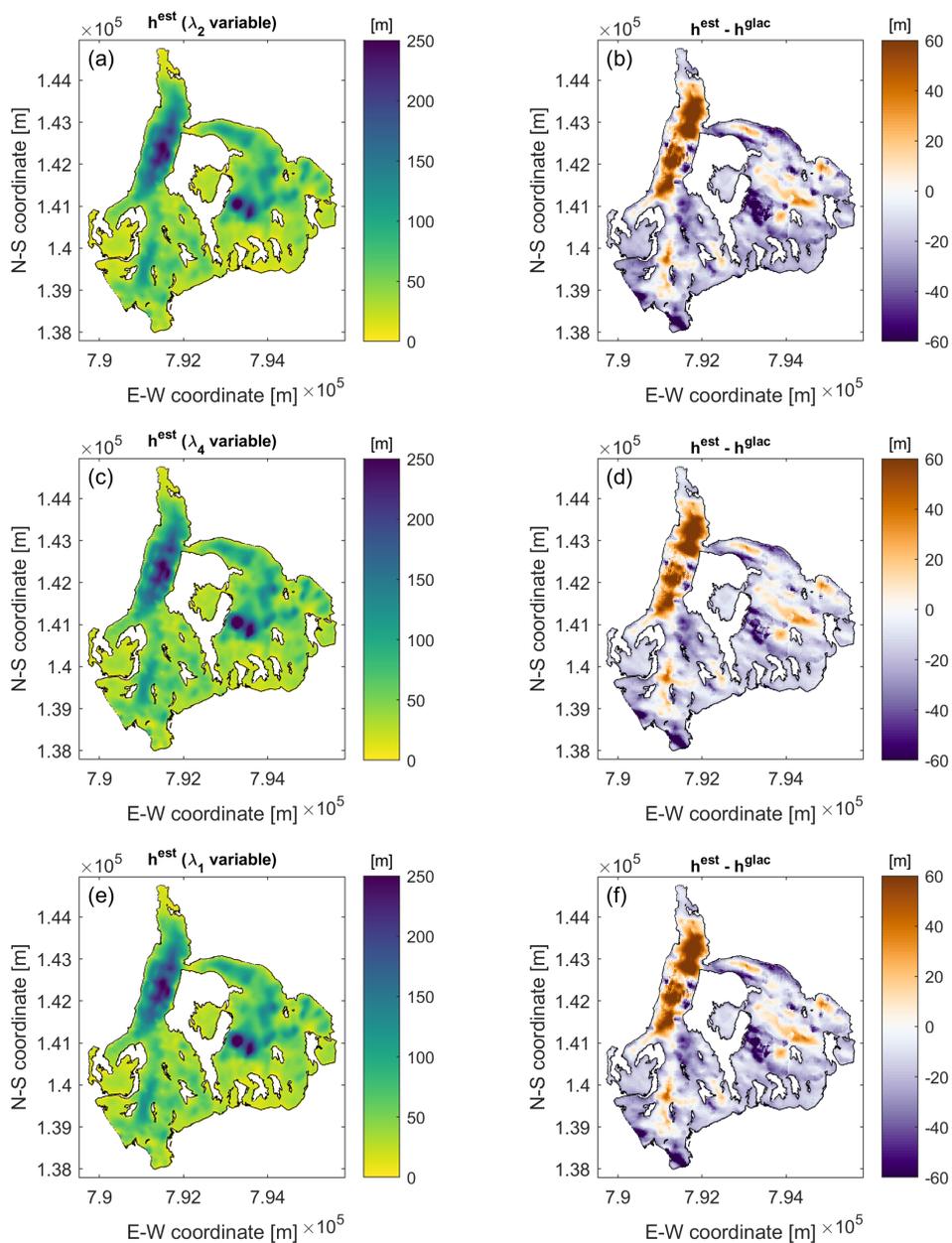
305
 306 It is instructive to study the effects of an overly small or large (fixed) λ_4 value. As shown
 307 in Table 1, we employed a prescribed value of 10 for λ_4 . This value was chosen by trial
 308 and error. There was a range of λ_4 values around 10 that yielded similar results (not
 309 shown). Choosing very low or high λ_4 values (i.e., $\lambda_4 = 2$ resp. $\lambda_4 = 50$) has a
 310 detrimental effect on the results, as shown in Figure 4. For $\lambda_4 = 2$, the inversion fits the
 311 ice thicknesses obtained from the GPR data only along the profile lines and maintains the
 312 glaciological modeling results in the remaining areas. This produces artificial features in
 313 the thickness map (Figure 4a). In contrast, $\lambda_4 = 50$ produces overly smooth images,
 314 which is obscuring small-scale variations from the glaciological constraints in regions



315 poorly covered by GPR data (Figure 4e). It is also noteworthy that even with $\lambda_2 = 0$ only
316 approx. 70% of the discrepancies $\|\mathbf{G}\mathbf{h}^{\text{est}} - \mathbf{h}^{\text{GPR}}\|$ were below ϵ^{GPR} (Figure 5e).
317

