

Uncertainty of Satellite-gravity-derived Moho Estimates: Contribution of Data Reductions

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Background

- Global satellite-only gravity models provide unparalleled spatial homogeneity in coverage and quality, at length scales suitable for lithospheric density modelling.
- Geophysical inverse problems require isolating an anomalous signal in the observed gravity field, through removal of the effect of known masses (data reduction, e.g. topography, sediments ...)
- Error characteristics of gravity models: 3 orders of magnitude smaller than reduction uncertainty at the same length scales. Data reduction and inversion parameters are the main error sources.

Forward Modelling Algorithm

We rely on the SHTOOLS [1] implementation of **Wieczorek & Phillips (1998)** algorithm [2] spectral forward modelling algorithm for the potential of a relief with lateral variations of density, referenced to a spherical interface.

from Wieczorek (2007) [3]

Sph. harmonics coefficients of the relief (to the n-th power)

$$(ph^n)_{lm} = \frac{1}{4} \int_{\Omega} [\rho(\theta, \phi) h^n(\theta, \phi)] Y_{lm}(\theta, \phi) d\Omega \quad \text{with } \Omega = (\theta, \phi) \quad (1)$$

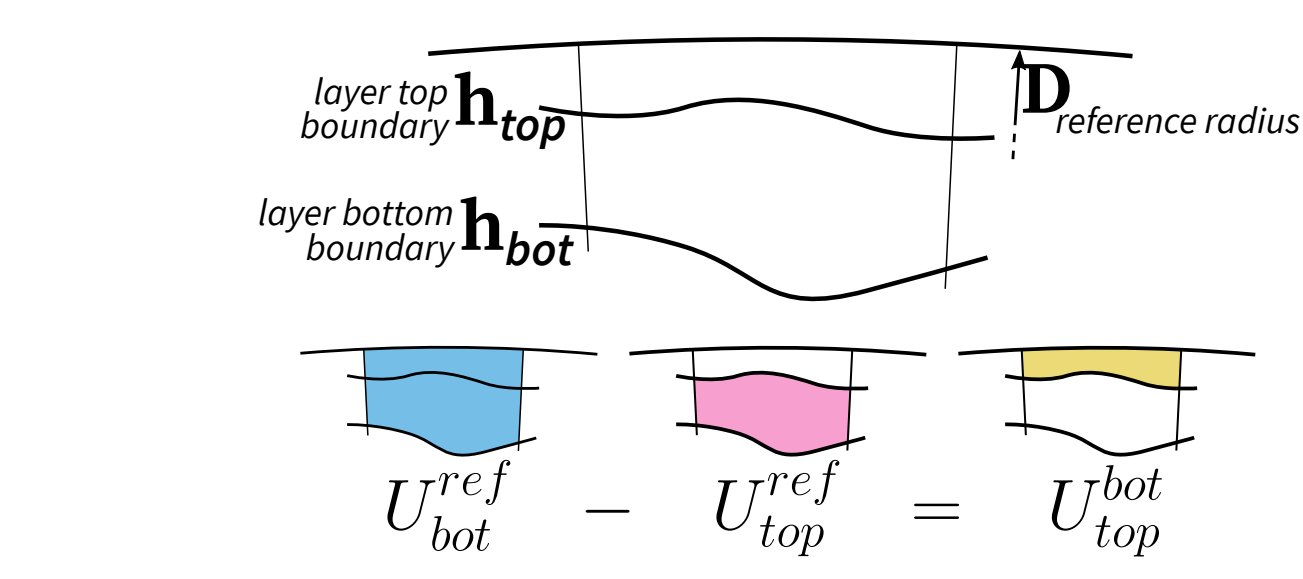
Sph. harmonics coefficients of potential

$$C_{lm} = \frac{4\pi D^3}{M(2l+1)} \sum_{n=1}^{n_{max}} \frac{(ph^n)_{lm}}{D^n n!} \frac{\Pi_{j=1}^{l+4-j}}{(l+3)} \quad (2)$$

