

1 Renormalisation of Global Mantle Dynamic Topography Predictions using  
2 Residual Topography Measurements for “Normal” Oceanic Crust

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4 **L. Cowie\*** <sup>1,2</sup> and **N. Kuszniir** <sup>2</sup>

5 <sup>1</sup> Present address: Centre for Earth Evolution and Dynamics (CEED), University of Oslo, Norway

6 <sup>2</sup> Earth, Ocean and Ecological Sciences, University of Liverpool, Liverpool, L69 3BX, United Kingdom

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9 **Corresponding Author:** Leanne Cowie

10 **Address:** Centre for Earth Evolution and Dynamics (CEED), University of Oslo, P.O. Box 1028 Blindern, NO-  
11 0315 Oslo, Norway

12 **Email:** leanne.cowie.87@gmail.com

13 **Phone:** 07765673285

15 **Abstract**

16 We compare mantle dynamic topography predicted from mantle convection modelling with residual  
17 topography measurements for oceanic regions, where crustal basement thickness is 10.0 km or less.  
18 Measurements of residual topography, calculated by removing the isostatic effects of crustal thickness  
19 variation, bathymetry, sediments, ice and lithosphere thermal anomalies, from the observed topography,  
20 are inaccurate for continents and oceanic plateaus due to uncertainties in determining their crustal  
21 thickness and density. As a consequence, residual topography measurements for these regions are  
22 unsuitable for testing mantle dynamic topography predictions. Residual topography is more accurately  
23 determined for oceanic crust. We use global mapping of crustal basement thickness using gravity anomaly  
24 inversion to identify oceanic crust of 10.0 km thickness or less to select measured residual topography for  
25 comparison with predicted mantle dynamic topography. For these oceanic regions we compare mantle  
26 dynamic topography and residual topography and, using amplitude histogram matching and grid searches,  
27 compute the amplitude rescaling and shift which needs to be applied to predicted mantle dynamic  
28 topography to fit the observed residual topography. We examine three global compilations which use  
29 different approaches to determine mantle dynamic topography: (i) Steinberger (2007), which uses seismic  
30 topography deeper than 220 km to determine mantle density; (ii) Flament et al. (2013), which uses plate  
31 velocity and subduction history; and (iii) Steinberger et al. (2017), which uses seismic tomography, including  
32 that above 220 km, to determine shallow upper mantle densities. Our analysis shows that for the  
33 Steinberger (2007) and Flament et al. (2013) compilations, the predicted mantle dynamic topography for  
34 oceanic regions requires a rescaling of approximately  $\times 0.5$  and a negative shift of approximately -500 m to  
35 match the observed residual topography. In contrast Steinberger et al. (2017), which includes shallow  
36 upper mantle densities above 220 km, requires only a small shift (+50 m) but a greater scaling of  $\times 0.375$ .  
37 Maps of renormalised (rescaled and shifted) mantle dynamic topography for Steinberger et al. (2017) show  
38 a close resemblance to measured residual topography.

39

40 **Keywords**

41 Mantle Dynamic Topography

42 Residual Topography

43 Oceanic crust

44 Gravity Inversion

## 46 **1. Introduction**

47 Large-scale variations in the Earth's surface topography originate from both changes in crustal and  
48 lithosphere thickness, lithosphere temperature structure and composition, and from convective viscous  
49 flow within the mantle (e.g. Ricard et al. (1993); Lithgow-Bertelloni and Silver (1998); Gurnis et al. (2000)  
50 and Steinberger (2016)). These two contributions to the support of the Earth's topography have been  
51 termed isostatic and mantle dynamic (Allen, 1997). Mantle dynamic topography includes both the  
52 consequences of whole-mantle thermal convection resulting in the large scale three-dimensional variations  
53 in mantle density, and smaller smaller-scale convection processes associated with mantle plume activity  
54 and subduction. The magnitude of mantle dynamic topography is related to the intensity and depth of  
55 mantle flow, whilst the wavelength is proportional to the scale and depth of the flow (Richards and Hager,  
56 1984).

57 Two approaches may be used to examine mantle dynamic topography; (i) predictive forward modelling  
58 using 3D mantle convection and (ii) measurement by subtracting the isostatic component of topography  
59 from observed topography to give what is often termed residual topography. Within this paper we use the  
60 term mantle dynamic topography to refer to present day predictions and residual topography to refer to  
61 observations and measurements.

62 Knowledge of mantle dynamic topography is important, in particular the amplitude and what drives it. Due  
63 to its importance, attempts have been made to constrain mantle dynamic topography using observations of  
64 its geological pattern, and many authors have used the observations of residual topography to test  
65 predictive models of mantle dynamic topography (e.g. Lithgow-Bertelloni and Silver (1998); Kaban et al.  
66 (1999); Gurnis et al. (2000); Panasyuk and Hager (2000); Flament et al. (2013) and Steinberger et al. (2017)).

67 While at long wavelengths a global comparison of measured residual topography with predicted mantle  
68 dynamic topography shows some similarity, significant differences are observed particularly in the Pacific  
69 regions where mantle dynamic topography is substantially more positive than residual topography. Many

70 authors (e.g. Watkins and Conrad (2018); Steinberger et al. (2017); Hoggard et al. (2016) and Yang and  
71 Gurnis (2016)) believe that predicted mantle dynamic topography amplitudes are too great when  
72 compared to observations of residual topography. Steinberger (2016) suggests that at long wavelengths,  
73 corresponding to degree 2, the amplitude of predicted mantle dynamic topography may be twice as large  
74 as the observed residual topography. Hoggard et al. (2016) suggest that the power spectrum for predicted  
75 mantle dynamic topography at long wavelengths may be an order of magnitude greater than the observed  
76 residual topography, while at short wavelengths the opposite occurs with greater amplitudes in the  
77 residual topography compared with predicted mantle dynamic topography.

78 It is therefore important to better constrain the mantle dynamic topography models. Many groups have  
79 used observations of residual topography to constrain predictions of mantle dynamic topography. Watkins  
80 and Conrad (2018) developed new constraints on the amplitude of long-wavelength mantle dynamic  
81 topography by examining asymmetries in seafloor bathymetry across mid-ocean ridges. Steinberger (2016)  
82 compares predicted mantle dynamic topography, from a new model incorporating shallow upper mantle  
83 density structure, with observations of residual topography, in terms of both correlation and amplitude  
84 ratio. Yang and Gurnis (2016) investigated the relationship between free-air gravity and mantle dynamic  
85 topography as a function of wavelength and also investigated the possibility that the observed free-air  
86 gravity anomalies and the large amplitude long-wavelength mantle dynamic topography can be reconciled.  
87 They attempted to verify that at long wavelengths their predicted mantle dynamic topography is consistent  
88 with observed residual topography (from Hoggard et al. (2016)), in both pattern and amplitude.  
89 Steinberger et al. (2017) used the unclear nature of the interpretations of two large, seismically slow  
90 regions in the lower mantle beneath Africa and the Pacific oceans (and whether they are large-scale active  
91 upwellings or represent collections of regular mantle plumes) to investigate the implications of these  
92 upwellings for mantle dynamic topography. In order to do this, they compared their modelled mantle  
93 dynamic topography with a new compilation of observed residual topography.

94 While there is much uncertainty in the model prediction of mantle dynamic topography, there is also  
95 uncertainty in the measured residual topography particularly for the continental regions. The calculation of

96 the isostatic topography component for the continents requires knowledge of crustal thickness and density,  
97 both of which are uncertain. Similar problems in determining the isostatic correction for topography exist  
98 for oceanic plateaus and micro-continents in oceanic regions. Residual topography is more reliably  
99 measured in oceanic regions where oceanic crust is of normal or near normal thickness.

100 In this paper we use an alternative and independent method to compare predicted mantle dynamic  
101 topography with observed residual topography. In addition, we produce global maps of renormalised  
102 mantle dynamic topography which provide a better fit the measured residual topography. As a strong  
103 contribution to the present-day surface topographic signal arises from crustal thickness variations, we  
104 restrict the comparison of predicted mantle dynamic topography with residual topography to oceanic  
105 regions with normal or near normal crustal thickness where uncertainty in crustal thickness and density are  
106 minimized. To achieve this, we use global mapping of crustal thickness, from gravity inversion, to identify  
107 normal thickness oceanic crust and avoid oceanic plateaus and micro-continents.

## 108 **2. Compilations of Observed Residual Topography and Predicted Mantle Dynamic** 109 **Topography Examined in this Study**

110 We compare mantle dynamic topography with residual topography for three predictive models. These are:  
111 Steinberger (2007); Flament et al. (2013); and Steinberger et al. (2017), as shown in Figure 1. We choose  
112 these because they represent distinct approaches for the prediction of mantle dynamic topography.

