



- 1 The long-term GIA signal at present-day in Scandinavia, northern Europe and the British Isles
- 2 estimated from GPS and GRACE data
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10 Abstract

- 11 The long-term glacial isostatic adjustment (GIA) signal at present-day is constrained via joint inversion
- 12 of GPS vertical land motion rates and GRACE gravity data for a region encompassing Scandinavia,
- 13 northern Europe and the British Isles, and the Barents Sea. The best-fit model for the vertical motion
- 14 signal has a χ^2 value of approximately 1 and a maximum posterior uncertainty of 0.3-0.4 mm/yr. An
- 15 elastic correction is applied to the vertical land motion rates that accounts for present-day changes to
- 16 terrestrial hydrology as well as recent mass changes of ice sheets and glaciered regions. Throughout
- 17 the study area, mass losses from Greenland dominate the elastic vertical signal and combine to give
- 18 an elastic correction of up to +0.5 mm/yr in central Scandinavia. Neglecting to use an elastic correction
- 19 may thus introduce a small but persistent bias in model predictions of GIA vertical motion even in
- 20 central Scandinavia where vertical motion is dominated by long-term GIA. The predicted gravity signal
- 21 is generally less well-constrained than the vertical signal, in part due to uncertainties associated with
- 22 the correction for contemporary ice mass loss in Svalbard and the Russian Arctic. The GRACE-
- 23 derived gravity trend is corrected for present-day ice mass loss using estimates derived from the
- 24 ICESat and CryoSat missions, although a difference in magnitude between GRACE-inferred and
- altimetry-inferred regional mass loss rates suggests the possibility of a non-negligible GIA response
- 26 here either from millennial-scale or Little Ice Age GIA.
- 27





28 1. Introduction

29 Long-term glacial isostatic adjustment (GIA) is the process by which the Earth's solid surface and 30 underlying mantle deform in response to loading by the large ice sheets that existed during the last 31 glaciation. Because the time-scale of Earth's viscoelastic relaxation is up to several thousand years, 32 ongoing GIA is usually the dominant present-day deformation signal in formerly glaciated areas that 33 are tectonically guiescent (for example, up to 1 cm/yr land uplift around the northwestern Gulf of 34 Bothnia). Outside formerly glaciated regions, the GIA signal often remains large enough to form a 35 significant component of observed present-day deformation and sea-level change rates. Constraint of 36 the long-term GIA signal at present-day is therefore required for accurate separation of the paleo and the more recent contributions to present-day land deformation and gravity change. This problem is 37 38 complicated further by the fact that the GIA signal itself is temporally and spatially complex, and poorly 39 constrained by models designed to describe both ice cover during the last glaciation and the structure of the Earth. 40

41

42 This paper constrains the long-term GIA signal in Scandinavia and northern Europe through the 43 simultaneous inversion of vertical land motion rates from GPS and gravity change rates from GRACE. 44 Corresponding uncertainties are also empirically estimated for the preferred model(s). Forward GIA 45 model predictions typically have no formal uncertainty estimation although parameter variation 46 suggests that forward model uncertainty is large. The estimation of formal model uncertainty is 47 therefore a notable advantage of semi-empirical or data-driven methodologies. Similar empirical and 48 semi-empirical approaches have been implemented to estimate regional long-term GIA signals in 49 Antarctica (Riva et al. 2009, Gunter et al. 2014), North America (Sasgen et al. 2012, Simon et al. 50 2017), Alaska (Jin et al. 2016) and Fennoscandia (Hill et al. 2010, Zhao et al. 2012). Here, our 51 methodology is based on that of Hill et al. (2010); relative to their previous work, we update both the 52 GPS and GRACE datasets, expand the study area to include northern Europe and the British Isles to 53 the south and the Barents Sea to the north, and incorporate a second model of ice sheet history into 54 the a priori input. There are three main goals: i) to model the paleo GIA signal at present-day in a 55 continuous region between Scandinavia and the British Isles, ii) to estimate empirically the uncertainty





- of the modelled signal, and iii) to assess the importance of applying an elastic correction to the vertical
- 57 land motion data.
- 58

59 2. Model Inputs and Method

- 60 2.1 GPS Data
- 61 Rates of vertical land motion measured by GPS are taken from both Kierulf et al. (2014) and the
- 62 Nevada Geodetic Laboratory (Blewitt et al. 2016) (Figure 1). The Kierulf et al. (2014) dataset has
- 63 relatively dense coverage within the region of the former load centre of the Fennoscandian Ice Sheet
- 64 (FIS), particularly in Norway, but sparse coverage elsewhere. The data from Blewitt et al. (2016) are
- thus used in the region outside the former ice sheet margin. In total, there are 459 stations. The data
- span the years 1996-2016; the time series length varies station to station from 3-20 years, with an
- 67 average time series length of approximately 10 years.