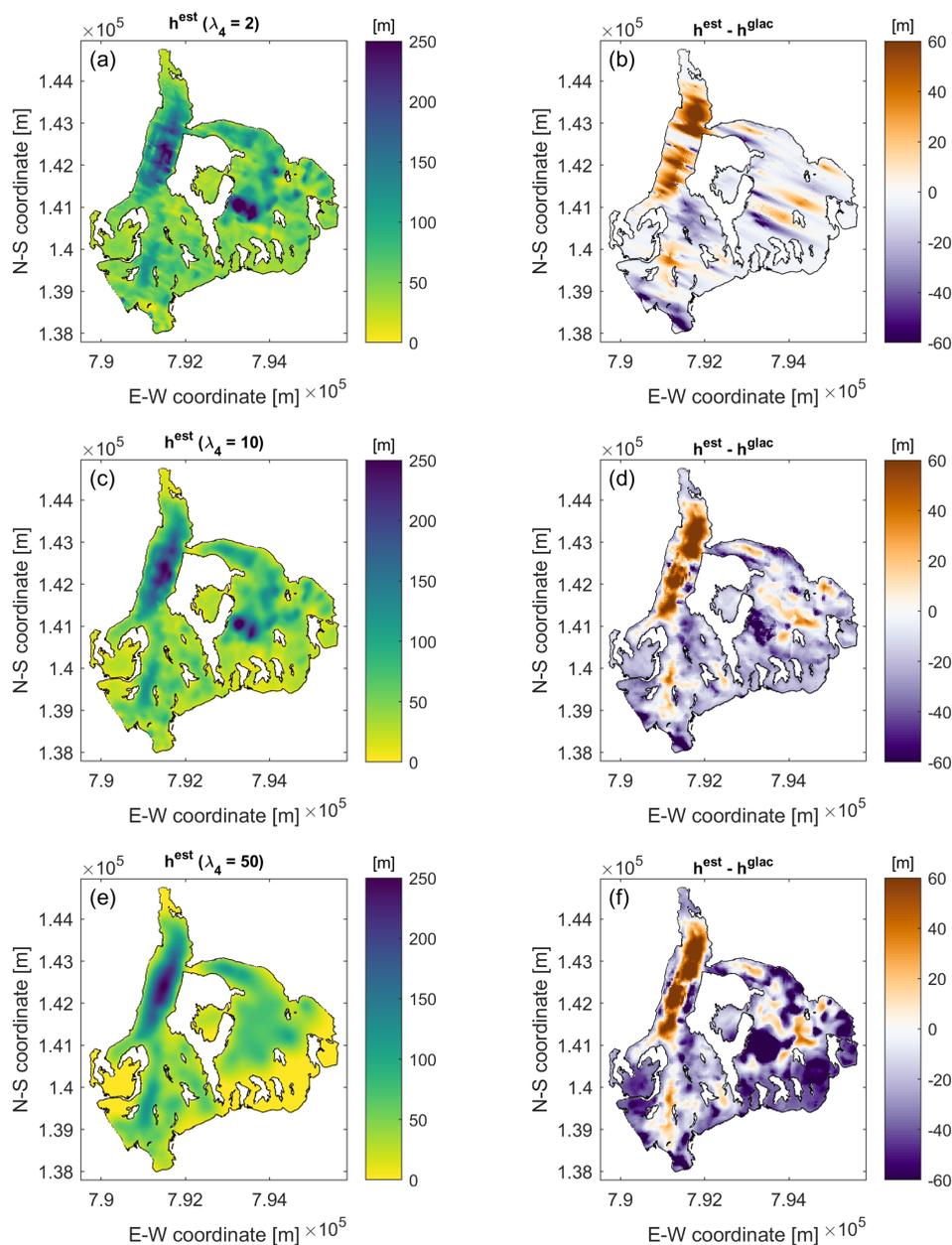


318
319 *Figure 2: Results from Morteratschgletscher only using (a) glaciological constraints and*
320 *(b) GPR constraints. Colors indicate ice thickness. Available thickness data obtained*
321 *from GPR profiles are marked with black lines.*



322
 323
 324
 325
 326

Figure 3: Results from Morteratschgletscher using different strategies for choosing weighting parameters λ (see text for more explanations). Left panels show ice thickness distributions and right panels show differences to glaciological model without GPR constraints (Figure 2a).



327
 328 *Figure 4: Results from Morteratschgletscher using fixed λ_1 and λ_4 values and varying*
 329 *λ_2 . (a) and (b) are the results for $\lambda_4 = 2$, (c) and (d) for $\lambda_4 = 10$ and (e) and (f) for*
 330 *$\lambda_4 = 50$. Left panels show ice thickness distributions and right panels show differences to*
 331 *glaciological model without GPR constraints (Figure 2a).*



332

333 In the following, we consider only the scheme in which λ_1/λ_2 is kept fixed

334 ($\lambda_1 = 1, \lambda_2 = 1$), and λ_4 is varied for analyzing the Glacier Plaine Morte and the Dom

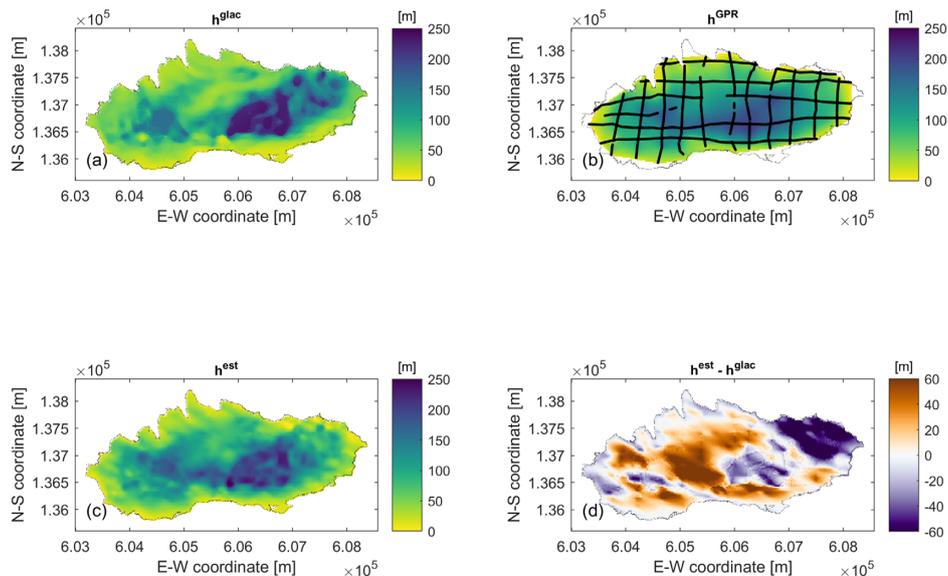
335 region data. The results are shown in Figures 5 and 6. For the Glacier Plaine Morte, the
336 glaciological model suggests a deep isolated trough slightly east of the center (Figure 5a).

337 This is not supported by the GPR data, which rather indicate a larger E-W oriented
338 elongated zone of increased thickness (Figure 5b). Such a feature is also contained in the
339 GlaTE inversion results (Figure 5c). Furthermore, the glaciological model in Figure 5a
340 overestimates the ice thickness in the northeastern part of the glacier.