Potential (exterior to relief)

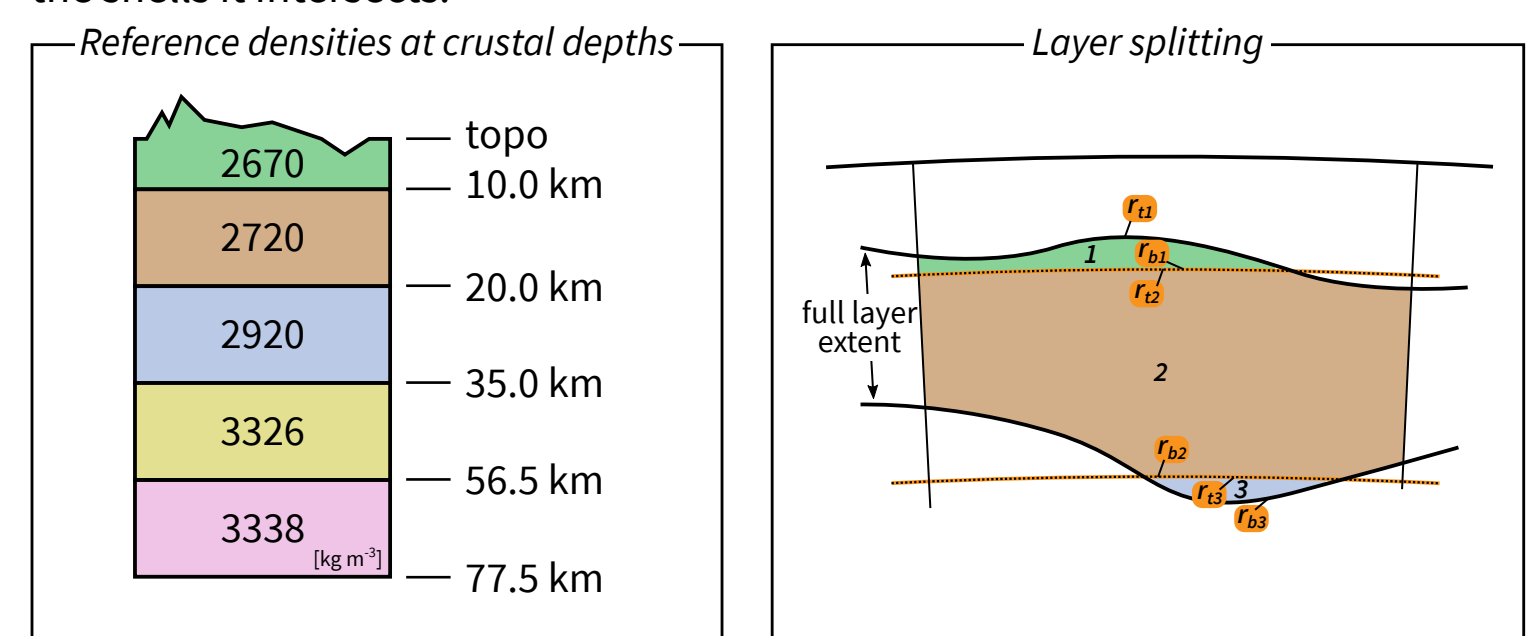
$$U(\mathbf{r}) = \frac{GM}{r} \sum_{l=0}^{\infty} \sum_{m=0}^l \left(\frac{R_0}{r}\right)^l C_{lm} Y_{lm}(\Omega) \quad (3)$$

We set up a layer-wise forward modelling scheme:



Density reference and layer splitting

Global density reference: adapted from **AK135**[4], discretized in geocentric ellipsoidal shells of constant density. The "known densities" of the modelled layers are expressed against this reference, after slicing each layer according to the shells it intersects.

