

## Towards Backus-Gilbert style inversions for mantle anisotropy using normal mode data

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## Why use the SOLA method?

Over the past decades, global tomographic models have exploited variants of the damped leastsquares (DLS) linear inversion approach. However, DLS-based models can have **amplitudes locally biased** towards lower or higher values, especially in regions of poor data illumination. This can potentially cause physical misinterpretations of tomographic images. Moreover, the computation of the full DLS-generalised inverse - required to infer both the model resolution and its covariance - is **prohibitive** for large-scale problems.

The SOLA (Subtractive Optimised Locally Averages) method (Zaroli, 2006)

- is based on a **Backus–Gilbert** (B-G) approach
- allows to produce quantitative **unbiased** tomographic images
- provides the estimated model together with the full appraisal associated to this model, composed of uncertainties and resolution
- retains all the advantages of the B-G approach but is computationally more efficient and versatile.



## DLS vs. SOLA (1)

#### What are we looking for?

DLS

Estimates of the (damped) model parameters themselves in specific locations

SOLA A linear combination of the true model parameters  $\rightarrow$  their average with uncertainties

#### How do we remove the non-uniqueness of the solution?

#### DLS

Introducing *ad hoc,* subjective regularisation constraints on the model itself (norm damping or smoothing) SOLA

Through the process of averaging: a unique combination of the parameter can be found even if the parameters are not uniquely defined

Amplitudes can be biased

Toy problem: in regions with isolated receivers (bottom right corner), DLS-amplitudes are biased (dark blue spot in the red circle, not present in the input model or in the SOLA solution). From Zaroli et al., 2017



#### Amplitudes are constrained to be unbiased









Zaroli, C. et al. (2017). Geophys. Res. Lett.

## DLS vs. SOLA (2)

#### How do we formulate the minimisation problem?

DLS Aims to minimise the data misfit and the model complexity

#### SOLA

Aims to minimise the difference between target and resolving kernels (linked to the model resolution) and model uncertainties

**Key idea of SOLA**: we specify an *a priori* target form for each resolving kernel. In this way, we add *a priori* information on the model resolution, but not on the model itself!

#### Do we always have info about uncertainties and resolution?

DLS

**NO**, the full generalised inverse is often not computed for large-scale problems

**SOLA YES**, it is inherent to the method



## Why study seismic anisotropy?

Seismic anisotropy can provide us information about mantle flow, and therefore the thermal state, composition and mechanical properties of the mantle.



While isotropic models are consistent at least on long wavelengths (as evidenced in the vote maps above), agreement on anisotropy is still lacking, particularly in the deep mantle.



The lack of consensus and uncertainty information makes the interpretation of seismic anisotropy difficult!

### What are normal modes and why use them?

Normal modes are the long-period (~3000-100 s) standing waves of the Earth observed after very large earthquakes ( $M_w \ge 7.4$ ). They can be observed as peaks at discrete frequencies of a Fourier transformed seismogram.



Normal modes have been chosen because they provide **global coverage** and are directly **sensitive to**  $V_s$  **and**  $V_p$ **anisotropy**, whose interpretation provides information about mantle flow.





Park, J., et al. (2005). Science Lockwood, O. G., & Kanamori, H. (2006). Geochem, Geophys, Geosys

### SOLA and normal modes: 3D noise

- Normal modes are sensitive to multiple parameters simultaneously
- With SOLA, we treat the additional sensitivity of the parameters not of interest as noise called *3D* noise (Masters et al., 2003), instead of including scaled sensitivity kernels
- We can make an estimation of the 3D noise by computing the splitting function coefficients from models with only VP, Vs,  $\rho$  or topography at internal boundaries perturbations
- For example, the total 3D noise when resolving for e.g. Vs will be given by the contribution from VP,  $\rho$  and topography
- These uncertainties are added to the data uncertainties in the inversion
- The 3D noise characterisation is vital, since the inversion is largely driven by errors.



## First inversions set-up

For simplicity, we start by inverting for isotropic velocities.

The SOLA method requires the user to define:

- Sensitivity matrix, which contains the sensitivity kernels of the normal modes

   → sensitivity kernels discretised in 52 layers
- Data vector, which contains the splitting function coefficients and their uncertainties (and, in our case, the 3D noise)
  - → same dataset as SP12RTS (*Koelemeijer at al., 2016:* 143 spheroidal normal modes)
- A vector that contains the thickness of the layers of the model parameterisation
  - $\rightarrow$  model space parameterised in 52 layers
- Target kernels
  - $\rightarrow$  52 wide and overlapping target kernels



We first invert for Vs, as it has the largest amplitudes in the Earth.

### Isotropic inversions set-up (1) Sensitivity kernels





Example of two continuous Vs sensitivity kernels (blue) and Vs sensitivity kernels discretised in 52 layers (red) for modes 19S11 (left) and 2S14 (right).

#### Isotropic inversions set-up (2) 3D noise 1000 800 600 VP 400 200 0 Mean 5 - 55-5 5 - 5s=2, t=0 s=2, t=1 s=2, t=-1 s=2, t=2 s=2, t=-2 SD 1000 800 600 ρ 400 200 0-2 2-2 2-2 2-2 2 - 20 n 0 s=2, t=0 s=2, t=1 s=2. t=-1 s=2. t=2 s=2. t=-2

Quantification of 3D noise for VP and  $\rho$  computed using the splitting function predictions for 16 existing models, each time only including VP or  $\rho$ .

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Assembly For each degree *s* and order *t* (*s*=2 shown in the figure above), we evaluate the standard deviation (SD) of the distribution, imposing the mean to be zero.

#### Isotropic inversions set-up (2) 3D noise





# Isotropic inversions set-up (3) **Target kernels**

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Similar to *Masters et al. (2003),* we aim to obtain resolving kernels in the shape of a boxcar.

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### Isotropic inversions preliminary results Resolving kernels and uncertainties

The resolving kernels (i.e. resolution) are unaffected by the addition of 3D noise but uncertainties increase substantially.

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Target kernels
---- Resolving kernels

- With 3D noise
- Without 3D noise

## Summary

- Anisotropy can provide us with valuable information about the mantle flow, but anisotropic tomography models are not consistent → difficult interpretation!
- The SOLA method allows us to obtain unbiased tomography images, provided with uncertainty and resolution information
- We aim to develop a new anisotropy model based on normal mode data using SOLA. Normal modes are sensitive to VS and VP anisotropy, and provide global coverage
- 3D noise characterisation is a vital step: errors largely drive the inversions
- Our preliminary results are able to produce resolving kernels similar to the target kernels, with 3D noise having a big effect on model uncertainties and not on model resolution.

