

## 1 Overview

- Global radial anisotropic models are inconsistent (Fig. 1) meaning geological interpretation of active/ancient mantle flow/deformation is challenging.
- One recent interpretation (Priestley et al., 2020) suggests that negative radial anisotropy in the CAM2016 model at ~150km depth within cratons reflects their formation by horizontal shortening/vertical thickening. This mechanism is not easily reconciled with other available models however.
- Using variable parameterizations within both LSQR and Bayesian inversions of Rayleigh and Love (R&L) surface wave dispersion curves, we test whether negative radial anisotropy is reliably recovered at upper mantle depths using synthetic models (Section 3) and whether anisotropic anomalies are required below cratons using real data inversions (Section 4).
- Both algorithmic and parameterization choices can affect 'recovered' radial anisotropy (Xi) meaning existing geological interpretations may be biased.
- Benchmarking against 'active' mid-ocean ridge, hotspot & active mountain belt locations (Fig. 2) available on request.

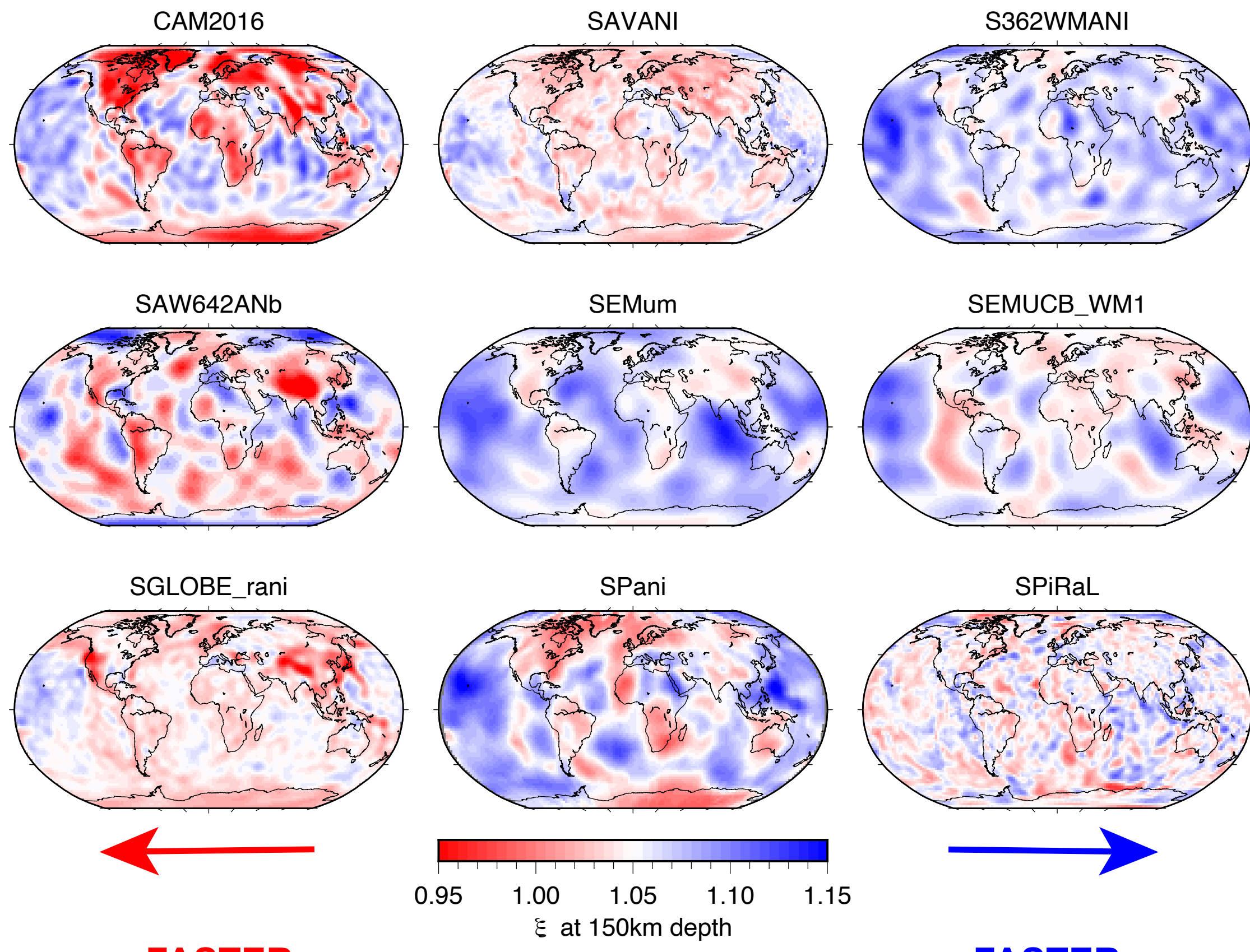