113 The Steinberger (2007) model used a mantle convection model following the approach of Hager and  
114 O'Connell (1979, 1981). Density distribution within the convection model (below 220 km) was derived from  
115 seismic tomography assuming velocity anomalies are the result of temperature variations. The model  
116 surface boundary condition used present-day plate velocities from NUVEL (DeMets et al., 1990). The  
117 effects of the 660 and 410 km phase transitions were included in the model.

118 The Flament et al (2013) model used an approach based on subduction history similar to Ricard et al. (1993).  
119 The surface boundary conditions of the mantle convection models used plate velocities for the last 200 Myr  
120 derived from Seton et al. (2012) to compute the subduction input into the mantle. In order to suppress the

121 effect of surface traction imposed by plate velocities, the dynamic topography itself is computed with a no-  
122 slip surface boundary condition. This modelling methodology only produces the long wavelength  
123 components of mantle dynamic topography, in contrast to the approach used by Steinberger (2007) which  
124 also produces shorter wavelength components.

125 The Steinberger et al. (2017) model uses a methodology based on that used in Steinberger (2007) but with  
126 additional features. In particular it differs in that it includes the density anomalies derived from  
127 tomography above 220 km (as well as those below). For the comparison of predicted mantle dynamic  
128 topography and measured residual topography it is important that the residual and mantle dynamic  
129 topography models are derived in a mutually consistent manner (e.g. the oceanic lithosphere thermal  
130 correction). As a consequence, for the comparison we use the paired grids compilations of mantle dynamic  
131 and residual topography, which have been provided by Steinberger and Flament.

132 The Steinberger (2007) residual topography grids were prepared using an isostatic correction for ocean  
133 lithosphere cooling with oceanic ages from Mueller (1997) and an isostatic correction for continental  
134 crustal thickness variation using Crust2.0. The Steinberger (2007) residual topography (and mantle dynamic  
135 topography) grids are air loaded for both oceans and continents.

136 For oceanic regions the Flament et al (2013) residual topography grids were prepared using the plate  
137 cooling model of Crosby & McKenzie (2009) and a sediment correction. For continental regions a mean  
138 elevation correction was applied rather than a correction for crustal thickness variation. Consequently, the  
139 residual topography for the continents (and continental shelves) is likely to be unreliable. The oceanic  
140 values are however, more reliable and are used in this study. The Flament et al (2013) residual topography  
141 (and mantle dynamic topography) grids are water loaded for oceans and air loaded for the continents.

142 For the oceans the Steinberger et al. (2017) residual topography grids were based on the detailed  
143 compilation of Hoggard et al. (2016) which were made using corrections for ocean lithosphere age,  
144 sediment thickness and oceanic crustal thickness variations. For the continents, the corrections for crustal  
145 thickness variations were made using Crust2.0. The Steinberger et al. (2017) residual topography (and  
146 mantle dynamic topography) grids are air loaded for both oceans and continents.

147 All three compilations show positive mantle dynamic topography and residual topography in the SW Pacific  
148 (Figure 1). However, the predicted mantle dynamic topography for Steinberger (2007) and Flament et al.  
149 (2013) is much larger in amplitude and extent than the observed residual topography. For the Steinberger  
150 (2007) compilation the residual topography is positive for east Asia and eastern South America, while the  
151 corresponding mantle dynamic topography is negative. For the Flament et al. (2013) compilation a major  
152 discrepancy is seen for Eurasia and the west coast of South America where in both cases the residual  
153 topography is positive while the mantle dynamic topography is negative. We attribute these differences in  
154 the Flament et al. (2013) compilations to an inappropriate procedure used to determine their continental  
155 residual topography. In contrast we do not see such obvious discrepancies within the Steinberger et al.  
156 (2017) compilations.

### 157 **3. Identification of the Global Distribution of “Normal” Thickness Oceanic Crust**

158 The aim of this study is to compare predicted mantle dynamic topography with observed residual  
159 topography for oceanic regions where the isostatic correction for crustal thickness variation can be made  
160 more accurately. This requires the identification of oceanic crust with “normal” thickness oceanic crust,  
161 which we determine using global mapping of oceanic crustal thickness using gravity anomaly inversion as  
162 shown in Figure 2 (a).

163 The data used within the gravity anomaly inversion are bathymetry (Amante and Eakins, 2009), satellite  
164 derived free air gravity (Sandwell et al., 2014), three-dimensional sediment thickness (Divins, 2003; Laske  
165 and Masters, 1997) and ocean age isochrons from Müller et al. (1997). The sediment thickness, used within  
166 the gravity anomaly inversion, is a merged compilation of both the Laske and Masters (1997) and NOAA  
167 (Divins, 2003) sediment thickness data sets, in which the merged resulting sediment thickness is the larger  
168 of the two values.

169 The gravity anomaly inversion method is carried out in the 3D spectral domain, using the scheme of Parker  
170 (1972) to predict Moho depth and hence determine crustal basement thickness. A detailed description of  
171 the gravity anomaly inversion technique methodology is given in Chappell and Kuszniir (2008), Alvey et al.

172 (2008), Cowie and Kuszniir (2012) and Kuszniir et al. (2018). A lithosphere thermal gravity anomaly  
173 correction is incorporated to account for the elevated geothermal gradient within oceanic and rifted  
174 continental margin lithosphere; failure to include the correction for the lithosphere thermal gravity  
175 anomaly can lead to predictions of Moho depth and crustal basement thickness which are substantially too  
176 great.

177 The gravity inversion methodology to produce crustal thickness requires a reference Moho depth to  
178 determine 3D Moho depth (Chappell and Kuszniir, 2008). The reference Moho depth is the notional depth  
179 of the Moho for a region with zero topography or bathymetry, zero sediment thickness, zero long  
180 wavelength free-air gravity anomaly and zero lithosphere thermal gravity anomaly. The reference Moho  
181 depth has a global average of approximately 38 km from calibration against seismic observation, assuming  
182 a crustal basement density of  $2850 \text{ kgm}^{-3}$ , and is independent of whether the crust is oceanic or continental.  
183 Because the long wavelength components of the Earth's gravity anomaly field are controlled by deep  
184 mantle dynamic processes, which are not related to lithosphere and crustal structure, the reference Moho  
185 depth varies globally and correlates weakly with mantle dynamic topography. In this study a constant value  
186 of reference Moho depth of 38 km has been used, which assumes zero mantle dynamic topography.

187 The resulting global map of crustal basement thickness determined from gravity anomaly inversion is used  
188 to identify crust with thicknesses greater than that of "normal" oceanic crust, in order to eliminate and  
189 filter out the less accurate measurements of residual topography for regions of thicker crust. White et al.  
190 (1992) report that seismic refraction results show that "normal" oceanic crust averages  $7.1 \pm 0.8$  km away  
191 from anomalous regions, such as fracture zones and hotspots. The upper and lower bounds of "normal"  
192 oceanic crustal thickness range between 5.0 km and 8.5 km. When identifying regions of thick oceanic  
193 crust, sensitivities to oceanic crustal basement thickness filters with maximum values of 5.0 km, 7.0 km and  
194 10.0 km (Figure 2 (b, c and d)) have been considered, in order to be inclusive and ensure magma poor and  
195 magma rich regions are accounted for. Whilst we show the results for these sensitivities, our preference is  
196 for the 10.0 km crustal basement thickness filter to maximise the sampling of oceanic regions.

#### 197 **4. Comparison of Observed Residual Topography and Predicted Mantle Dynamic** 198 **Topography for "Normal" Thickness Oceanic Crust**

199 Cross-plots of global predicted mantle dynamic topography against measured residual topography at  
200 identical locations for the Steinberger (2007), Flament et al. (2013) and Steinberger et al. (2017)  
201 compilations are shown in Figure 3 (a, e and i). Both the Steinberger (2007) and Flament et al. (2013)  
202 compilations show a highly scattered cloud of points with poor correlation between predicted mantle  
203 dynamic topography and measured residual topography. In contrast, the Steinberger et al. (2017)  
204 compilation shows a tighter cloud of points, with a more positive correlation. The global cross-plots, in  
205 Figure 3, show points that plot generally within the range -1000 m and 1000 m for both mantle dynamic  
206 topography and residual topography. The measured residual topography from the Flament et al. (2013)  
207 compilation shows a sharp cut-off at -750 m; the cause of this is explained by Flament et al. (2013) (see  
208 below also). The poor correlation of global mantle dynamic topography with residual topography shown in  
209 Figure 3 (a and e) for the Steinberger (2007) and Flament et al. (2013) compilations may be in part due to  
210 errors within the crustal thickness and density component of the determination of residual topography,  
211 particularly in continental regions.

212 When the crustal basement thickness filter is applied, for crust of 10.0 km thickness or less (Figure (d)), the  
213 cross-plot for the remaining regions of "normal" oceanic crust shows a tighter range for all three  
214 compilations (Figures 3 (b, f and j)). As the values for thicker crust are removed, the more positive residual  
215 topography values are filtered out.