68

Figure 1. Rates of vertical land motion (mm/yr) for the GPS data used in the inversion, after correction
 for elastic effects (Section 2.3). BS – Baltic Sea, FJL – Franz Josef Land, GB – Gulf of Bothnia, NZ –
 Novaya Zemlya, Sv – Svalbard, FJL and NZ = Russian Arctic. Dark red dashed line shows the
 approximate boundary of ice cover at the Last Glacial Maximum (LGM). The size of the circles is
 inversely proportional to the measurement uncertainty.

- 75 As further described in Kierulf et al. (2014), their rates were derived using the GAMIT/GLOBK GPS
- 76 analysis software (Herring et al. 2011), while the data from the Nevada Geodetic Laboratory were
- 77 calculated using the MIDAS trend estimator, an algorithm designed for automatic step detection in





- 78 time series (Blewitt et al. 2016). Although the processing technique differs for each dataset, the two
- 79 datasets are combined in order to achieve the best possible spatial coverage in the study area.
- 80 Common sites in the two datasets compare within the observational uncertainties at all but one of
- 81 thirty-one sites, and no apparent bias is observed between the differences at the shared sites.
- 82 Because the uncertainties are consistently larger for the data from the Nevada Geodetic Laboratory
- than for the data from Kierulf et al. (2014), we use the common sites to determine an average
- 84 uncertainty scaling factor (~2.25) to apply to the uncertainties in the latter dataset. The scaling avoids
- significantly biasing the inversion result towards fitting either dataset. Both datasets are aligned in the
- 86 International Terrestrial Reference Frame 2008 (Altamimi et al. 2011), which is consistent with the CM
- 87 frame to within ~0.2 mm/yr. As described in Section 2.3, an elastic correction is applied that accounts
- 88 for recent changes in ice sheet and glacier volumes and terrestrial hydrology.

89

90 2.2 GRACE

91 The GRACE data are processed as in Simon et al. (2017). Rates of gravity change for a 10.5 year 92 period from 2004.02-2014.06 are estimated using 113 GRACE Release-05 (RL05) monthly solutions 93 from the University of Texas at Austin Center for Space Research (CSR). The coefficients are 94 truncated at degree and order 96 (consistent with a spatial resolution of ~200 km). Values estimated 95 from Satellite Laser Ranging (Cheng et al. 2013) replace the C₂₀ coefficients. Following Klees et al. 96 (2008), the monthly fields are filtered with a statistically optimal Wiener filter. The optimal filter 97 incorporates the full variance-covariance information of the monthly solutions, and less aggressively 98 filters in regions where signal is stronger. A mass trend is estimated that accounts for bias, annual, 99 and semi-annual variations (Figure 2). The signal uncertainty is represented by the full variance-100 covariance matrix of the trend. Corrections for changes in the terrestrial hydrology cycle and ice mass 101 loss from Svalbard and the Russian Arctic are applied as described in Section 2.3.







102EWT (mm/yr)EWT σ (mm/yr)103Figure 2. (a) Total gravity change rates measured from GRACE, (c) correction for terrestrial hydrology104changes and present-day ice mass loss (Section 2.3), and (e) final corrected rates. (b,d,f) Same as105(a,c,e) but rates are the 2σ uncertainties associated with the signal. Units are mm/yr change in106equivalent water thickness (EWT).

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108

109 2.3 Corrections for Terrestrial Hydrology and Present-day Ice Melt

- 110 Changes in terrestrial hydrology as well as present-day ice mass loss from Greenland, and glaciers
- 111 and ice caps in Svalbard, the Russian Arctic, and Scandinavia may form a significant contribution to
- the total measured gravity change and vertical motion rates within the study area.
- 113





114 GRACE

- 115 In the continental region and south of approximately 71.5° N latitude, hydrological changes are
- 116 estimated with the model PCR-GLOBWB (Wada et al. 2014), which includes changes from
- 117 anthropogenic groundwater depletion and dam retention. The trend is computed for 2004-2014 from
- 118 11 annual means on a 2° x 2° grid, consistent with the resolution of the GRACE data (Figure 2c). In
- 119 glaciered regions (Scandinavia, Svalbard and the Russian Arctic), the hydrology model is not used to
- 120 correct the input rates. Rather, it is assumed that present-day estimates of regional ice melt derived
- 121 from altimetry observations should more accurately capture the dominant hydrological signals that
- 122 would be modelled by PCR-GLOBWB.

123

Estimates of present-day mass changes in Scandinavia, the Russian Arctic, and Svalbard are
summarized in **Table 1**, and vary considerably depending on estimation method and time period.
Mass loss in Scandinavia is consistently small and generally estimated to be between -1.2 to -2 Gt/yr.
Here, we apply a mass loss rate of -1.3 Gt/yr, determined by glaciological modelling (Marzeion et al.
2012, 2015).