341

342 Results from the Dom region show a relatively good match between the glaciological
343 model (Figure 6a) and the GlaTE inversion result (Figure 6c). The glaciological model
344 tends to underestimate the maximum thickness in the center of the glacier tongues, and to
345 overestimate the thickness towards the edges (Figure 6d). The isolated trough structures
346 (ice thickness > 200 m) in the northernmost glacier in the glaciological model (Figure 6a)
347 are only partially supported by the GPR data (Figure 6b) and the GlaTE inversion (Figure
348 6c). In the southernmost Weingartengletscher, no data constraints exist (Figure 6b). The
349 non-zero differences in this part (Figure 6d) are the result of the smoothing constraints.
350 Here, the thickness estimates from the glaciological model are thus more trustworthy.

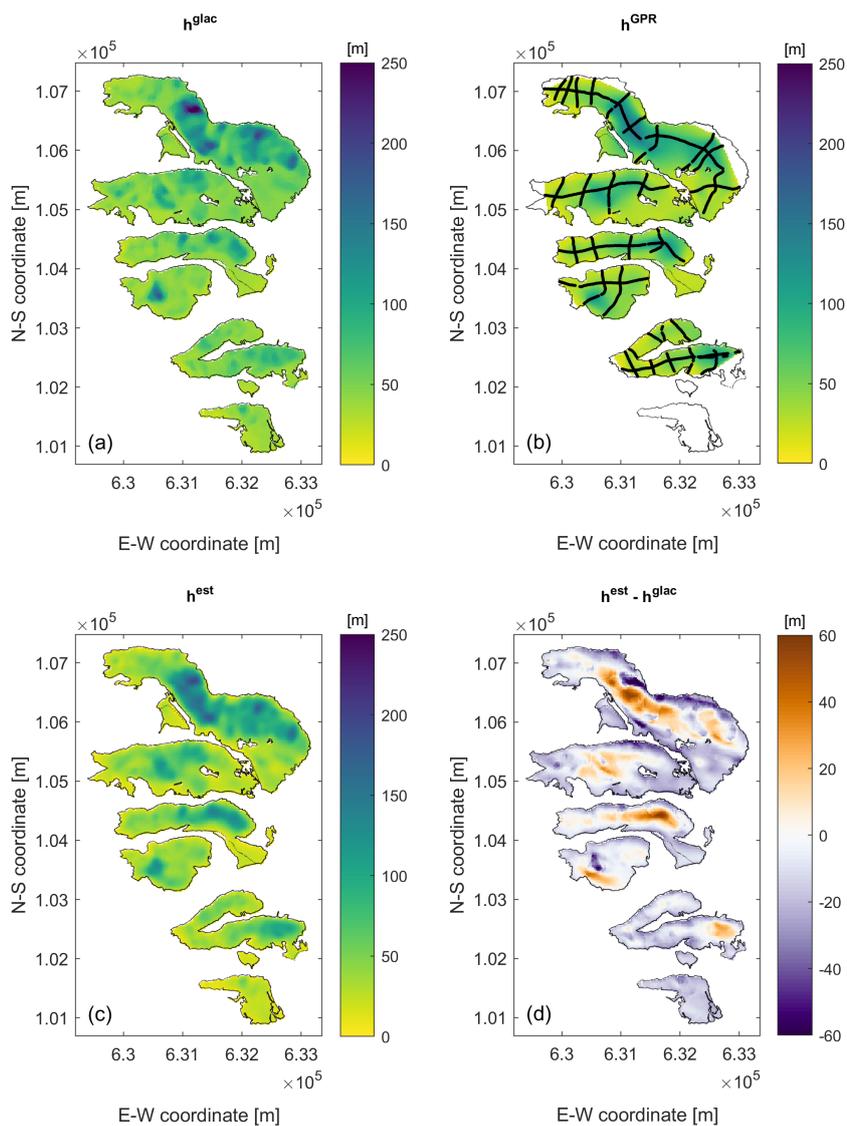
351



352

353 *Figure 5: Results from Plaine Morte Glacier. (a) only glaciological constraints, (b) only*
354 *GPR constraints (available thickness data from GPR profiles marked with black lines),*
355 *(c) GlaTE inversion, (d) difference between GlaTE inversion and glaciological model.*

356



357
 358 *Figure 6: Results from the Dom region. (a) only glaciological constraints, (b) only GPR*
 359 *constraints (available thickness data from GPR profiles marked with black lines), (c)*
 360 *GlaTE inversion, (d) difference between GlaTE inversion and glaciological model.*

361
 362
 363



364 **3 Optimized experimental design using GlaTE inversion**

365

366 All the investigations, described in Section 2, were based on existing GPR data. Their
 367 experimental layouts were designed heuristically using experience from prior surveys.
 368 Once a glacier model has been established, one may realize that another GPR survey
 369 layout may have provided better information. Therefore, a dense survey grid, as
 370 employed for 3D seismic reflection campaigns for hydrocarbon exploration for example
 371 (e.g., Vermeer, 2003) would be the best choice. This, however, would exceed by far the
 372 budgets typically available for glacier investigations.

373

374 Optimizing the glaciological constraints with only a limited number of GPR data is a
 375 chicken-and-egg problem: identifying the most useful GPR data to be added would
 376 require knowledge on where the true ice thickness distribution deviates most from the
 377 distribution in the glaciological model, but this would require advanced prior knowledge
 378 about the ice thickness that one wants to measure. The problem can be tackled
 379 nevertheless by making some specific assumptions (see below).

380

381 With our investigations, we address the following questions.

- 382 1. Was the experimental geometry and the amount of data acquired in the three
 383 investigation areas adequate?
- 384 2. Do better experimental layouts exist for constraining the ice thicknesses in a cost-
 385 optimized manner?
- 386 3. Can some general recommendations be made for designing helicopter-borne GPR
 387 surveys on glaciers?

388

389 Due to the lack of knowledge on the true ice thicknesses, we assumed that the GlaTE
 390 inversion results, shown in Figures 3, 5 and 6 are a good proxy for the actual thickness
 391 distributions. Without GPR data, the state of knowledge is represented by the
 392 glaciological model (Figures 2a, 5a and 6a). For these models, only 16% (Mortersatsch),
 393 10% (Plaine Morte) and 23% (Dom) of the GPR data constraints satisfy the condition
 394 $\|\mathbf{G}\mathbf{h}^{\text{glac}} - \mathbf{h}^{\text{GPR}}\| < \epsilon^{\text{GPR}}$, and the average ice thickness misfits over the entire glacier area

395 $\left(\text{mean}\left(\mathbf{h}^{\text{glac}} - \mathbf{h}^{\text{true}}\right)\right)$ (\mathbf{h}^{true} = “true” model) are 20 m, 25 m and 15 m for the three data

396 sets, respectively. It should be noted that the glaciological models are calibrated with
 397 α_{GPR} . If no GPR data would have been available, the performance of the glaciological
 398 models would be even worse.