216 The comparison of predicted mantle dynamic topography and measured residual topography can also be  
217 made using an amplitude histogram plot as shown in Figure 3 (c, g and k). The result of applying the crustal  
218 thickness filter to produce histograms of mantle dynamic and residual topography for crust of thickness  
219 10.0 km or less is shown in Figure 3 (d, h and i). The amplitude histograms for crust thinner than 10.0 km  
220 show that mantle dynamic topography, for all three compilations, has a larger amplitude compared with  
221 residual topography.

222 In the amplitude histograms of measured residual topography from Flament et al. (2013), shown in Figures  
223 3 (g and h), there is an additional peak at approximately -750 m. These peaks are the result of how the  
224 residual topography has been calculated for the area between the continental shelf and the ocean floor  
225 (see Flament et al. (2013) for more details). This additional peak is reduced but not entirely eliminated for  
226 crust less than 10.0 km thickness.

227 Both the Flament et al. (2013) and the Steinberger et al. (2017) residual topography (Figure 3 (h and i)) are  
228 symmetrical about zero, whilst the peak of residual topography for Steinberger (2007) (Figure 3 (d)) is  
229 slightly negative; the cause and implication of this is unknown.

## 230 **5. Renormalisation of Predicted Mantle Dynamic Topography**

231 The aim of renormalisation is to adjust the predicted mantle dynamic topography to better fit the  
232 measured residual topography. Two methods have been used to calculate the rescaling factor and the shift  
233 required to renormalise the predicted mantle dynamic topography; these include a histogram matching  
234 method and a grid search using a L1 norm regularization method.

235 The histograms shown in Figures 4, 5 and 6 show the un-normalised mantle dynamic topography and  
236 associated renormalised mantle dynamic topography for the Steinberger (2007), Flament et al. (2013) and  
237 Steinberger et al. (2017) compilations. Crustal basement thickness filters of 5.0, 7.0 and 10.0 km have been  
238 used. Comparison of mantle dynamic topography and residual topography shows a similar relationship for  
239 all crustal thickness filter sensitivities. As the crustal thickness value in the filter is decreased, fewer points  
240 for comparison are selected and consequently the height of the histograms decreases

### 241 **5.1. Renormalisation by Matching Histograms Amplitudes**

242 We match the histograms of mantle dynamic and residual topography by rescaling and shifting the mantle  
243 dynamic topography. We apply a shift to the mantle topography in order to match the average mantle  
244 dynamic topography with the average observed residual topography. The shift is sensitive to the crustal  
245 basement thickness filter used when eliminating thicker than “normal” oceanic crustal regions. At lower  
246 crustal thickness filters the results may be less reliable as there are less data used to calculate the shift. For

247 the histogram matching we examine two different approaches. The first approach matches the height of  
248 the histogram peaks of the rescaled mantle dynamic topography with that of the observed residual  
249 topography, while the second approach matches their standard deviation.

250 In Figure 4 histograms of shifted and rescaled mantle dynamic topography are compared with those for  
251 observed residual topography for the Steinberger (2007) compilations for crustal basement thickness filters  
252 of 5.0, 7.0 and 10.0 km. The required shift and rescaling values to match histograms are shown on Figure 4.  
253 For the Steinberger (2007) compilation a negative shift needs to be applied to the mantle dynamic  
254 topography for all crustal thickness filters; for a crustal basement filter of 10.0 km the required shift is -665  
255 m. The rescaling factor is  $\times 0.75$  using the matched histogram peak method and  $\times 0.85$  using the matched  
256 standard deviations method. For both histogram matching methods the scaling factor decreases as the  
257 crustal basement thickness filter decreases from 10.0 to 5.0 km. The dependency of the shift and the  
258 rescale factors on crustal thickness filter settings is shown in Figure 7 for all three compilations.

259 Corresponding histograms for the Flament et al. (2013) compilation are shown in Figure 5. For a crustal  
260 basement thickness filter of 10.0 km the required shift is -472 m while the rescaling factor is  $\times 0.775$  and  
261  $\times 0.76$  using the matched histogram peak and matched standard deviation methods respectively. The shift  
262 shows a similar dependency on the crustal thickness filter to the analysis of the Steinberger (2007)  
263 compilation (Figure 7). As before, decreasing the crustal thickness filter decreases the required scaling  
264 factor for the matched histogram peak method. However, for the matched standard deviation method the  
265 rescaling factor does not show a consistent trend with the crustal thickness filter.

266 Corresponding histograms for the Steinberger et al. (2017) compilation are shown in Figure 6. For a crustal  
267 thickness filter of 10.0 km the required shift is +50 m while the rescaling factor is  $\times 0.375$  using the matched  
268 histogram peak method and  $\times 0.45$  using the matched standard deviation method. The shift in this case is  
269 positive and is also much smaller in comparison to the shifts calculated for the Steinberger (2007) and  
270 Flament et al. (2013) compilations. For the Steinberger et al. (2017) compilation decreasing the crustal  
271 thickness filter decreases the required scaling factor for both the matched histogram peak and standard  
272 deviation methods.

## 273 **5.2. Renormalisation by Grid Search using L1 Norm Regularization Method**

274 We also determine the rescaling factor and shift to match mantle dynamic and residual topography by  
275 applying a grid search using a L1 norm regularization method (Figure 8). The L1 norm method minimizes the  
276 sum of absolute differences between the target value (residual topography in this paper) and the estimated  
277 value (mantle dynamic topography in this paper) at each identical point, and assumes that there is a  
278 correlation between these two parameters. A grid search is carried out varying the rescaling factor and  
279 shift to identify the minimum misfit. Within the grid search we are attempting to correlate a long  
280 wavelength parameter (mantle dynamic topography) with a measured parameter with both short and long  
281 wavelength components (residual topography). For obvious reasons it is inappropriate to spatially filter the  
282 residual topography.

283 The results of the L1 norm grid search method are shown in Figure 8 for all three compilations for crustal  
284 thickness filter sensitivities of 5.0, 7.0 and 10.0 km or less. The calculated shift of mantle dynamic  
285 topography determined using this grid search L1 norm method is approximately -800 m, -500 m and +100 m  
286 for the Steinberger (2007), Flament et al. (2013) and Steinberger et al. (2017) compilations respectively,  
287 which are similar to those calculated from the histogram matching methods. However, the re-scaling factor  
288 from the grid search L1 method is between  $\times 0.1$  and  $\times 0.2$  which is much less than the  $\times 0.8$  from histogram  
289 matching, for both Steinberger (2007) and Flament et al. (2013). In contrast, the L1 method re-scaling  
290 factor is approximately  $\times 0.3$  for Steinberger et al. (2017), which is much closer, yet still less than that  
291 calculated from the matched histograms method.

292 The smaller rescaling factors may be the result of a weak correlation between the predicted mantle  
293 dynamic topography and the measured residual topography, which is more apparent for the Steinberger  
294 (2007) and Flament et al. (2013) compilations. In order to test the weak correlation hypothesis, the mantle  
295 dynamic topography was set to zero and compared to the real residual topography within the grid search;  
296 the grid search yields a rescaling factor of zero. This suggests that the grid search method is not a reliable  
297 method for calculating the rescaling factor, if the correlation is weak between the predicted mantle  
298 dynamic topography and the measured residual topography.

299 Our preference is for the renormalisation rescaling factor and shift calculated from the histogram matching  
300 method.

## 301 **6. Comparison with other studies**

302 We have compared our rescaling of mantle dynamic topography to fit measured residual topography with  
303 other studies, in particular recent papers by Hoggard et al. (2016) and Steinberger (2016).

304 Hoggard et al. (2016) use power spectral analysis of their observed residual topography and published  
305 predictive models of mantle dynamic topography in order to investigate the differences between them.

306 They show that the power spectrum of the predicted mantle dynamic topography models shows a peak at  
307 long wavelengths, which decreases to zero at shorter wavelengths, whilst the power spectrum of the  
308 observed residual topography has a more consistent power for all wavelengths (i.e. its spectrum is flatter).

309 In comparison with the power spectral analysis for the mantle dynamic topography, the residual  
310 topography has approximately an order of magnitude less power at the longer wavelengths; however, at  
311 shorter wavelengths the residual topography has significantly greater power. Our renormalisation results  
312 are consistent with these long wavelength observations.

313 Steinberger (2016) presents a new mantle dynamic topography model incorporating densities shallower  
314 than 220 km depth, derived from seismic tomography, which he compares with observed residual  
315 topography using correlation and amplitude ratios. Various sensitivity analyses have been considered,  
316 including but not limited to mantle viscosity, velocity-density scaling and surface boundary conditions. His  
317 preferred model predicts a topography only slightly greater (x1.2) than residual topography, and also  
318 provides an acceptable fit to the geoid. This preferred mantle dynamic topography prediction is similar in  
319 amplitude to our renormalisation results for the Steinberger et al. (2017) compilation.