129

In the Russian Arctic, glaciological estimates of mass change are consistent within uncertainties for the different time periods and suggest mass change between -21.0 to -24.7 Gt/yr. These rates are approximately twice those estimated by the ICESat and CryoSat missions, which estimate mass changes in this region of between -10.5 to -14.9 Gt/yr, with a small acceleration observed after 2010 (Wouters, *pers. comm.*, 2016). The smallest net mass change estimate for the Russian Arctic comes from GRACE, with -5.7 Gt/yr mass change observed between 2003-2013 (Schrama et al. 2014).

136

In Svalbard, estimated mass change rates are more discrepant. Again, glaciological estimates are the largest, but two estimates of -42.0 Gt/yr and -17.0 Gt/yr between 2003-2009 are not consistent within uncertainties and differ in magnitude by more than a factor of 2. Laser and radar altimetry estimates are smaller, and suggest a clear acceleration in mass loss since 2010 (-4.6 Gt/yr between 2003-2009





- 141 and -16.5 Gt/yr between 2010-2014, Wouters, pers. comm., 2016). As with the Russian Arctic,
- 142 GRACE is the estimation technique that records the smallest net mass change, with -4.0 Gt/yr
- 143 estimated in Svalbard between 2003-2013 (Schrama et al. 2014).
- 144

Study/Source	Svalbard (Gt/yr)	Russian Arctic (Gt/yr)	Scandinavia (Gt/yr)
	2003-	-2009	
Marzeion et al. (2012, 2015) <i>(2003-2009)</i>	-42.0 ± 3.2 (gl)	-22.9 ± 4.7 (gl)	-1.2 ± 0.2 (gl)
Gardner et al. (2013) (2003-2009)	-17.0 ± 6.0 (gl) -5.0 ± 2.0 (l, G)	-21.0 ± 13.0 (gl) -11.0 ± 4.0 (l, G)	-2.0 ± 0.0 (gl)
Wouters (2016) (2003-2009)	-4.6 ± 1.2 (I)	-10.5 ± 1.3 (I)	-
	2010-	-2014	
Wouters (2016) <i>(2010-2014)</i>	-16.5 ± 1.6 (C)	-14.9 ± 1.2 (C)	-
	≥10 years t	ime period	
Marzeion et al. (2012, 2015) <i>(2004-2013)</i>	-39.8 ± 2.2 (gl)	-24.7 ± 3.0 (gl)	-1.3 ± 0.1 (gl)
Average Wouters (2016) (2003-2014)	-10.6 ± 2.0 (I, C)	-12.7 ± 1.8 (I, C)	-
Schrama et al. (2014) <i>(2003-2013)</i>	-4.0 ± 0.7 (G)	-5.7 ± 0.9 (G)	+1.3 ± 0.9 (G)
This study	-10.6 ± 2.0 (I, C)	-12.7 ± 1.8 (I, C)	-1.3 ± 0.1 (gl)
This study, with scaling	-2.7 ± 2.0 (I, C)	-2.5 ± 1.8 (I, C)	-1.3 ± 0.1 (gl)*

145 146

Table 1. Estimates of present-day mass change for Svalbard, the Russian Arctic, and Scandinavia for different time periods and from different sources. Letters in parentheses indicate estimation method; gl 147 - glaciological, I - IceSat, G - GRACE, C - CryoSat. All rates are in Gt/yr. *Not scaled.

148

149 The differing mass change estimates among measurement techniques for the Russian Arctic and

150 Svalbard raise the question of which value to use when applying a correction to the total GRACE trend

shown in Figure 2a. GRACE measures total mass changes (solid Earth plus cryosphere), while 151

152 glaciological and altimetry methods more accurately isolate changes to the cryosphere. Relative to





GRACE, the latter two methods both consistently infer larger mass losses, suggesting that GRACE may contain a significant mass gain signal from the solid Earth, either from glacial isostatic adjustment from the last glaciation, or from the Little Ice Age (LIA). For both Svalbard and the Russian Arctic, we choose to apply an estimate that averages the ICESat and CryoSat estimates over the years 2003-2014 (**Table 1**). Subtracting these averaged rates from the total GRACE estimates for a similar time period (2003-2013, Schrama et al. 2014, **Table 1**), infers a reasonably consistent total solid Earth or GIA signal of +6.6-7 Gt/yr in the region.