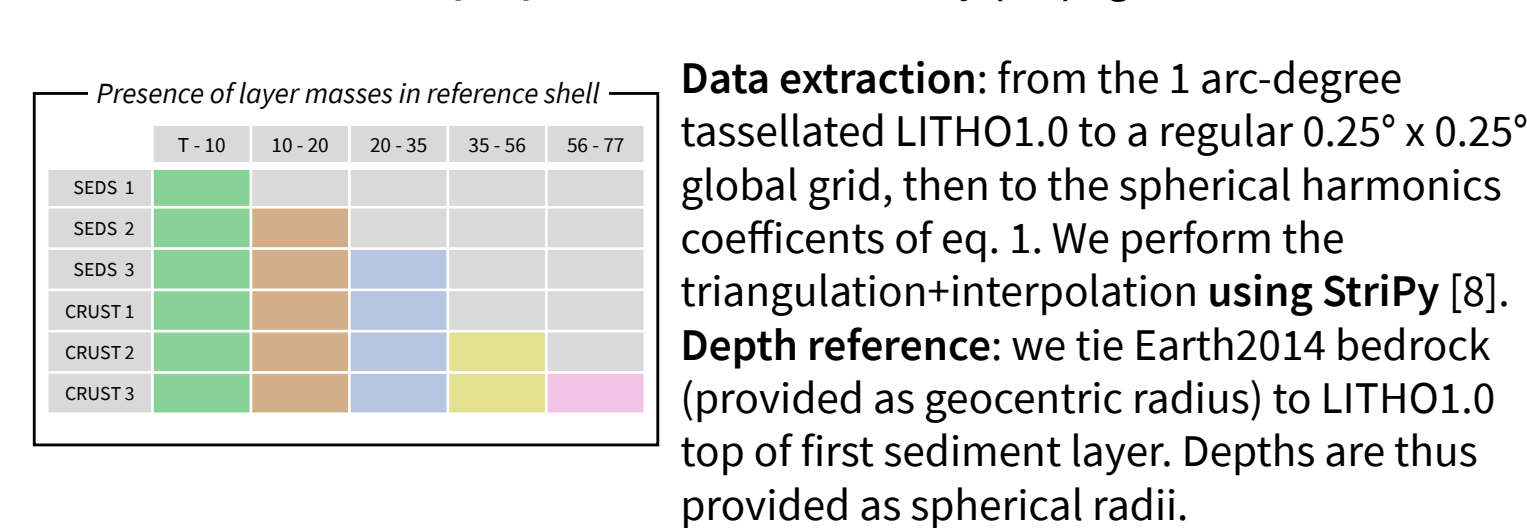


Terrain correction: input topography, water, ice

We use the **Earth2014**, 1 arc-min shape model [5] to obtain a terrain correction (TC). We forward modelled an **ellipsoid-referenced** solid topography effect, plus bathymetry and ice stripping. When this TC is removed from the observed gravity disturbance, we obtain "No Ellipsoidal Topography of Constant density" gravity disturbance (NETC, see [6]).

Sub-surface data: LITHO1.0 [7]