399

400 Subsequently, it is analyzed which of the profiles j ($j = 1 \dots \text{number of profiles}$) causes the
 401 largest discrepancies between \mathbf{h}^{GPR} and \mathbf{h}^{glac} . For that purpose we define

402 (12)
$$d_1^{\text{cost}} = \max_j \left(\frac{\sum_{i=1}^{i=n_j} P\left(|h_{ij}^{\text{GPR}} - h_{ij}^{\text{glac}}|\right)}{c_j} \right),$$



403

404 where index i runs over all n_j data points of profile j . h_{ij}^{GPR} and h_{ij}^{glac} represent the
 405 measured and modelled ice thickness at data point i of profile j . The function P is defined
 406 as
 407

$$408 \quad (13) \quad P(x) := \begin{cases} 1 & \text{if } x > \varepsilon^{GPR} \\ 0 & \text{if } x \leq \varepsilon^{GPR} \end{cases} .$$

409

410 Since longer profiles would be associated with higher (monetary) data acquisition costs,
 411 the discrepancy d_1^{cost} is normalized with a cost factor c_j , defined as

412

$$413 \quad (14) \quad c_j = \max(len_j, 200) ,$$

414

415 where len_j represents the length of profile j . This cost function assumes that the
 416 acquisition costs increase linearly with profile length, which is realistic, because the
 417 helicopter costs are typically charged per minute of flight time. To avoid that overly short
 418 profiles would dominate d_1^{cost} , the assumption was made that profiles with $len < 200$ m
 419 would incur the same costs (for such short profiles the flight time is typically governed
 420 by positioning the helicopter at the starting point of a profile).
 421

422 The profile associated with the largest discrepancy d_1^{cost} is expected to offer the largest
 423 amount of additional information per unit cost. In this virtual experiment, we assumed
 424 that one would acquire this profile and subsequently perform a GlaTE inversion, yielding
 425 an improved model \mathbf{h}^{est_k} . Index k indicates the actual state of the experimental design,
 426 that is, k is equal to 1, when adding the first profile. Then, the next profile line to be
 427 acquired is identified using
 428

$$429 \quad (15) \quad d_{k+1}^{cost} = \max_j \left(\frac{\sum_{i=1}^{i=n_j} P\left(\left|h_{ij}^{GPR} - h_{ij}^{est_k}\right|\right)}{c_j} \right)$$

430

431 Repeated application of Equation (15) identifies an optimized sequence for how the
 432 profiles should be acquired. Figures 7a, 7c and 7e show the evolution of what we call the
 433 “data fit curve”, i.e. the evolution of
 434

$$435 \quad (16) \quad d_{k+1}^{fit} = \max_j \left(\frac{\sum_{i=1}^{i=n_j} \hat{P}\left(\left|h_{ij}^{GPR} - h_{ij}^{est_k}\right|\right)}{n_j} \right)$$



436

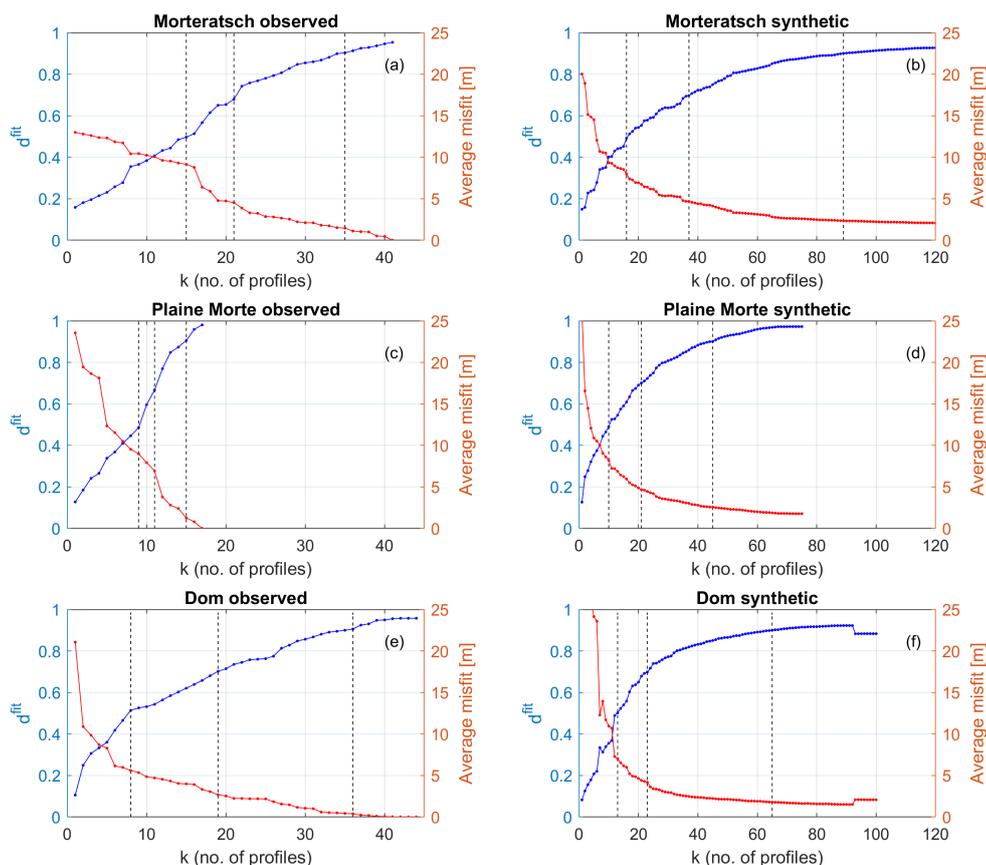
437 with

438

439 (17)
$$\hat{P}(x) := \begin{cases} 0 & \text{if } x > \varepsilon^{GPR} \\ 1 & \text{if } x \leq \varepsilon^{GPR} \end{cases} .$$