## 320 **7. Renormalised Mantle Dynamic Topography**

321 Figure 9 shows a comparison of the original modelled and renormalised mantle dynamic topography for  
322 Steinberger (2007), Flament et al. (2013) and Steinberger et al. (2017) where the rescaling factor and shift  
323 was derived from the matched heights of histogram peaks method for crustal thicknesses of 10.0 km or less

324 (Figures 4, 5 and 6). Renormalisation using the matched histogram standard deviations method produces  
325 maps which are not significantly different.

326 Both the renormalisation of mantle dynamic topography for the Steinberger (2007) and Flament et al.  
327 (2013) compilations (Figure 9 (d and e)) require a substantial negative shift (-665 m and -472 m respectively)  
328 and an amplitude reduction rescaling (x0.75 and x0.77 respectively). The renormalisation of mantle  
329 dynamic topography for the Steinberger (2017) compilation, in contrast, uses a much smaller shift (+50 m)  
330 but a more substantial rescaling factor is required (x0.375).

331 The renormalised mantle dynamic topography from Steinberger (2007) shows a Pacific Ocean with positive  
332 mantle dynamic topography restricted to the South West Pacific. Elsewhere in the Pacific and also in the  
333 Atlantic and Indian Oceans renormalised mantle dynamic topography is negative. Negative mantle dynamic  
334 topography is also predicted, by renormalisation, for the Americas, Africa and Eurasia. The negative  
335 renormalised mantle dynamic topography for Africa in particular suggests that the negative  
336 renormalisation shift is too great and incorrect.

337 The renormalised mantle dynamic topography from Flament et al. (2013) shows positive mantle dynamic  
338 topography for the Pacific and Atlantic Oceans and for South Africa. Renormalised mantle dynamic  
339 topographies for these regions are between +500 m and +1000 m. The Americas and Eurasia show negative  
340 mantle dynamic topography; for Eurasia the renormalised mantle dynamic topographies have a minimum  
341 between -1500 m and -2000 m. The Indian Ocean shows positive mantle dynamic topography in the west  
342 and negative in the east. It should be noted that the renormalisation of this and other predicted mantle  
343 dynamic topographies only use the oceanic residual topography; as a consequence the continental residual  
344 topographies of Flament et al. (2013), which are likely to be unreliable, have no impact on the  
345 renormalisation.

346 The renormalisation of Steinberger et al. (2017), as mentioned previously, requires a negligible offset but a  
347 more significant rescaling. As a consequence it shows a similar pattern to the original mantle dynamic  
348 topography but with lower amplitude. The renormalised mantle dynamic topography is positive in the  
349 south-west Pacific with amplitudes generally less than +500 m. Elsewhere in the east and north Pacific

350 values vary between  $\pm 350$  m. In the Atlantic renormalised mantle dynamic topographies are negative in  
351 the west; in the Argentine Basin values are  $-500$  m and greater. The north Atlantic shows positive regions  
352 corresponding to Iceland and the Azores. Elsewhere in the eastern and south Atlantic there are regions of  
353 low amplitude negative and positive renormalised mantle dynamic topography ( $< \pm 500$  m); similarly, in the  
354 Indian Ocean. South and east Africa show a positive renormalised mantle dynamic topography with values  
355 in north-east Africa exceeding  $+500$  m. Renormalised mantle dynamic topographies for Eurasia are  
356 generally negative with values down to  $-500$  m.

357 The original and renormalised mantle dynamic topographies for Steinberger et al. (2017) differ from those  
358 of Steinberger (2007) and Flament et al. (2013) in that they contain shorter wavelength components. This  
359 arises from the inclusion of seismic tomography data for depths of less than 220 km to provide lithosphere  
360 density information for the mantle convection modelling. It is conspicuous that while histogram matching  
361 of mantle dynamic and residual topography for the Steinberger (2007) and Flament et al. (2013)  
362 compilations required a substantial shift of  $-500$  m or more, the Steinberger et al. (2017) compilation does  
363 not. This zero offset for Steinberger et al. (2017) is confirmed by the grid search comparison of mantle  
364 dynamic and residual topography (Figure 8). We attribute this zero offset for the Steinberger et al. (2017)  
365 compilation to the inclusion of seismic tomography information above 220 km depth in the mantle  
366 convection prediction of mantle dynamic topography.

367 The renormalised mantle dynamic topography for Steinberger et al. (2017) is similar to the mantle dynamic  
368 topography model which includes the effect of lateral viscosity variations above 220 km from Osei Tutu et  
369 al. (2018) (illustrated in Figure 1 (c) of Steinberger et al. (2017)). This may suggest that the amplitude  
370 reduction rescaling in our renormalisation, based on oceanic comparisons, arises because lower upper  
371 mantle viscosity in oceanic regions reduces surface coupling of mantle convection mantle dynamic forces.  
372 The more complete lithosphere density structure within the mantle dynamic topography modelling and the  
373 inclusion of both long and short wavelengths makes the Steinberger et al. (2017) compilations our  
374 preference for the renormalisation results.

375 The comparison of modelled mantle dynamic and measured residual topographies, and the shift and  
376 rescaling to give renormalised mantle dynamic topography, has been carried out for oceanic crust of  
377 normal or near normal thickness with thicknesses less than or equal to 10.0 km. Due to the large errors  
378 and uncertainties in determining residual topography for the continents we do not believe that it is possible  
379 to carry out a meaningful comparison or recalibration for the continents. In Figure 10 we compare the  
380 observed residual topography with the renormalised topography for Steinberger et al. (2017) for crust less  
381 than or equal to 10 km thick. An important question that remains is whether the renormalisation derived  
382 for oceans can be applied to continental regions.

## 383 **8. Summary**

384 We have compared the modelled mantle dynamic topography predictions of Steinberger (2007), Flament et  
385 al. (2013) and Steinberger et al. (2017) with their associated measured residual topography. Due to the  
386 large errors and unreliability of measured residual topography in continental regions, oceanic plateaus and  
387 micro-continents, we restrict our comparisons to oceanic regions with crustal thickness less than 10.0 km,  
388 which corresponds to oceanic crust with normal (or near normal) crustal thickness where crustal thickness  
389 isostatic corrections may be made with more confidence. The amplitude rescaling and shift of predicted  
390 mantle dynamic topography to correspond to measured residual topography in oceanic regions has been  
391 examined using histogram matching and grid search methods. While the mantle dynamic topography  
392 predictions of Steinberger (2007) and Flament et al. (2013) were made using very different approaches (the  
393 former using seismic tomography and the latter subduction history to condition mantle convection models),  
394 the oceanic comparison with measured residual topography in both cases suggests that predicted mantle  
395 dynamic topography is too positive and also too large. Shifts of -665 m and -470 m and rescaling by  $\times 0.75$   
396 and  $\times 0.77$  of predicted mantle dynamic topography are required for the Steinberger (2007) and Flament et  
397 al. (2013) compilations respectively. In contrast, the compilation of Steinberger et al. (2017), which  
398 included density heterogeneities above 220 km derived from seismic tomography in the prediction of  
399 mantle dynamic tomography, requires no significant shift but a larger rescaling, by  $\times 0.375$ .

400 Our conclusion is that the predicted mantle dynamic topography in oceanic regions is too large and needs  
401 to be rescaled by approximately  $\times 0.35$  (between  $\times 0.3$  and  $\times 0.4$ ). At long wavelengths our analysis is  
402 consistent with the results of Hoggard et al. (2016) who use a different approach. They show that at long  
403 wavelengths predicted mantle dynamic topography is between  $\times 2$  and  $\times 10$  larger than observed residual  
404 topography (while at shorter wavelengths, they show that residual topography has greater amplitude than  
405 the predicted dynamic topography from models that exclude uppermost mantle density anomalies). The  
406 required rescaling of oceanic predicted mantle dynamic topography is consistent with the suggestion made  
407 by Steinberger et al. (2017) that mantle convection mantle dynamic forces are poorly coupled to the  
408 surface because of sub-lithosphere low viscosity structure.

409 We show that the renormalised (rescaled) predicted mantle dynamic topography from Steinberger et al.  
410 (2017) shows good agreement with the observed residual topography for the Pacific, Atlantic and Indian  
411 Ocean. As the renormalisation can only be calibrated for oceanic regions, it is uncertain whether the same  
412 rescaling of predicted mantle dynamic topography can be applied to the continents.

413

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420

421 **Figure Captions:**

422 **Figure 1:** Comparison of measured residual topography and model predicted mantle dynamic topography  
423 for the (a & d) Steinberger (2007), (b & e) Flament et al. (2013) and (c & f) Steinberger et al. (2017)  
424 compilations. **Figure 2:** (a) Global crustal basement thickness calculated from our gravity anomaly  
425 inversion. (b) Areas with crustal thickness of 5.0 km or less; (c) 7.0 km or less; (d) 10.0 km or less.