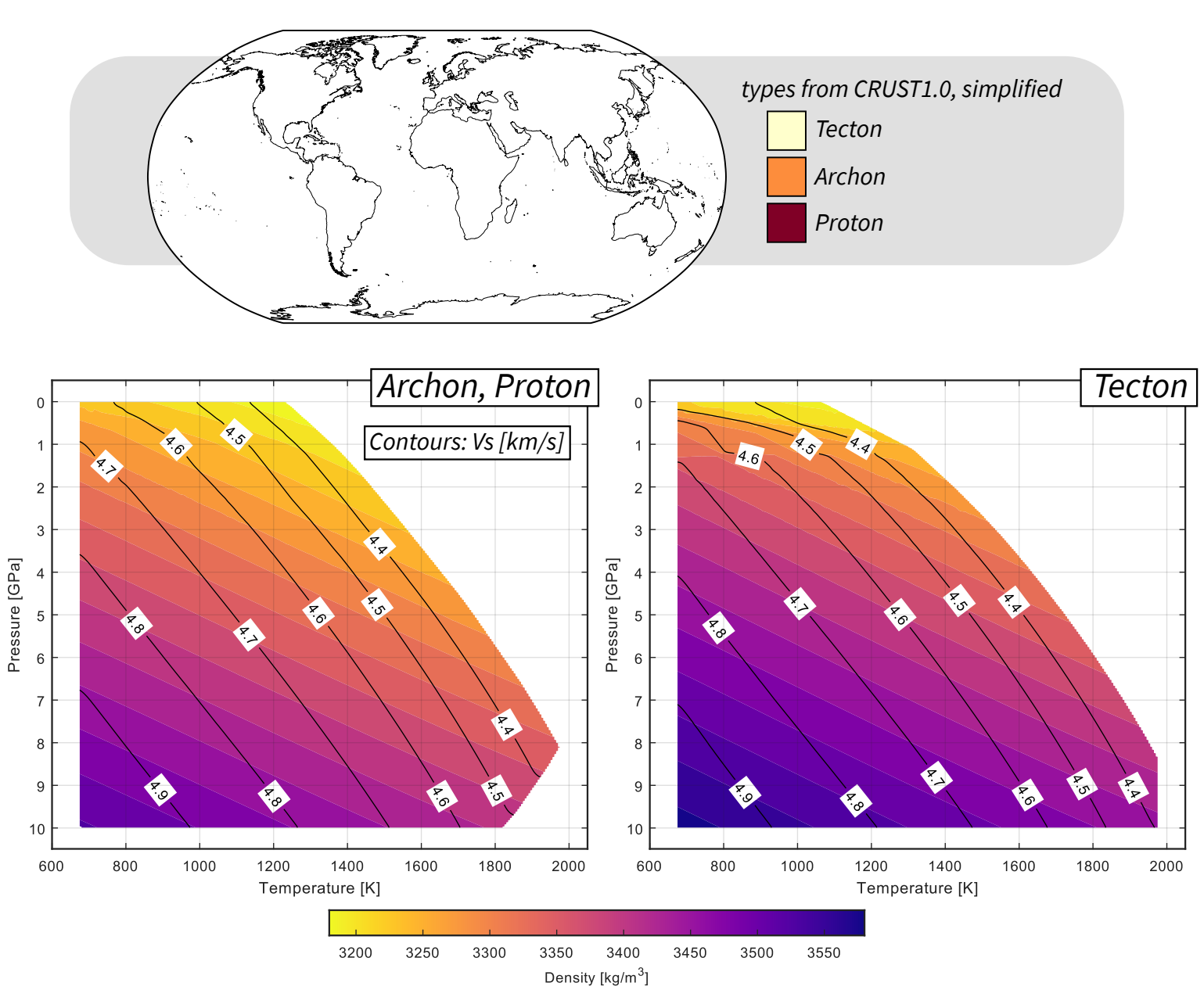
- Readily available, global depth-density model, layer defined: topography to lithosphere-asthenosphere boundary.
- Surface wave based, from an integrated starting model (multiple sources): no information on coverage and data uncertainty, this suggest caution.
- We consider it **fit-for-purpose** for this uncertainty-propagation test.



Data extraction: from the 1 arc-degree tassellated LITHO1.0 to a regular 0.25° x 0.25° global grid, then to the spherical harmonics coefficients of eq. 1. We perform the triangulation+interpolation using **Stripy** [8]. **Depth reference:** we tie Earth2014 bedrock (provided as geocentric radius) to LITHO1.0 top of first sediment layer. Depths are thus provided as spherical radii.