160

161 However, applying the averaged ice melt corrections to Svalbard and the Russian Arctic creates a 162 large mass gain signal over these two areas and a relatively smaller signal in the central Barents Sea; 163 this pattern is generally inconsistent with ice coverage in the Barents Sea region suggested by several different Pleistocene ice sheet reconstructions (Auriac et al. 2016), and therefore inconsistent with the 164 paleo GIA signal that the input signal should represent. Possible explanations for this inconsistency 165 166 are: i) models of LGM ice cover in the region require thicker ice over Svalbard and the Russian Arctic than in the Barents Sea, ii) there is a large Little Ice Age GIA signal over these two regions, and/or iii) 167 the Wiener filter applied to the GRACE data too aggressively filters signal in these small regions. The 168 169 first explanation is unlikely because glacial margin chronology suggests that Svalbard and the Russian 170 Arctic were located on or near the margin of the Barents Ice Sheet where ice cover would have been 171 thinnest. To counteract the effect of either of the latter two explanations (LIA rebound or signal loss in 172 GRACE), we apply ad-hoc scaling factors of 0.25 and 0.2 to the ice mass loss estimates in Svalbard 173 and the Russian Arctic (Table 1), so that their removal from the total GRACE signal results in a spatial 174 pattern in the residual (i.e., paleo GIA) signal that is approximately consistent with thicker LGM ice 175 cover over the Barents Sea than around its margins (Figure 2e). Such a scaling factor approach is 176 certainly not ideal, but serves to provide a GRACE input signal in the Barents Sea region that has a 177 spatial pattern broadly consistent with expectations of the paleo GIA response to loading and 178 unloading from the Barents Ice Sheet.

179





181 GPS

182	Vertical land motion rates may likewise be affected by present-day ice mass loss and the terrestrial
183	hydrology cycle. As with the GRACE data, the GPS data are corrected for changes to terrestrial
184	hydrology south of 71.5° N latitude using predictions from the PCR-GLOBWB model, although here,
185	the hydrology trend has been estimated from 1993-2014 to be more consistent with the length of the
186	GPS time series. North of 71.5° N latitude, the same scaled corrections derived from ICESat and
187	CryoSat are applied for present-day ice mass changes in Svalbard and the Russian Arctic.
188	Throughout the study area, the GPS measurements are also corrected for additional elastic vertical
189	motion from mass loss of the Greenland Ice Sheet, the Antarctic Ice Sheet and glaciers and ice caps
190	in northern Canada. Mass loss of the Greenland Ice Sheet is estimated from 1993-2014 using surface
191	mass balance estimates from RACMO2.3 (Noël et al. 2015) and ice discharge with a constant
192	acceleration of 6.6 Gt/yr ² (van den Broeke et al. 2016). Mass loss of the Antarctic Ice Sheet is also
193	estimated from 1993-2014 assuming a constant acceleration in ice discharge of 2 Gt/yr ² . The
194	scenarios for both Greenland and Antarctica are consistent with the mass balance estimates from
195	Shepherd et al. (2012). For the Canadian Arctic, a constant mass loss rate of 60 Gt/yr is used
196	(Gardner et al. 2013). All signals combine to yield a total net uplift of approximately 0.2-0.5 mm/yr
197	throughout most of the study area, with Greenland mass loss providing the largest contribution
198	(Figure 3). The additional uncertainties are also computed and added in quadrature to the
199	measurement uncertainties; correction of the GPS data for non-GIA signals adds < ± 0.05 mm/yr
200	uncertainty in most of the study area and $\sim \pm 0.1$ mm/yr in Svalbard (Figure 3).

201

202 Finally, in addition to present-day ice mass loss signals, a correction of 4.33 ± 0.40 mm/yr is removed 203 from the vertical motion rates for the two GPS sites on Svalbard (NYAL and LYRS). This value is an 204 average of 3 scenarios from Mémin et al. (2014) which estimate the vertical land motion at Ny-Ålesund 205 due to Pleistocene and Little Ice Age GIA signals; their estimates range from 3.31-4.95 mm/yr; thus 206 the averaged correction of 4.33 mm/yr that is applied assumes that the signal from Pleistocene GIA is 207 small and that most residual land motion here is from LIA rebound. After correction for present-day ice 208 mass changes and approximated LIA uplift, the residual (inferred paleo GIA) vertical uplift rates at 209 NYAL and LYRS are 2.64 ± 0.80 and 1.10 ± 2.64 mm/yr, respectively.







210 211

Figure 3. GPS-measured rates of vertical land motion before and after the applied elastic correction 212 (top left and right). An elastic correction is computed for mass loss changes from Greenland, the West 213 Antarctic Ice Sheet (WAIS), glaciers and ice caps in northern Canada, Svalbard and the Russian 214 Arctic, and loading from the terrestrial hydrology cycle. Sites on Svalbard are additionally corrected for 215 LIA uplift as discussed in the text.