440

441 For the Morteratsch and Plaine Morte data, there is an approximately linear increase of
442 the data fit curve. Likewise, we observe a corresponding linear decrease of the average
443 model misfit. As discussed in Maurer et al. (2010), benefit-cost curves, such as the d^{fit}
444 graphs in Figure 7, typically enter into the area of diminishing returns at some stage, that
445 is, the curves exhibit a characteristic kink and flatten out at larger numbers of profiles.
446 This indicates that it becomes increasingly expensive to obtain additional information.
447 The curves in Figures 7a and 7c therefore indicate that the area of diminishing returns
448 was not reached during the Morteratsch and Plain Morte campaigns, and that it would
449 have been useful to acquire more profiles. In contrast, the d^{fit} and average misfit curves
450 for the Dom region (Figure 7e) start flattening out, although we do not observe a
451 characteristic kink in the curves. This indicates that it would have been very expensive to
452 obtain a more accurate ice thickness distribution for the Dom field site.
453



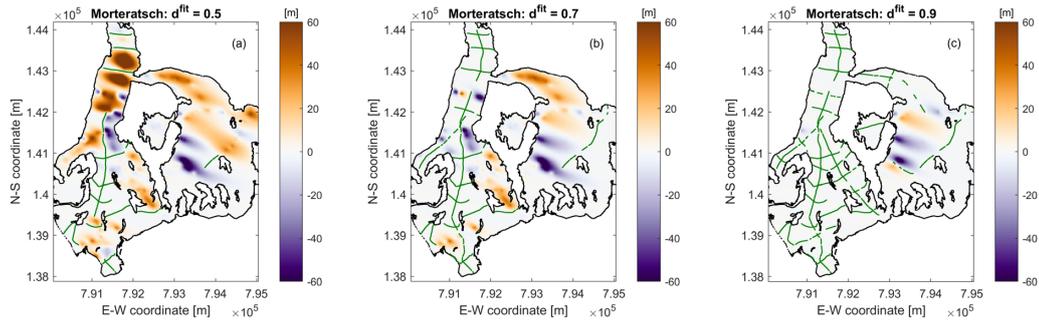
454
 455

Figure 7: Evolution of data fit d^{fit} (blue curves) and average data misfit
 $\text{mean}(\mathbf{h}^{\text{est}_k} - \mathbf{h}^{\text{true}})$ (red curves). Panels a), c) and e) show the results for the observed
 data, and panels b), d) and f) show the results for the synthetic data generated on a
 densely spaced grid of hypothetical profiles. Vertical dashed lines indicate the number of
 profiles required to achieve d^{fit} values of 0.5, 0.7 and 0.9 (see also Figures 8 to 13).

460

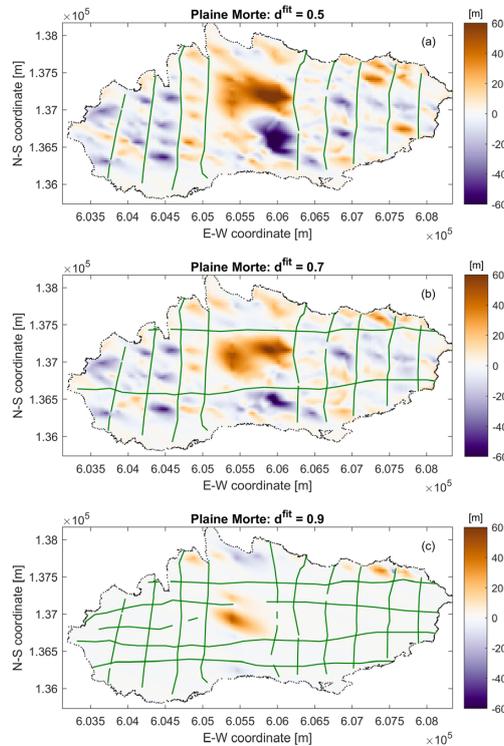
461 Figures 8 to 10 show examples of model misfit plots ($\mathbf{h}^{\text{est}_k} - \mathbf{h}^{\text{true}}$) superimposed with the
 462 selected profile lines. The corresponding stages of the selection procedure are indicated
 463 with black dashed lines in Figures 7a, 7c and 7e. For the Morteratschgletscher, profiles
 464 are selected preferentially in the western part, because the model fit is already quite good
 465 in the eastern region. For Plaine Morte, it is interesting to note that most N-S profiles are
 466 selected before the longer and thus more expensive E-W oriented profiles are considered.
 467 In the Dom region, no obvious selection patterns can be recognized.

468
 469



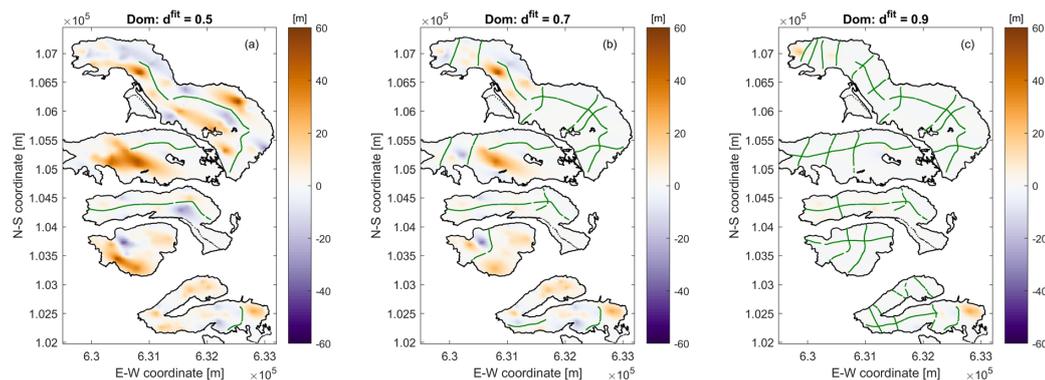
470
 471
 472
 473

Figure 8: Morteratsch model misfit $h^{true} - h^{est_k}$ after selected stages of the experimental design procedure using observed data (see also vertical dashed lines in Figure 7). The selected GPR profiles are superimposed with green lines.



474
 475
 476
 477

Figure 9: Plaine Morte model misfit $h^{true} - h^{est_k}$ after selected stages of the experimental design procedure using observed data (see also vertical dashed lines in Figure 7). The selected GPR profiles are superimposed with green lines.



478
479
480
481

Figure 10: Dom model misfit $h^{\text{true}} - h^{\text{est}_k}$ after selected stages of the experimental design procedure using observed data (see also vertical dashed lines in Figure 7). The selected GPR profiles are superimposed with green lines.