Lithospheric mantle: velocity-to-density conversion

- Vs to density for LITHO1.0 'LID' layer (Moho to LAB)
- density and Vs forward modelling using **Perple_X** [9]
- simple compositional model: Archon/Tecton, according to Griffin et al. 2009 [10]



Uncertainty propagation through random modelling

- error assumptions on the input data: **depth and density**
- random modelling on 5000 independent draws
- depth uncertainty, st. dev 5% of depth
- density uncertainty, st. dev 100 kg m⁻³
- simple error criteria (realistic, but no spatial variability)
- no error covariance information is included: each node assumed independent
- criteria for 5000 draws: high enough to observe power-law decay in error degree variances (i.e. effect of assuming uncorrelated errors attenuates)

Implementation:

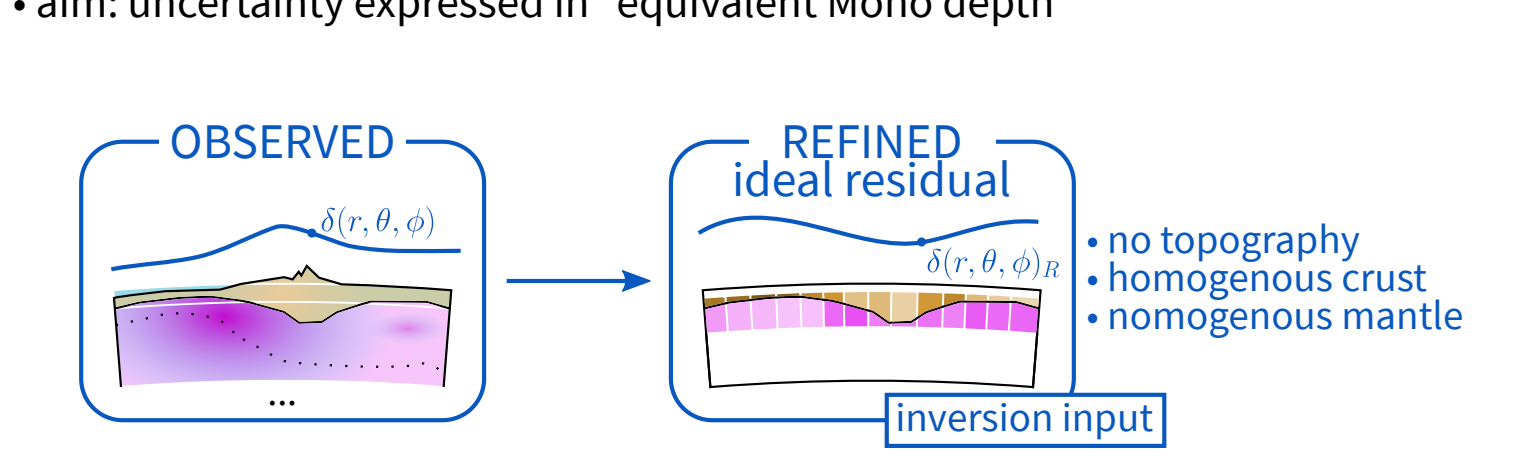
- parallel implementation, using the **multiprocessing** Python module [11]
- 3,3 seconds per sliced-layer, per worker, per draw (e.g. 2 hours on 40 workers)
- random draws are partitioned in 100-draws blocks
- the variance of g partitions of k draws is consolidated, using the following:

$$\text{Var}(X_1, \dots, X_n) = \frac{k-1}{gk-1} \left(\sum_{i=1}^k \frac{k(g-1)}{k-1} \text{Var}(E_i) \right) \quad \text{with } E_i, V_i \text{ mean and variance of each partition}$$

→ could be easily scaled to more complex, larger schemes

From disturbance-uncertainty to Moho-uncertainty

- aim: uncertainty expressed in "equivalent Moho depth"