216

2.4 A Priori Model Information 217

- 218 The prior model covariance matrix contains predictions from a set of forward GIA models that varies
- 219 ice sheet history and mantle viscosity and is constructed as described in Hill et al. (2010) and Simon
- et al. (2017). Here, two different ice sheet histories are coupled to a suite of three-layer Earth models 220
- 221 with an elastic lithosphere and varying upper and lower mantle viscosities.
- 222
- 223 The first ice sheet model is the global ICE-5G model (Peltier 2004). In the second ice sheet model, the
- 224 glacial history over Fennoscandia and the British Isles is described by the model(s) from the Australian
- 225 National University (ANU, Lambeck et al. 2010). This second version of the ice sheet model contains





- 226 ICE-5G coverage over Greenland and Antarctica and the model of North American coverage
- 227 presented in Simon et al. (2015, 2016). Tests indicate that varying the ice sheet history over North
- 228 America has little impact on the predictions in Fennoscandia, although this variation is useful for
- 229 studies that wish to expand the study area outside of the current study area. Relative to ICE-5G, LGM
- 230 ice cover in the ANU model is thinner over the Barents Sea, thicker over Svalbard and Scotland, and
- discontinuous between Scandinavia and the British Isles (Figure 4).

232





Figure 4. Last glacial maximum (LGM) ice cover in Scandinavia, the Barents Sea and the British Isles from ICE-5G (a) and the ANU model (b).

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237
       The Earth models have a 90 km thick elastic lithosphere and the upper and lower mantle viscosities
        span 0.2 - 2 \times 10^{21} Pa s and 1 - 60 \times 10^{21} Pa s, respectively. These viscosities span a range of
238
239
        plausible values in the upper and lower mantle. We note however that both the ICE-5G and ANU ice
240
        sheet models have been fit to a particular viscosity profile. While the coupling of differing Earth models
241
        to a 'tuned' ice sheet history may introduce artificially high variances, this concern may be countered
242
        by considering that the variances in such an a priori Earth-ice model set could almost certainly be
243
        made larger if any combination of 3D Earth structure, non-linear mantle rheology or glaciological and
244
       climatological constraints were additionally incorporated. A full covariance matrix is generated that
245
       relates the variances of each model prediction relative to the suite's average. All models are
246
       represented at spherical harmonic degree and order 256. The average response and uncertainties of
247
       the a priori set is shown in Figure 5.
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 248
 (mm/yr)
 EWT (mm/yr)

 249
 Figure 5. Averaged a priori rates of the Earth-ice model set. (a, c) Vertical rates and uncertainties. (b, d) Gravity change rates and uncertainties in units of equivalent water thickness (EWT) change.

251

252 2.5 Method

253 The least-squares adjustment method is based on the methodology of Hill et al. (2010) and extended

by Simon et al. (2017). The method simultaneously inverts the data constraints (GPS, GRACE or

- both) with the a priori GIA model information and minimizes the misfit to both input types. As in Simon
- 256 et al. (2017), variance component estimation (VCE) is also used to weight the input uncertainties. The
- 257 prior models are combined with the data in three scenarios: inversion with the GPS data alone (D1),
- 258 inversion with the GRACE data alone (D2), and inversion with both datasets (D3).

259

260 3. Results and Discussion

- 261 3.1 Prediction of Vertical Motion and Gravity Change
- 262 Vertical Motion





263	The predicted GIA response and uncertainties for the D1-D3 scenarios are shown for vertical land
264	motion (Figure 6). The incorporation of the GPS data in scenarios D1 and D3 leads to a similar
265	pattern of regional uplift although relative to D1, the D3 scenario predicts slightly lower rates of uplift
266	over the northern British Isles and in the Barents Sea. D1 and D3 have respective peak uplift rates of
267	9.8 and 9.2 mm/yr. When only the gravity data are inverted in the D2 scenario, the region of uplift is
268	broader and the peak uplift rate is smaller at 7.1 mm/yr. In all cases, the peak uplift is centred over the
269	northwestern region of the Gulf of Bothnia. The peak (1 σ) uncertainty rates are ±0.36, ±0.43 and ±0.28
270	mm/yr for the D1-D3 cases. Similar to the results of Simon et al. (2017), the predicted uncertainties
271	are largest where the signal is largest (around the Gulf of Bothnia) and/or the data coverage is
272	sparsest and most poorly constrained (around the Barents Sea). In Finland, for example, the relatively
273	large signal and the relatively sparse data coverage combine to create a region of larger uncertainty
274	than in surrounding areas. The inclusion of VCE does not significantly impact the signal prediction but
275	in general somewhat increases the estimation of posterior model uncertainty; the weighting factors
276	determined by VCE are shown in Table 2 .

277

278 Gravity Change

279 The predicted gravity change rates for D1-D3 are comparable to the predicted vertical motion rates in

both the spatial pattern and relative magnitude (not shown). The peak mass change rates are again

281 centred over the northern Gulf of Bothnia, and are 33.7, 24.3, and 32.3 mm/yr of equivalent water

thickness change for the D1-D3 scenarios. The peak associated 1σ uncertainties are ±1.59, ±1.59 and ±1.22 mm/yr EWT.