426 **Figure 3:** Comparison of the relationship between the observed residual topography and predicted  
427 mantle dynamic topography for the Steinberger (2007), Flament (2013) and Steinberger et al. (2017)  
428 compilations. Cross-plots showing the relationship between measured residual topography and predicted  
429 mantle dynamic topography (a, e & i) for all crustal basement thicknesses, and (b, f & j) for crustal  
430 thicknesses of 10 km or less. Histogram showing the relationship between measured residual topography  
431 and predicted mantle dynamic topography (c, g & k) for all crustal basement thicknesses, and (d, h & i) for  
432 crustal thicknesses of 10 km or less.

433 **Figure 4:** Histograms showing the un-normalised and renormalised relationships between measured  
434 residual topography and predicted mantle dynamic topography for the Steinberger (2007) compilation. (a,  
435 b & c) Histograms showing the un-normalised relationship between measured residual topography and  
436 predicted mantle dynamic topography filtered for crustal thickness less or equal to 5, 7 and 10 km. (d, e & f)  
437 Histograms showing residual topography and re-normalised mantle dynamic topography using the  
438 histogram matched heights method for crustal thickness less or equal to 5, 7 and 10 km. (g, h & i)  
439 Histograms showing residual topography and re-normalised mantle dynamic topography using the  
440 histogram matched standard deviations method for crustal thickness less or equal to 5, 7 and 10 km.

441 **Figure 5:** Histograms showing the un-normalised and renormalised relationships between measured  
442 residual topography and predicted mantle dynamic topography, and sensitivity to crustal thickness filter,  
443 for the Flament et al. (2013) compilation. Explanation of (a - i) similar to Figure 4. **Figure 6:** Histograms  
444 showing the un-normalised and renormalised relationships between measured residual topography and

445 predicted mantle dynamic topography, and crustal thickness filter, for the Steinberger et al (2017)  
446 compilation. Explanation of (a - i) similar to Figure 4.

447 **Figure 7:** (a) Sensitivity of the maximum crustal thickness filter to the calculated shift and (b) rescaling  
448 factor for the renormalisation of predicted mantle dynamic topography from the Steinberger (2007),  
449 Flament et al. (2013) and Steinberger et al. (2017) compilations. Within (b) the solid lines represent the  
450 matched peak (MP) method, whilst the dashed lines represent the standard deviation (SD) method.

451 **Figure 8:** Grid search of shift and scaling factor parameter space showing L1 norm misfit between  
452 renormalized predicted dynamic topography and observed residual topography for the Steinberger (2007),  
453 Flament et al (2013) and Steinberger et al. (2017) compilations, and sensitivity to crustal thickness filters of  
454 5, 7 and 10 km.

455 **Figure 9:** Comparison of model predicted mantle dynamic topography and renormalised mantle dynamic  
456 topography for the (a & d) Steinberger (2007), (b & e) Flament et al. (2013) and (c & f) Steinberger et al.  
457 (2017) compilations.

458 **Figure 10:** Comparison of the (a) observed residual topography; (b) renormalised mantle dynamic  
459 topography for crust less than 10 km thick, for the Steinberger et al. (2017) compilation.