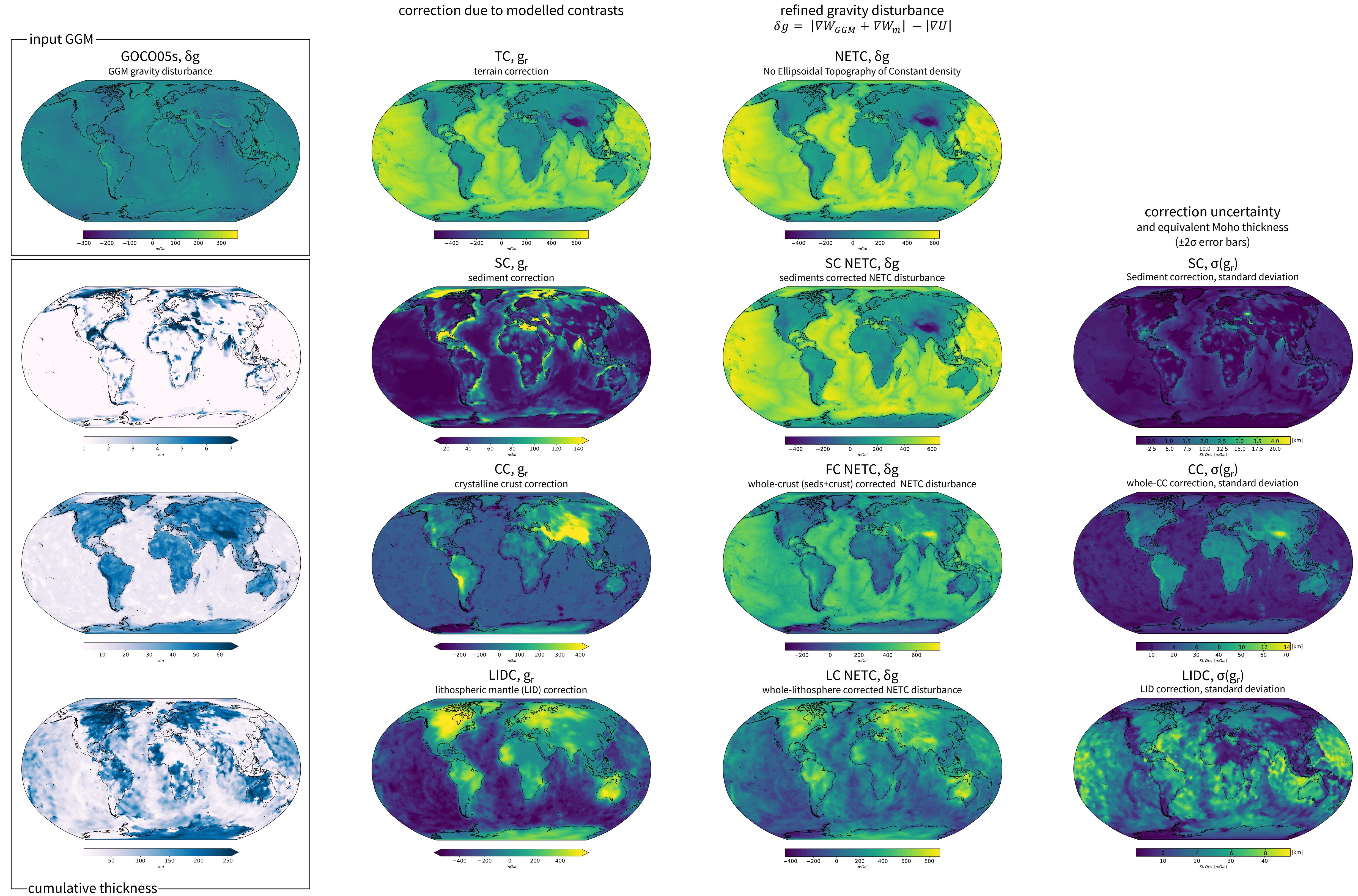


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