287 GIA for the D1-D3 scenarios.

288

289

Data Incorporated	σ^2 Squared Value		I	Ratios	
	σ_1^2 (Vertical)	σ_2^2 (Gravity)	σ_{μ}^2 (Prior)	σ_1^2/σ_2^2	σ_1^2/σ_μ^2 , σ_2^2/σ_μ^2
D1: Vertical only	0.85	-	0.94	-	0.92, -
D2: Gravity only	-	13.51	0.61	-	-, 22.02
D3: Vertical+Gravity	1.02	20.55	0.64	0.05	1.60, 32.29

Table 2. Results of the variance component analysis. σ_1^2 and σ_2^2 are the variance factors applied to the 290

291 vertical motion data (dataset 1) and gravity change data (dataset 2), respectively, and σ_{μ}^2 is the variance factor applied to the prior information. The ratios describe how each input is weighted relative

292 293 to the other(s).





294 3.2 Misfit Values and Residuals

- 295 For both χ^2 and RMS values, the D1 model provides the best fit to the vertical data, the D2 model
- 296 provides the best fit to the gravity data, and the D3 model provides the best fit overall (Figure 7). The
- 297 χ^2 values of the vertical prediction for both D1 and D3 are approximately equal to 1. The χ^2 values for
- the gravity data are relatively large with the smallest value of 15.9 obtained for the D2 model. Scaling
- the gravity data uncertainties by the VCE-determined scaling factors in **Table 2** reduces the overall χ^2
- 300 values for the gravity prediction to approximately 1.2 for the D2 and D3 models. However, the
- 301 statistical fit of the models to the gravity data remains generally worse than the fit to the vertical motion
- 302 data.





Figure 7. Fractional χ^2 and RMS values for each of the D1-D3 models. Fractional values are determined relative to the value of the worst fitting model for both the vertical motion and gravity change predictions (i.e., fractional χ^2 values of the vertical motion prediction are relative to D2 for which $\chi^2 = 2.94$). χ^2 values are not VCE-scaled; see **Figure 8** for all χ^2 values including with and without VCE scaling, where applicable.

309





- Figures 8-9 summarize the spatial residuals for the best-fit D3 model and the binned residuals for all
 models. The vertical motion residuals are unbiased and generally small. Regionally, the D3 model
 underpredicts vertical motion in Scotland and conversely overpredicts vertical motion along parts of
 the southern Norwegian coast and the Netherlands. The gravity residuals for D3 are relatively low for
 much of the study area, although there is noticeable overprediction in central Scandinavia and in the
 Barents Sea.
- 316





317

Figure 8. Spatial residuals for the D3 model for vertical motion (top) and gravity change (bottom). In top panel, triangles indicate model prediction is outside the 1σ uncertainty of the measurement, circles













- **Figure 9.** Histogram of residuals for models D1-D3, for prediction of vertical motion (left) and gravity change (right). Pink and blue shading indicate model overprediction and underprediction, respectively.
- 325 Where given, χ^2 values in brackets show the VCE-scaled χ^2 value.
- 326
- 327
- 328 3.3 Comparison of Vertical Motion Prediction to Other Models
- 329 We compare the vertical motion prediction of D1 to two other models that estimate the long-term GIA
- 330 signal at present-day. The first model is the semi-empirical land uplift model NKG2016LU (Vestøl et al.
- 331 2016) designed by several researchers in collaboration with the Nordic Geodetic Commission (NKG).
- 332 This model is constrained with GPS-measured vertical land motion rates updated from the dataset of
- 333 Kierulf et al. (2014), levelling measurements and GIA model predictions. The second model is the
- 334 forward GIA model ICE-6G (Peltier et al. 2015) which is constrained by a global dataset of vertical
- 335 land motion measurements. The majority of the these data are GPS measurements from the global
- 336 solution of JPL; within the study area of Scandinavia and northern Europe, additional measurements
- 337 come from the BIFROST GPS network as well as a small number of SLR, DORIS and VLBI
- 338 measurements (Argus et al. 2014, Peltier et al. 2015).

339

- 340 Figure 10 compares the vertical land motion predictions of D1, NKG2016LU, and ICE-6G. All
- 341 comparisons are made relative to the vertical motion dataset presented in this paper, although as
- 342 stated above, both NKG2016LU and ICE-6G were constrained with different variants of regional
- 343 vertical land motion data. As well, NKG2016LU predictions are available on a smaller grid centred
- over Scandinavia, thus, we limit our comparison with this model to within these bounds (reducing the
 comparison dataset from 459 to 280 sites).