482
483
484
485
486
487
488
489
490
491
492

A major limitation of this design experiment is that the “true” model and the recorded GPR profiles have a strong dependency. When all profiles of a particular region are selected, there is a perfect match between h^{est_k} and h^{true} . However, this is the result of our choice of the “true” model, and thus not indicate that this data set is optimal. To reduce, at least partially, this dependency, we have generated synthetic data sets that are covering all glacierized areas with a dense grid. We assumed a line spacing of 100 m and an inline sampling interval of 0.5 m, which is representative for the helicopter-borne GPR data that we acquired. With such a comprehensive data set, the experimental design procedure should have more flexibility to choose cost-optimized suites of profiles.

493
494
495
496
497
498
499
500
501
502
503
504

The resulting benefit-cost curves are shown in Figures 7b, 7d and 7f. As expected, the curves start flattening out after selecting a sufficiently large number of profiles. For the Morteratschgletscher (Figure 7b), it seems to be worthwhile acquiring more than the 43 profiles acquired during the actual experiment. After about 70 profiles, there is no significant benefit observed. Likewise, the curves for the Glacier Plaine Morte (Figure 7d) indicate clearly that acquiring a larger number of profiles would have been beneficial. After adding about 40 profiles, the d^{fit} curve starts flattening out. Only for the Dom region, the amount of profiles chosen for the actual survey seems to be adequate (Figure 7f). Note that the decrease in d^{fit} at about $k = 8$ and $k = 90$ in Figure 7f are the result of the applied smoothing constraints interfering with the data fit, but this does not affect the general shape of the curve.

505
506

Using the d^{fit} curves in Figure 7 seems to be a good option for selecting an appropriate number of profiles, but it is also insightful to consider the associated model misfit curves.



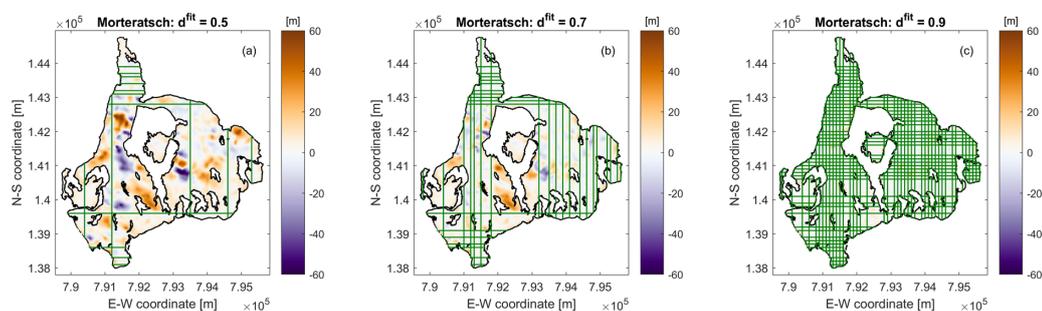
507 Figures 7b, 7d and 7f indicate that the average misfit $\varepsilon^{GPR} = 5m$ is typically reached well
508 before the \mathbf{d}^{fit} curves start flattening out.

509

510 For the experimental design with the synthetic data, Figures 11 to 13 shows examples of
511 model misfit plots ($\mathbf{h}^{est_k} - \mathbf{h}^{true}$) superimposed with the selected profile lines. In contrast
512 to the selection based on observed data from the Morteratschgletscher (Figure 8), the
513 design based on the dense synthetic grid (Figure 11) yields a better balance of profiles
514 among the eastern and western portions of the glacier. This is the consequence of the
515 larger flexibility of choosing profiles with the dense grid. For the Glacier Plaine Morte
516 (Figure 12), it is interesting to note that exclusively N-S oriented profiles were chosen. In
517 contrast, predominantly E-W oriented profiles were chosen for the Dom region (Figure
518 13). Both observations are governed primarily by the cost factor c_j in Equation (15).

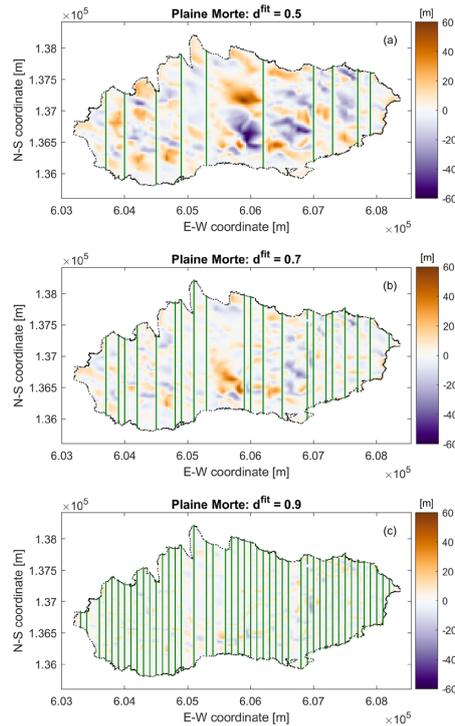
519

520



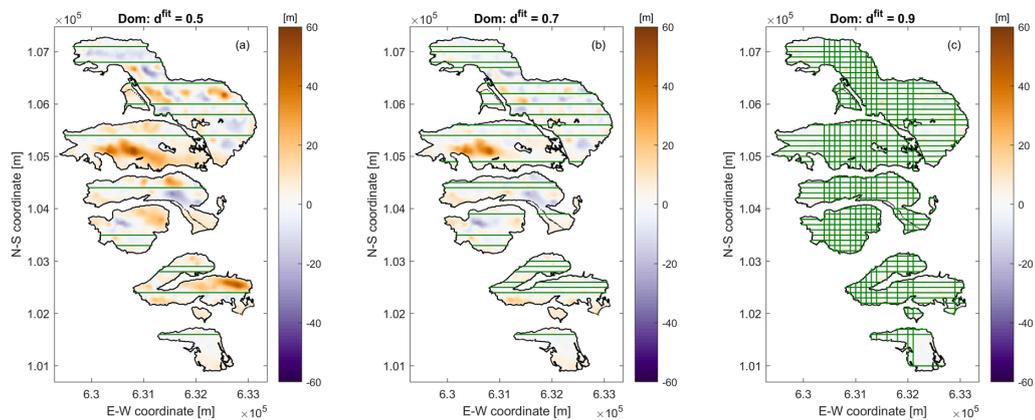
521

522 *Figure 11: Morteratschgletscher model misfit $\mathbf{h}^{true} - \mathbf{h}^{est_k}$ after selected stages of the*
523 *experimental design procedure using synthetic data (see also vertical dashed lines in*
524 *Figure 7). The selected GPR profiles are superimposed with green lines.*



525
 526
 527
 528

Figure 12: Glacier Plaine Morte model misfit $h^{\text{true}} - h^{\text{est}_k}$ after selected stages of the experimental design procedure using synthetic data (see also vertical dashed lines in Figure 7). The selected GPR profiles are superimposed with green lines.