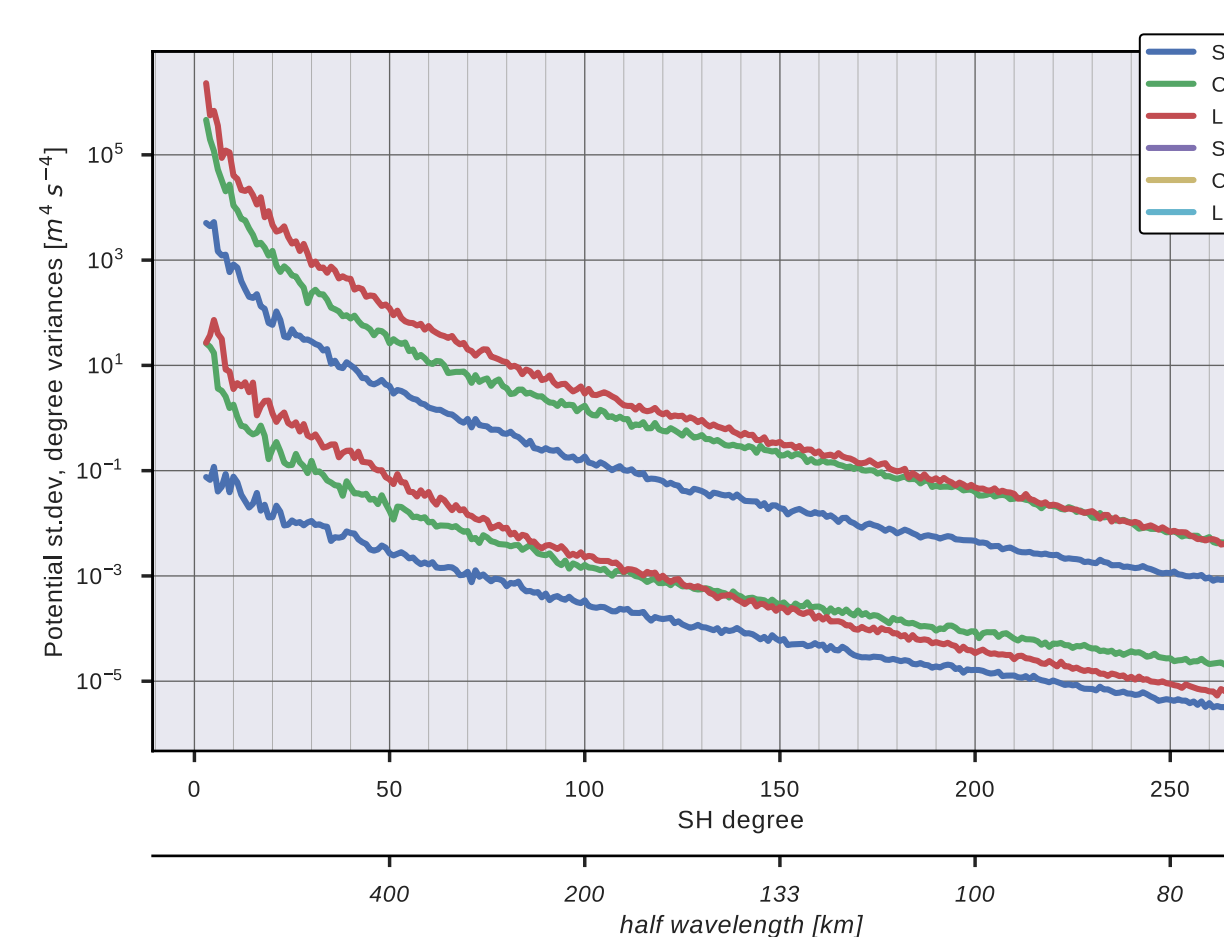
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Results: forward modelled reductions