346

With no significant bias and a χ^2 value of less than 1, the D1 model provides a good fit to the data. As with the D3 model, the D1 model underpredicts vertical motion over the northern British Isles, and appears also to overpredict vertical motion around the Netherlands. The NKG2016LU model has a χ^2 value of less than 1 and a bias towards overprediction of 0.13 mm/yr. The overall bias towards overprediction is small, but is persistent particularly over Scandinavia (**Figure 10**). For the region north





352 of 55°N (so approximately Scandinavia, 185 sites), the bias of the NKG2016LU model increases to 353 0.42 mm/yr. This bias is most likely attributable to the elastic correction applied to our GPS dataset, 354 which is approximately +0.2-0.5 mm/yr over Scandinavia (Figure 3). Without an elastic correction 355 applied to the GPS data, the NKG2016LU model has a bias of only -0.06 mm/yr in the region north of 55°N. The ICE-6G model underpredicts vertical motion at several sites in Scandinavia and has an 356 overall χ^2 value of 1.33, somewhat higher than that of either D1 or NKG2016LU. At station NYAL on 357 358 Svalbard, both the D1 and ICE-6G models underpredict vertical motion by more than 2 mm/yr, even 359 after the applied corrections for present-day mass loss and possible LIA uplift. That the statistical fit to the data of both D1 and NKG2016LU is slightly better than the fit of the ICE-6G forward model is 360 361 expected due to the fundamental difference in model type: unlike ICE-6G, both of the semi-empirical 362 models explicitly incorporate the data into the prediction via formal inversion. Conversely, an 363 advantage of ICE-6G and other models of its type is the direct insight they offer into the space-time 364 evolution of the ice sheets, which cannot be inferred from a present-day empirical prediction alone.







365

Figure 10. Spatial (left) and binned (right) vertical motion residuals for the D1, NKG2016LU and ICE 6G models (the latter two are abbreviated 'NKG' and '6G' in right-hand plots). Triangles indicate model
 prediction is outside the 1σ uncertainty of the measurement, circles indicate model prediction is inside
 the 1σ uncertainty of the measurement.

- 371 3.4 Tide Gauge Comparison
- 372 To assess the effect of GIA on regional sea-level change, we remove model D1's predictions of long-
- 373 term GIA from mean sea-level trends at 17 tide gauge sites along the coast of the North Sea (Figure
- 11). The mean trends are taken from Frederikse et al. (2016a) who developed a state-space model to





- 375 compute time-varying trends in tide gauge records, thereby taking into account unexplained (multi-)
- 376 decadal variability. The rates shown here are averaged time-varying trends from Model C of
- 377 Frederikse et al. (2016a), which removes decadal variability from the tide gauge time series using a
- 378 hydrodynamic model developed to predict storm surge heights along the North Sea coast.
- 379

380	When corrected for the long-term GIA trends, which are assumed to be linear over decadal time-
381	scales, the standard deviation of the trends decreases somewhat from 1.08 mm/yr to 0.89 mm/yr. The
382	GIA correction is small at most sites, and at all sites except 10 and 11 (Hirtshals and Tregde), the
383	averaged sea-level trends appear dominated by processes other than long-term GIA (Figure 11). At
384	Hirtshals and Tregde, which are located nearest to the centre of the former FIS, the predicted GIA-
385	induced sea-level trend is more than twice the magnitude of the averaged sea-level trend and
386	removing the GIA signal shifts the original trend at these locations closer to the mean of the 17
387	locations. Regionally, the average GIA model trend is \sim -0.4 mm/yr for the North Sea which is larger in
388	magnitude than the GIA trend of ~0 mm/yr in the North Sea sea-level analysis of Frederikse et al.
389	(2016b); this difference may in part be due to the influence of the ANU ice sheet model in the prior
390	model, which predicts stronger subsidence over the North Sea than ICE-5G.

391

392 Removal of the GIA signal from all 17 locations increases the North Sea mean sea-level trend from 393 1.31 mm/yr to 1.58 mm/yr The GIA-corrected rates at 4 sites along the British Isles coastline (12, 13, 394 14 and 16) fall outside the standard deviation of the mean corrected rate. In the northern British Isles, 395 around sites 13 and 14 (Wick and Aberdeen), model D1 underpredicts the magnitude of vertical 396 motion and thus also the magnitude of relative sea-level change. However, even if the magnitude of 397 RSL fall were larger in this region by up to 0.5 mm/yr, the GIA-corrected sea-level rates at Wick and 398 Aberdeen would remain outside the standard deviation of the mean. At station Wick, the sea-level 399 trend is particularly variable and non-linear at decadal scales (Frederikse et al. 2016a), suggesting 400 that one averaged time-varying rate cannot be expected to adequately describe sea-level variation at 401 this location. At any rate, such variability is insensitive to application of the relatively small and linear 402 GIA correction for this region and it appears unlikely that the variability in sea-level trends along the





- 403 British coast can be explained by GIA-induced sea-level change. Conversely, the variability in sea-
- 404 level trends in the northeast North Sea, near the former FIS, is easily attributed to GIA.