529
 530
 531
 532

Figure 13: Dom Region model misfit $h^{\text{true}} - h^{\text{est}_k}$ after selected stages of the experimental design procedure using synthetic data (see also vertical dashed lines in Figure 7). The selected GPR profiles are superimposed with green lines.



533

534

535 **4 Discussion and conclusions**

536

537 The GlaTE inversion scheme presented in this paper offers numerous beneficial features.
538 Its main advantage is its versatility, as there are several parameters, by which the
539 algorithm can be tuned to the peculiarities of a particular investigation area. However,
540 this is also one of the method's major drawbacks, since the choice of the control
541 parameters may include a considerable amount of subjectivity. This applies primarily to
542 the choice of the weighting factors λ_1 , λ_2 and λ_4 . Finding an appropriate value for λ_4
543 can be particularly awkward, since there is typically no ground-truth information
544 available on the lateral smoothness of the ice thickness distribution. Therefore, we have
545 chosen to keep λ_1 and λ_2 fixed and to determine λ_4 automatically. Quantifying our
546 (relative) confidence in the GPR constraints (λ_1) and glaciological constraints (λ_2) is
547 also a non-trivial task. For this problem, however, some physical arguments may exist.
548 Nevertheless, it might be helpful to repeat the GlaTE inversions with a range of λ_1 / λ_2
549 ratios and to check the corresponding variations in the resulting models.

550

551 Another potential problem is the determination of the scaling factor α_{GPR} in Equation (7).
552 It is largely dependent on the available GPR data, and it is assumed that the GPR profiles
553 have a good areal coverage, which might not be always the case. If values for α_{GPR}
554 would be available for a large number of glaciers, a statistical analysis might be used to
555 correlate the values with specific features of the glaciers (e.g., average steepness,
556 elevation above sea level, size or shape of the glacier, exposure, etc.). This may be
557 helpful in areas, where the GPR data coverage is poor or even non-existent.

558

559 In principle, any observations (e.g., boreholes) can be employed as data constraints in
560 Equation (1), but GPR measurements are typically the main source of information.
561 Migration of the GPR data allows the bedrock reflections to be imaged at the correct
562 positions and slopes along a profile, but it is possible that the reflections originated from
563 locations away from the profile lines (off-plane reflections). This may cause systematic
564 errors affecting the reliability of the results. We note, however, that this is not a problem
565 specific to GlaTE, but rather a general issue affecting GPR data acquired on a sparse grid.

566

567 As mentioned in Section 2, the system of equations in (11) can be augmented by any
568 linear constraints. An obvious, and in our view particularly useful set of constraints
569 would be offered by surface displacement measurements. They can be obtained from
570 differential satellite images and offer full coverage over a glacier. Such constraints could
571 possibly substitute the smoothness constraints in Equation (11) with a physically more
572 meaningful quantity.

573

574 Despite the limitations of our approach, we judge that our results provided useful insights
575 for designing GPR experiments, and some answers to the questions posed in Section 3
576 can be provided.



577

578 1. *Was the experimental geometry and the amount of data acquired in the three*
579 *investigation areas adequate?*

580

581 The benefit-cost curves in Figure 7 indicate that, at least for the Morteratsch and
582 Glacier Plaine Morte, it would have been useful to acquire more data.

583

584 2. *Do better experimental layouts exist for constraining the ice thicknesses in a cost-*
585 *optimized manner?*

586

587 The experimental layouts in Figures 8 to 13 do not provide unexpected features,
588 but indicate that acquiring a larger number of shorter profiles, instead of recording
589 a few long ones, could be beneficial, but it should be noted that we do not take
590 into account the flight time required to move to the next profiles. This could be
591 significant on glaciers with steep mountain flanks.

592

593 3. *Can some general recommendations for designing helicopter-borne GPR surveys*
594 *on glaciers be made?*

595

596 Based on our results, it is difficult to offer general recommendations. For
597 estimating the overall amount of data to be collected, the benefit-cost curves are
598 most indicative. However, in our case studies they do not flatten out clearly,
599 thereby indicating that it would be worthwhile acquiring more data. When high-
600 precision ice thickness maps are required, it is therefore advisable to acquire as
601 much data as can be afforded.

602

603 It is common practice to acquire crossing profiles, but from the experimental
604 layouts, shown in Figure 12, it could be concluded that it is not necessary to
605 acquire a large amount of crossing profiles. From a practical point of view, this
606 recommendation cannot be fully supported. When the signal-to-noise ratio of the
607 GPR profiles is poor, it can be difficult to identify the bedrock reflections
608 unambiguously.

609

610 It is not realistic to adopt a real-time experimental design strategy (i.e., choosing
611 the next profile based on the results of the previously acquired data), as assumed
612 in our virtual experiments in Section 3. However, if logistically feasible, it might
613 be useful to employ a two-step acquisition strategy. Initially, only a few profiles
614 could be acquired. After analyzing these data sets, regions, where large
615 discrepancies between h^{est} and h^{glac} exist, could be identified, and a suitable set
616 of additional profiles could be acquired with a second campaign.

617

618

619



620 **Acknowledgments**

621

622 We thank Patrick Lathion, Philipp Schaer and Kevin Déléze from GEOSAT SA, Patrick
623 Fauchère from Air Glacier, Hansueli Bärffuss from Heli-Bernina, as well as Lasse
624 Rabenstein and Lino Schmid for acquiring the data. Furthermore, we thank Matthias
625 Huss for fruitful discussions and Daniel Farinotti for an insightful in-house review, which
626 improved the clarity of the manuscript. Financial support was provided by ETH Zurich
627 (Grant ETH-15 13-2), the Innosuisse program SCCER-SOE (Swiss competence center
628 for energy research, supply of electricity), and the Swiss Geophysical Commission and
629 ETH Zurich.