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532

Figure 1

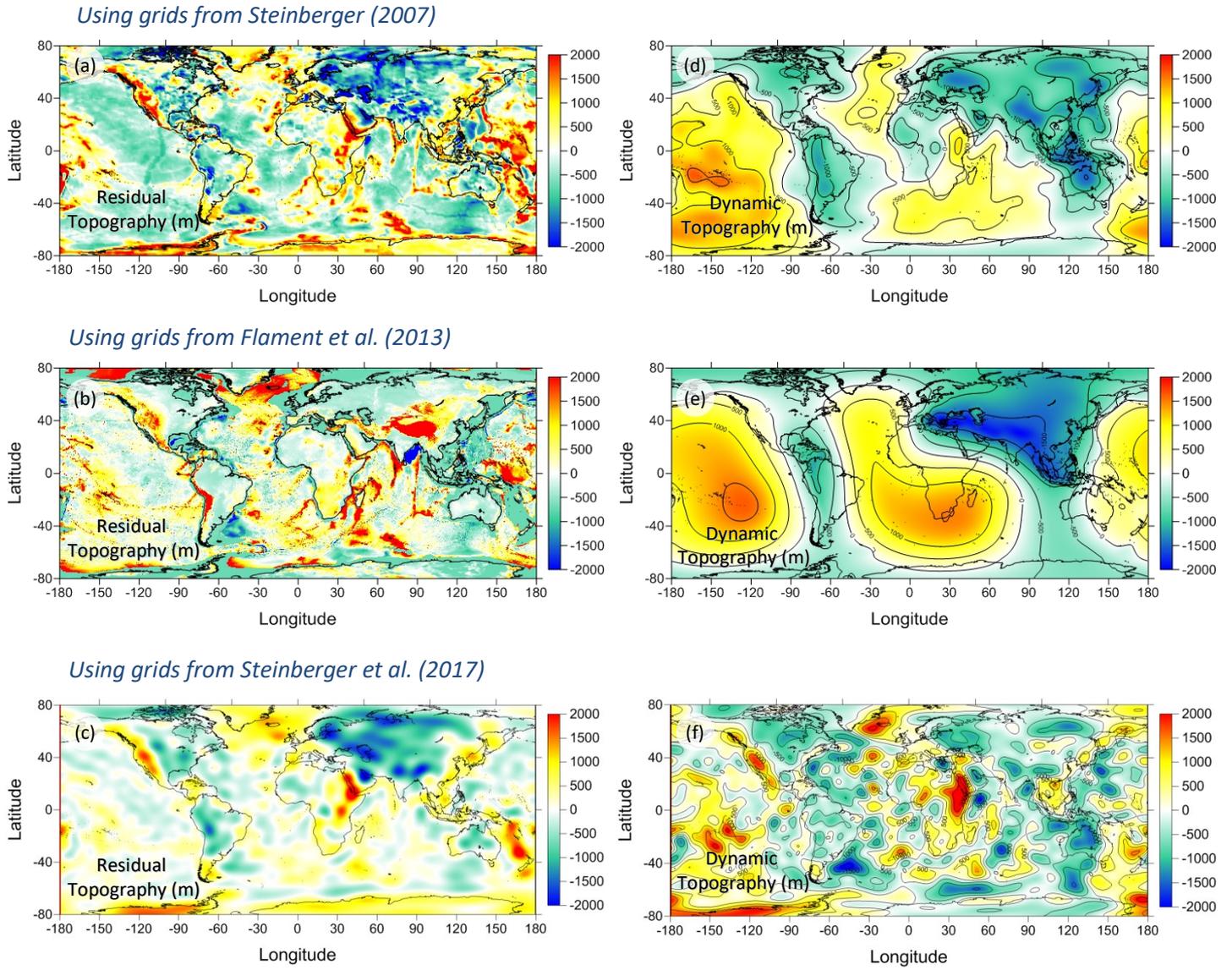


Figure 2

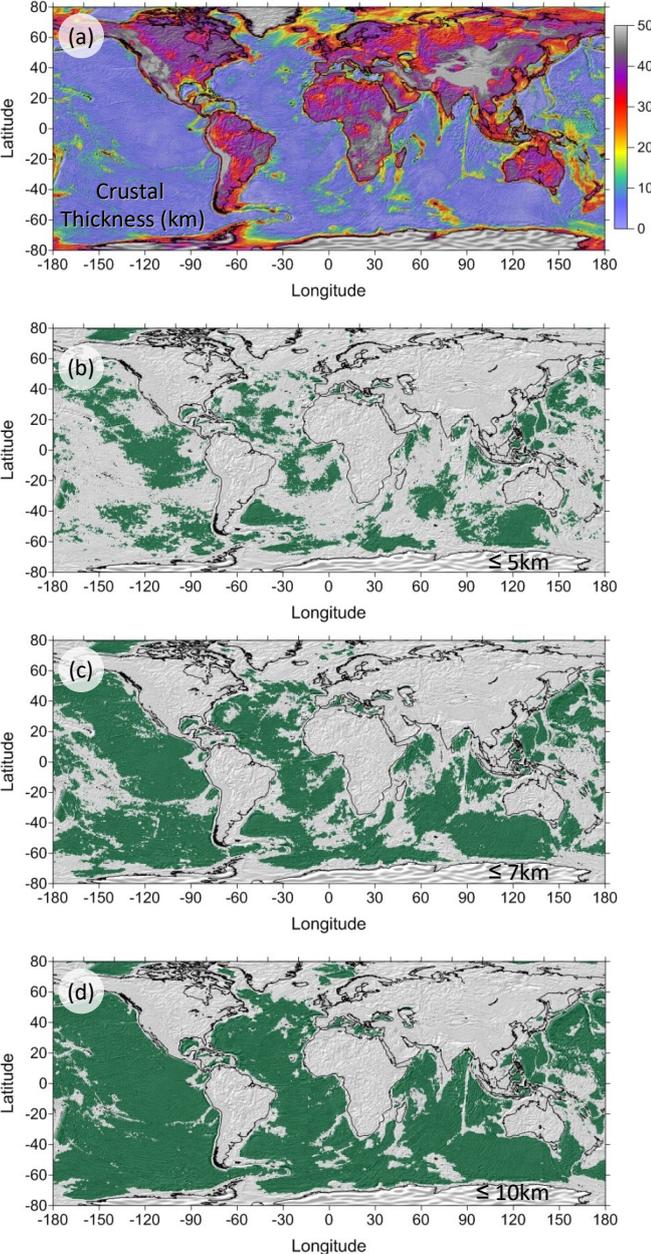


Figure 3

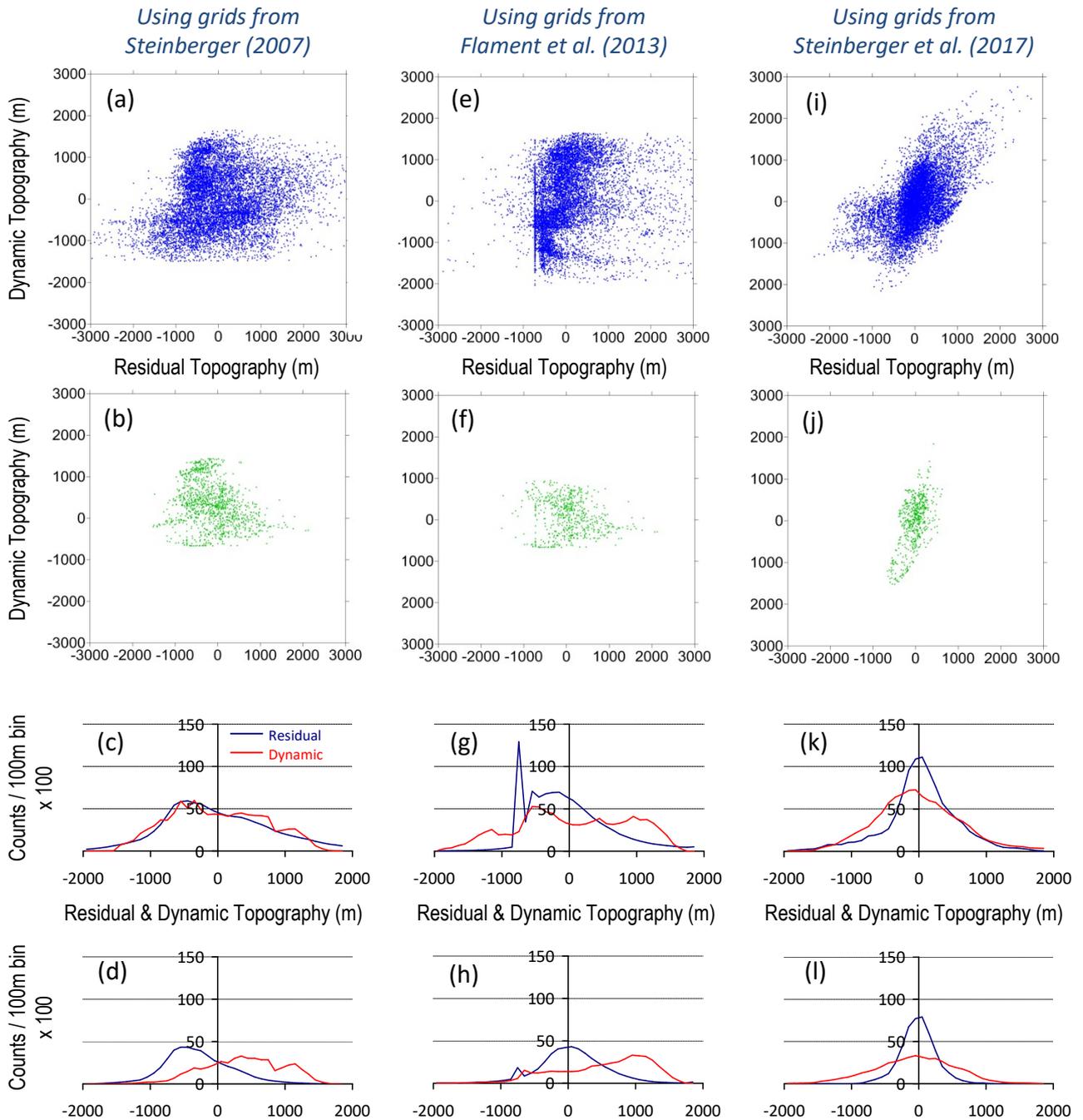


Figure 4

Using grids from Steinberger (2007)