405

Figure 11. Comparison of mean total, long-term GIA and non-GIA sea-level trends (grey boxes, blue
 triangles, red circles) for 17 tide gauge stations in the North Sea. Long-term GIA trends are from
 model D1, mean sea-level trends are from Model C of Frederikse et al. (2016a).

409

410 4. Conclusion

- 411 We generate a data-driven prediction of the long-term GIA response at present-day in Scandinavia,
- 412 northern Europe and the Barents Sea through the simultaneous inversion of GPS-measured vertical
- 413 motion rates, GRACE-measured gravity change rates, and a priori GIA model information. In models
- 414 D1-D3, we predict GIA motions for the inversion of the vertical motion data, the gravity data, and both
- 415 datasets. In both the χ^2 and RMS sense, the vertical motion data alone have the poorest ability to
- 416 predict gravity change, and vice versa. Predictions of the D3 model provide the best overall fit to both
- 417 datasets.





In general, prediction of the gravity signal is problematic, with larger χ^2 values than those obtained for the vertical motion prediction. The poorer prediction of gravity change is in part due to the uncertainty of the present-day mass loss effect in the Barents Sea region. The mass loss signal estimated by GRACE over Svalbard and the Russian Arctic is significantly smaller than estimates obtained from satellite altimetry. This difference may be the result of signal loss in the GRACE data from application of the Wiener filter or may also indicate that there is a non-zero component of ongoing glacial isostatic adjustment from the LIA.

426

427 The vertical motion signal is overall better predicted than the gravity signal. Both the D1 and D3 428 models have χ^2 values of \leq 1 and predict rates of vertical motion that are within the 1₅ uncertainty of 429 the observations throughout most of the study area. Regions of misfit persist in Scotland and around 430 the Netherlands, where the model underpredicts and overpredicts rates of vertical motion, 431 respectively. The misfit in Scotland may be partly due to both positive and negative rates of vertical 432 motion that are present in the data over relatively short distances. Further analysis and filtering of the 433 GPS dataset may be useful in this region. In the Netherlands, Kooi et al. (1998) found that present-day 434 subsidence from sediment compaction as well as tectonic movements may contribute significantly to 435 vertical land motion; correction for these effects may serve to reduce some of the misfit in this region.

436

437 The prediction of vertical land motion has a small but non-negligible sensitivity to the application of an 438 elastic correction. The elastic correction applied in this study is between 0.2-0.5 mm/yr; the largest 439 contribution comes from mass loss of the Greenland Ice Sheet which yields regional uplift with a 440 southeastward decreasing gradient. When the model predictions from another semi-empirical model of 441 GIA vertical motion, NKG2016LU, are compared to the corrected GPS data, a small but uniform bias of +0.42 mm/yr is present in the model predictions over Scandinavia. Conversely, when D1 model 442 443 predictions generated with the corrected data are compared to the uncorrected data from the same region, a uniform bias of -0.35 mm/yr is present, consistent with expectations. Both NKG2016LU and 444 D1 (and D3) have vertical motion χ^2 values \leq 1 over their respective study areas. However, while the 445 446 magnitude of the bias is smaller than the observational uncertainty on many of the measurements, it is 447 generally larger than the estimated posterior model uncertainty. Also, because only anthropogenic





- 448 hydrological signals (and not natural hydrological signals) were included in the elastic correction, it is
- 449 likely that the applied elastic correction is conservative in this region.

450

451	Therefore, the presence of such a bias in the vertical motion prediction suggests that while long-term
452	GIA is the dominant contributor to vertical motion in central Scandinavia, that it is still worthwhile to
453	correct GPS land motion rates for present-day elastic signals, so long as these signals are adequately
454	approximated (e.g., Riva et al. 2017). This conclusion however highlights a fundamental assumption
455	that underpins the data-driven methodology: that the input data can be adequately 'cleaned' for
456	processes not arising from long-term GIA. Even with applied corrections for hydrology and
457	contemporary ice mass loss, this assumption may not always be adequate, especially in regions
458	where model misfits relative to the data are spatially coherent. Thus, the success of data-driven GIA
459	predictions are evaluated by two criteria: i) the estimation of realistic posterior uncertainties that are
460	smaller than those associated with a priori knowledge and measurement uncertainty, and ii) the ability
461	of the final model to provide a good fit to the data. The vertical motion predictions of models D1 and
462	D3 satisfy both criteria for most of the study area and thus can provide a useful tool with which to
463	separate long-term GIA signals from shorter-term forcing.
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467





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