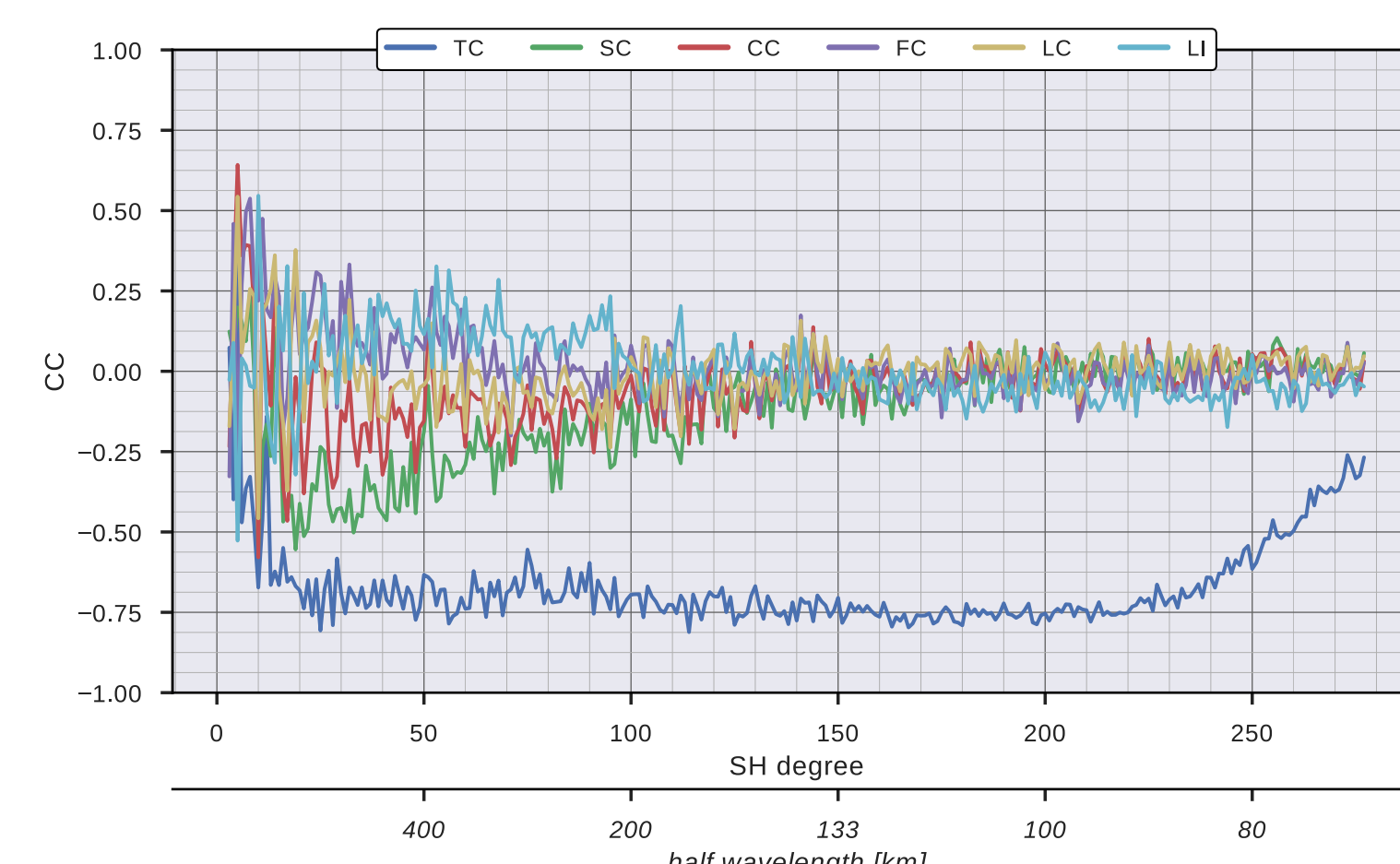
All functionals were computed at 10 km over GRS80, up to SH degree = 280



degree variances spectra: layer against error



correlation coefficients, against GGM



concluding remarks

- Outcome (and collaterals):**
- error estimate of gravity model "after reduction" relying on a layer-based, spectral domain forward modelling of reductions and propagated errors
 - reproduction of "common" and "novel" reductions from a topography-free to an (ideally) lithosphere-free disturbance (albeit with simple assumptions!)
- Room for improvement:**
- realistic, data dependent, error estimates e.g. weight according to data density, observable - and error propagation from conversion
 - upper mantle model and velocity conversion integrate available models, removal of lithosphere only shows LAB as artifact compositional model: refine or assess effect of "coarse" assumptions?
 - gravity-model-aware adaptation of reductions truncating at maximum SH degree is not enough e.g. take into account high-degree regularization of sat-only models