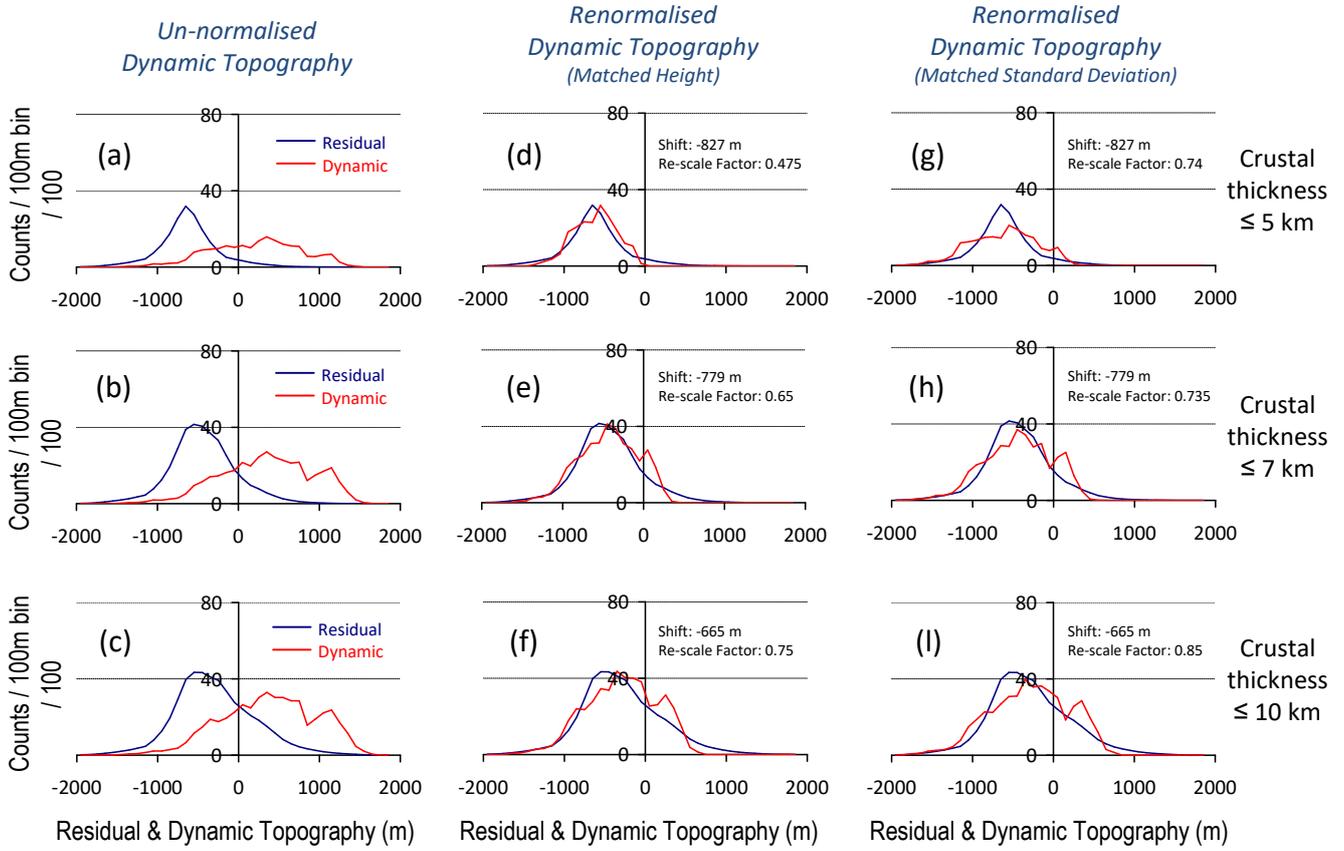


Figure 5

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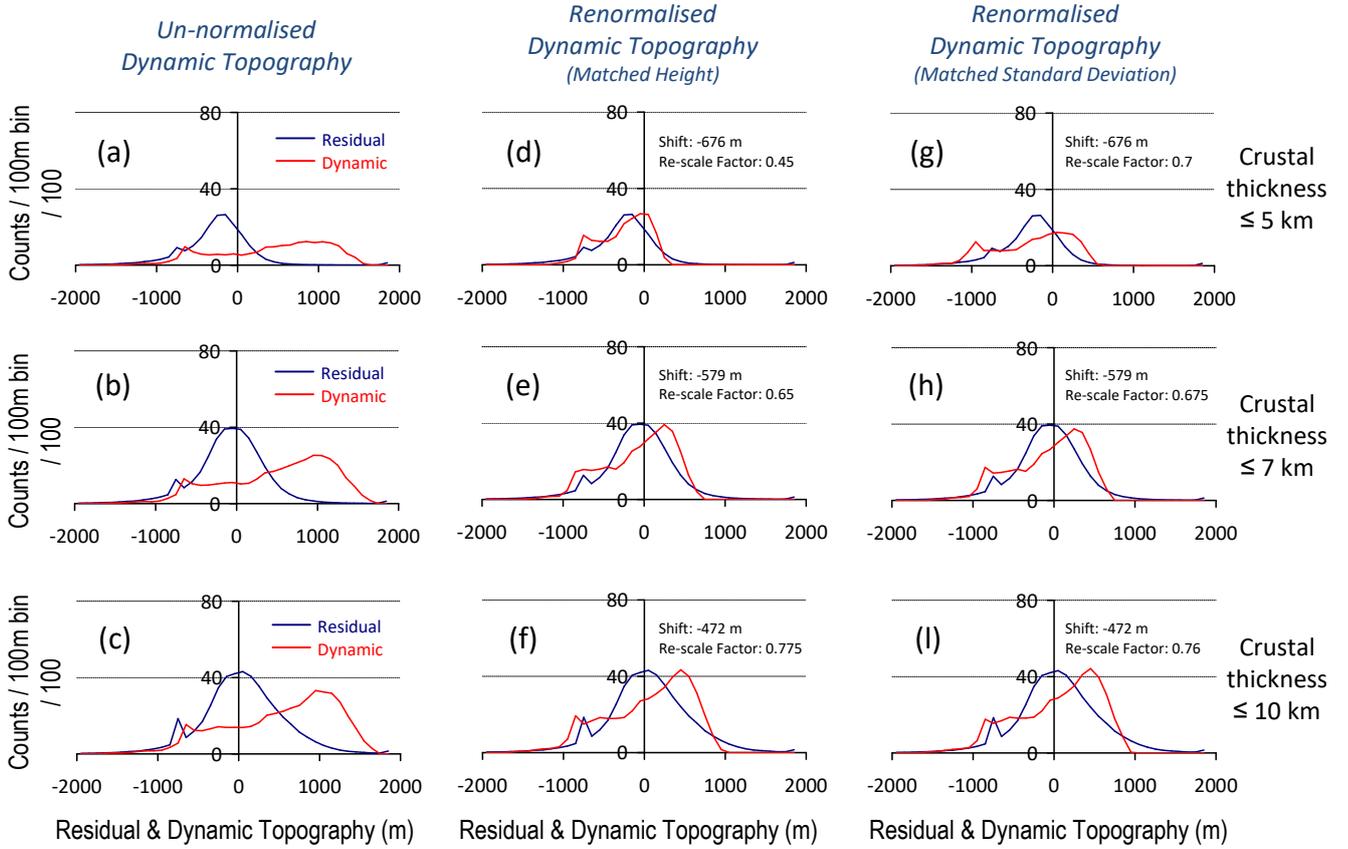


Figure 6

Using grids from Steinberger et al. (2017)

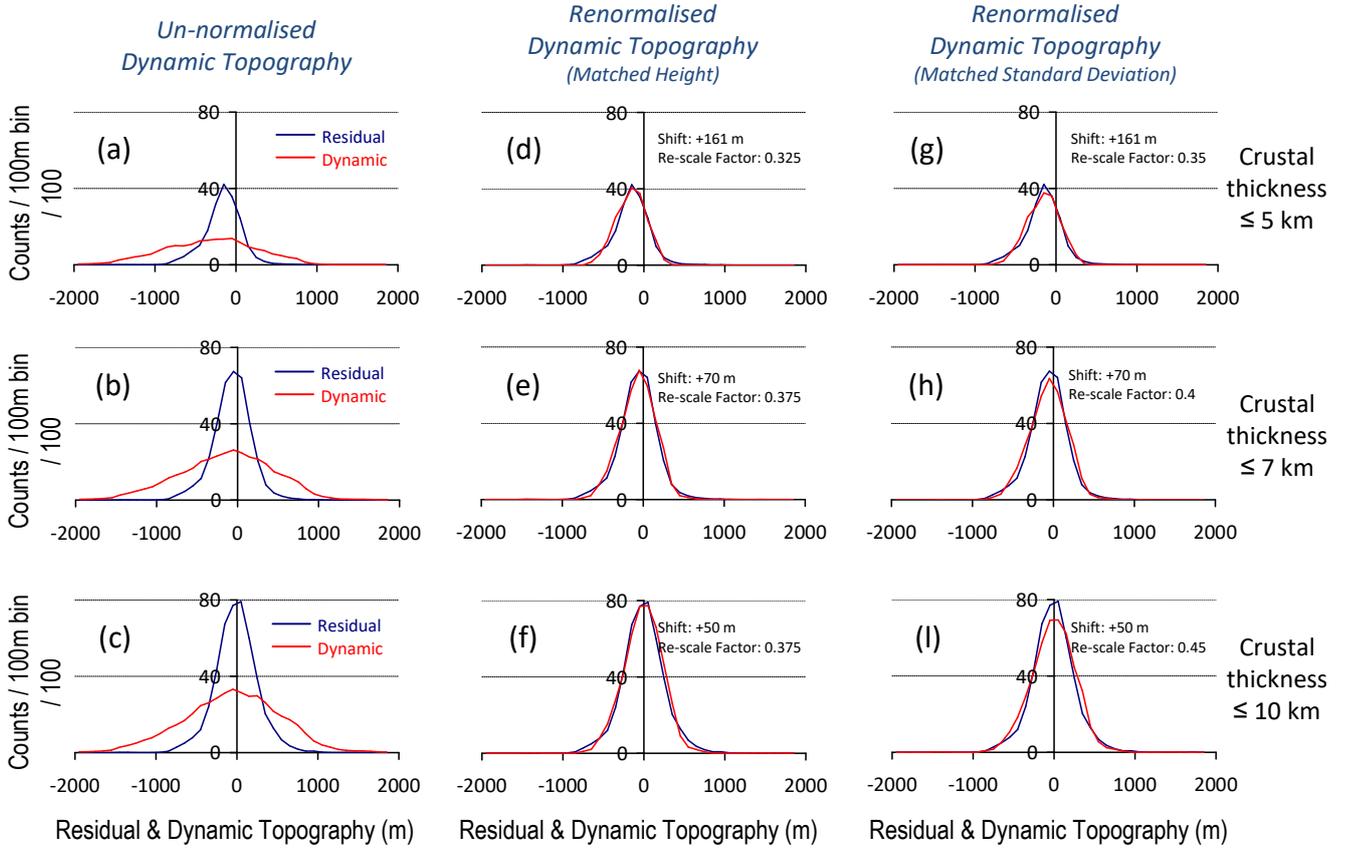


Figure 7

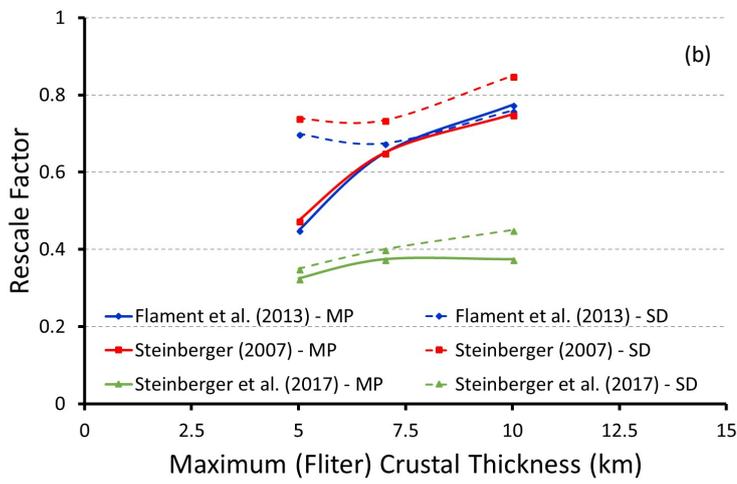
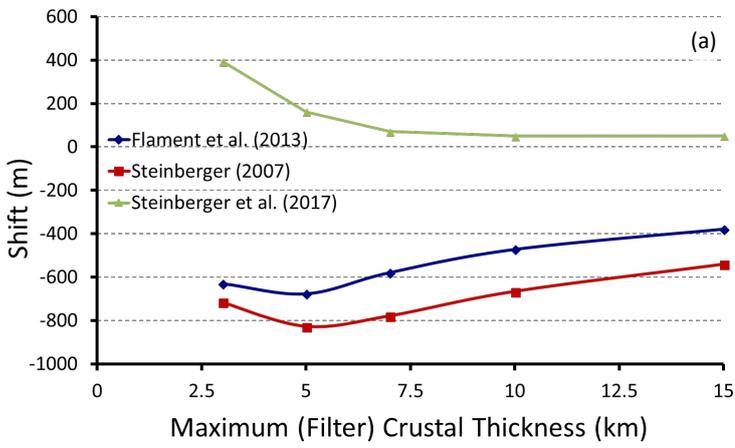