630

631

632 **References**

633

634 Clarke, G. K., Anslow, F. S., Jarosch, A. H., Radić, V., Menounos, B., Bolch, T., and
635 Berthier, E.: Ice volume and subglacial topography for western Canadian glaciers from
636 mass balance fields, thinning rates, and a bed stress model, *J Climate*, 26, 4282-4303,
637 2013.

638 Constable, S. C., Parker, R. L., and Constable, C. G.: Occam's inversion: A practical
639 algorithm for generating smooth models from electromagnetic sounding data,
640 *Geophysics*, 52, 289-300, 1987.

641 Evans, S.: Radio techniques for the measurement of ice thickness, *Polar Record*, 11, 406-
642 410, 1963.

643 Farinotti, D., Huss, M., Bauder, A., Funk, M., and Truffer, M.: A method to estimate the
644 ice volume and ice-thickness distribution of alpine glaciers, *J Glaciol*, 55, 422-430, 2009.
645 Farinotti, D., Brinkerhoff, D. J., Clarke, G. K., Fürst, J. J., Frey, H., Gantayat, P., Gillet-
646 Chautet, F., Girard, C., Huss, M., and Leclercq, P. W.: How accurate are estimates of
647 glacier ice thickness? Results from ITMIX, the Ice Thickness Models Intercomparison
648 eXperiment, *Cryosphere*, 11, 949-970, 2017.

649 Grab, M., Bauder, A., Ammann, F., Langhammer, L., Hellmann, S., Church, G., Schmid,
650 L., Rabenstein, L., and Maurer, H.: Ice volume estimates of Swiss glaciers using
651 helicopter-borne GPR—an example from the Glacier de la Plaine Morte, 2018 17th
652 International Conference on Ground Penetrating Radar (GPR), 2018, 1-4,

653 Huss, M., and Farinotti, D.: Distributed ice thickness and volume of all glaciers around
654 the globe, *Journal of Geophysical Research: Earth Surface*, 117, 2012.

655 Huss, M., Voinesco, A., and Hoelzle, M.: Implications of climate change on Glacier de la
656 Plaine Morte, Switzerland, *Geographica Helvetica*, 68, 227-237, 2013.

657 Iken, A.: Adaption of the hot-water-drilling method for drilling to great depth,
658 *Mitteilungen der Versuchsanstalt für Wasserbau, Hydrologie und Glaziologie an der*
659 *Eidgenössischen Technischen Hochschule Zurich*, 211-229, 1988.

660 Kamb, B., and Echelmeyer, K. A.: Stress-gradient coupling in glacier flow: I.
661 Longitudinal averaging of the influence of ice thickness and surface slope, *J Glaciol*, 32,
662 267-284, 1986.

663 Langhammer, L., Rabenstein, L., Schmid, L., Bauder, A., Grab, M., Schaer, P., and
664 Maurer, H.: Glacier bed surveying with helicopter-borne dual-polarization ground-
665 penetrating radar, *J Glaciol*, 1-13, 10.1017/jog.2018.99, 2018.



- 666 Linsbauer, A., Paul, F., and Haerberli, W.: Modeling glacier thickness distribution and bed
667 topography over entire mountain ranges with GlabTop: Application of a fast and robust
668 approach, *Journal of Geophysical Research: Earth Surface*, 117, 2012.
- 669 Maurer, H., Curtis, A., and Boerner, D. E.: Recent advances in optimized geophysical
670 survey design, *Geophysics*, 75, 75A177 – 175A194, 2010.
- 671 Maurer, H., Nuber, A., Martiartu, N. K., Reiser, F., Boehm, C., Manukyan, E.,
672 Schmelzbach, C., and Fichtner, A.: Optimized Experimental Design in the Context of
673 Seismic Full Waveform Inversion and Seismic Waveform Imaging, in: *Advances in*
674 *Geophysics*, Elsevier, 1-45, 2017.
- 675 Menke, W.: *Geophysical data analysis : discrete inverse theory*, Matlab ed., Academic
676 Press, Waltham, MA, 293 p. pp., 2012.
- 677 Morlighem, M., Rignot, E., Seroussi, H., Larour, E., Ben Dhia, H., and Aubry, D.: A
678 mass conservation approach for mapping glacier ice thickness, *Geophys Res Lett*, 38,
679 2011.
- 680 Morlighem, M., Rignot, E., Mouginot, J., Seroussi, H., and Larour, E.: High-resolution
681 ice-thickness mapping in South Greenland, *Annals of Glaciology*, 55, 64-70, 2014.
- 682 Nye, J.: A method of calculating the thicknesses of the ice-sheets, *Nature*, 169, 529-530,
683 1952.
- 684 Paige, C. C., and Saunders, M. A.: Lsq - an Algorithm for Sparse Linear-Equations and
685 Sparse Least-Squares, *Acm T Math Software*, 8, 43-71, 1982.
- 686 Rutishauser, A., Maurer, H., and Bauder, A.: Helicopter-borne ground-penetrating radar
687 investigations on temperate alpine glaciers: A comparison of different systems and their
688 abilities for bedrock mapping, *GEOPHYSICS*, 81, WA119-WA129,
689 doi:10.1190/geo2015-0144.1, 2016.
- 690 Steinhage, D., Nixdorf, U., Meyer, U., and Miller, H.: New maps of the ice thickness and
691 subglacial topography in Dronning Maud Land, Antarctica, determined by means of
692 airborne radio-echo sounding, *Annals of Glaciology*, 29, 267-272, 1999.
- 693 Vermeer, G. J.: 3d seismic survey design optimization, *The Leading Edge*, 22, 934-941,
694 2003.
- 695 Watts, R. D., and England, A. W.: Radio-echo sounding of temperate glaciers: ice
696 properties and sounder design criteria, *J Glaciol*, 17, 39-48, 1976.
- 697 Zekollari, H., Huybrechts, P., Fürst, J., Rybak, O., and Eisen, O.: Calibration of a higher-
698 order 3-D ice-flow model of the Morteratsch glacier complex, Engadin, Switzerland,
699 *Annals of Glaciology*, 54, 343-351, 2013.
- 700
- 701