Figure 8

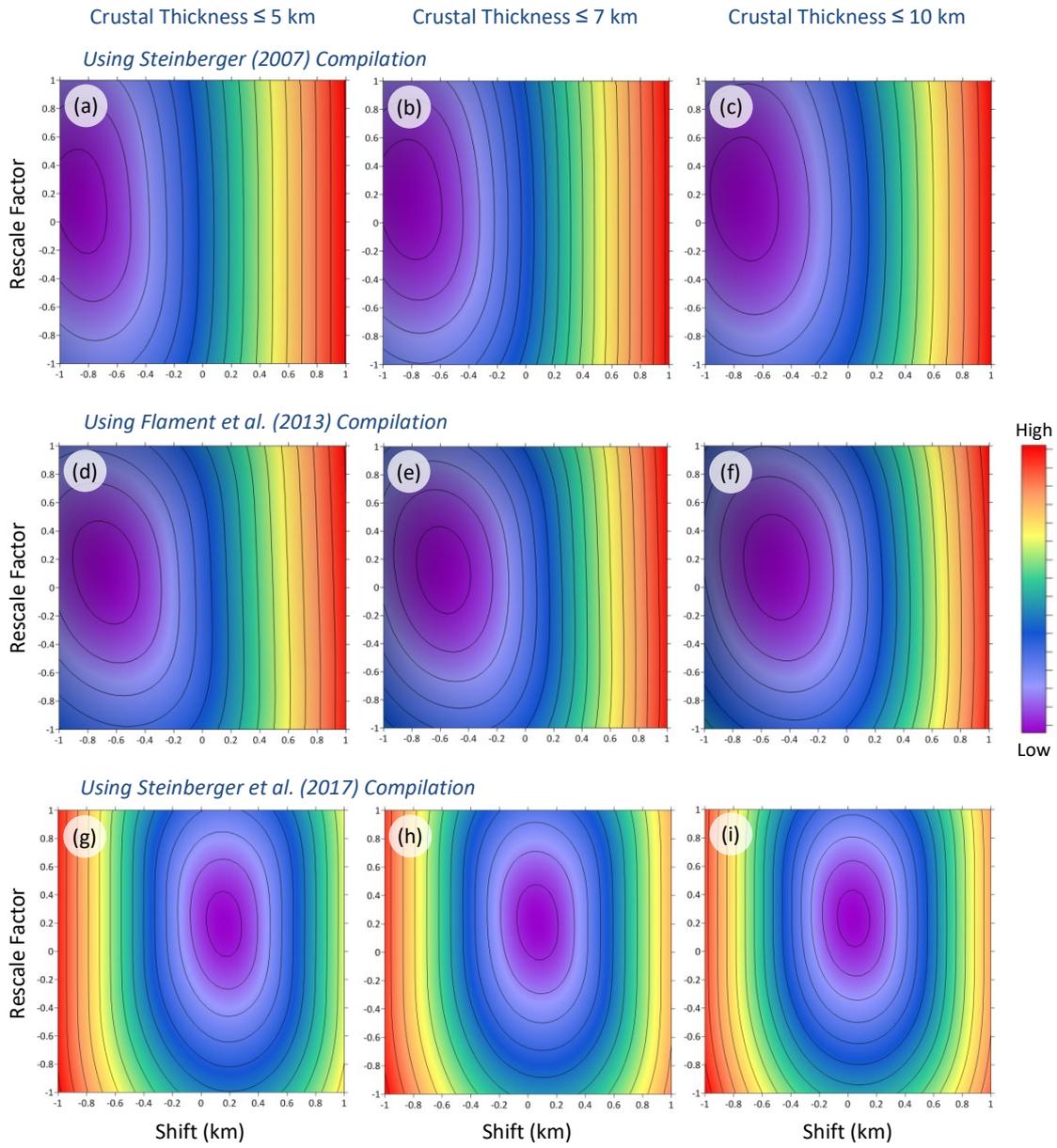
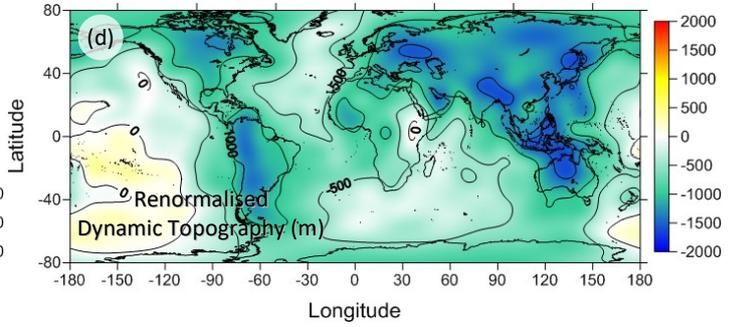
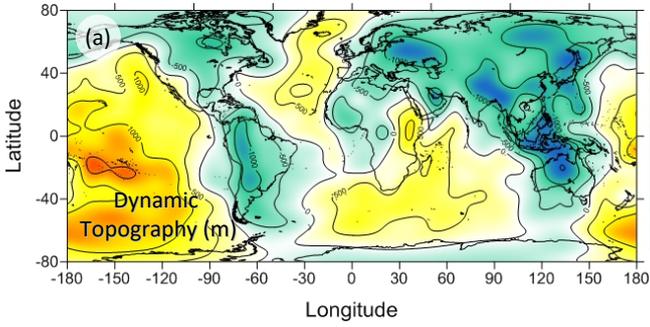
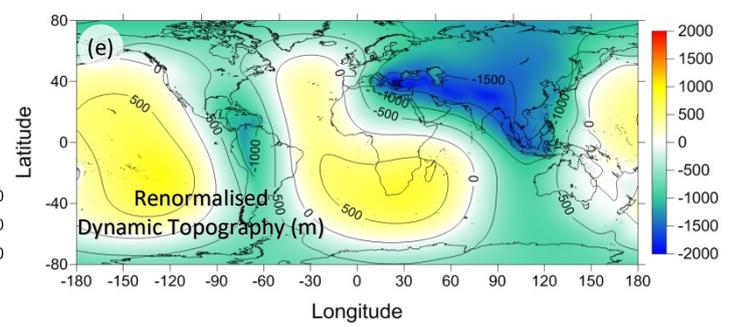
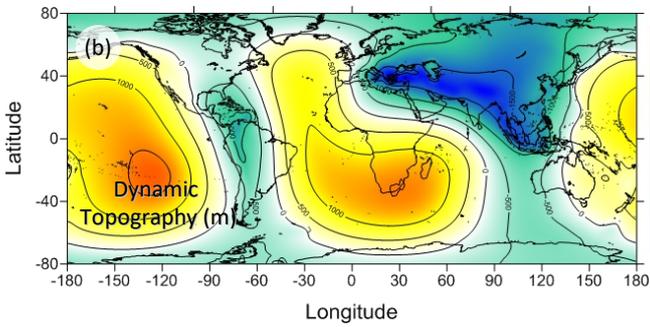


Figure 9

Using grids from Steinberger (2007)



Using grids from Flament et al. (2013)



Using grids from Steinberger et al. (2017)

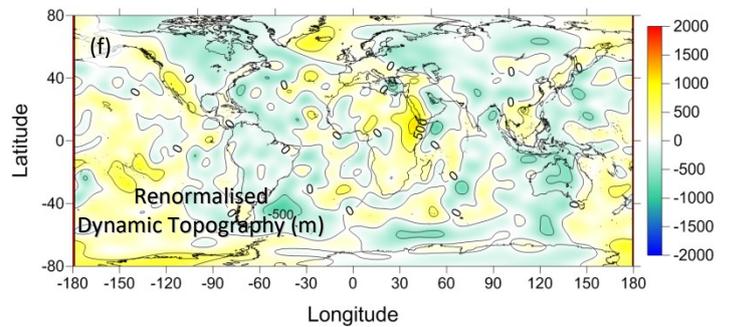
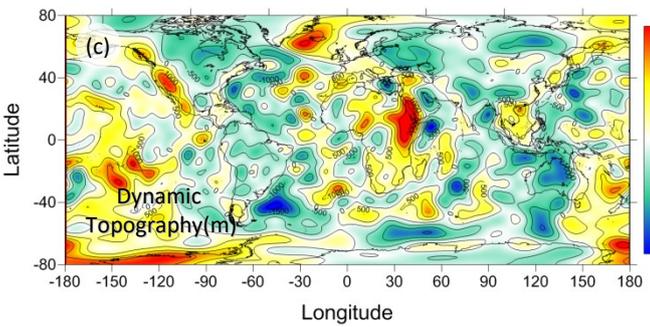


Figure 10

Cowie & Kuszniir, 2018

Using grids from Steinberger (2017)