FIGURE 1: Nine published global radial anisotropy models (see references) at 150km depth plotted on a diverging color scale around  $\xi=1.05$ , the approximate  $\xi$  in PREM at 150km depth.

## 2 Methodologies

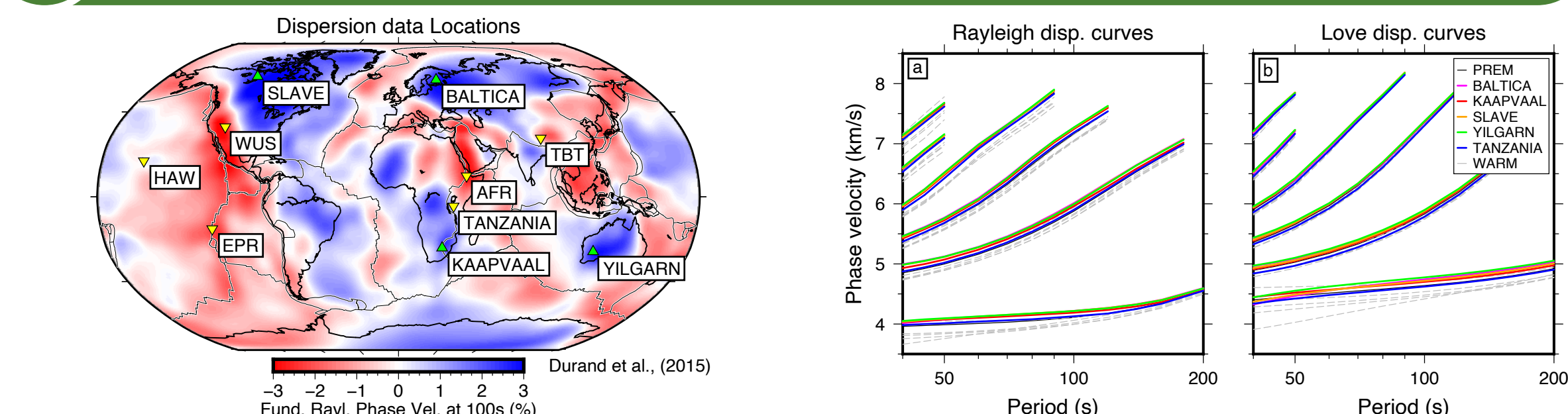


FIGURE 2: Fundamental mode Rayleigh wave phase velocity map at 100s period (Durand et al., 2015) w.r.t. mean value. Locations of cratonic data inversions in Section 4 (green triangles) and 'tectonically active' region data for benchmarking (yellow triangles) are indicated.

- Reference model for all inversions: Modified PREM without 220km disc. & upper mantle radial anisotropy, 35km depth Moho.
- Sensitivity kernels (LSQR) & R&L dispersion curve forward modeling (Bayesian & synthetic models): Mineos (Masters et al., 2011).
- 2D LSQR inversion based on Tarantola & Valette (1982) adapted from Debayle & Ricard (2012). Conservative regularization parameters chosen on L-curve.
- 2D hierarchical transdimensional Bayesian inversion to 700km depth using reversible jump Markov chain Monte Carlo sampling (adapted from Bodin et al., 2016). Number of layers, presence/absence of anisotropy & data error are all free parameters. Prior: +/-20% for  $V_{SV}$ , +/-40% for  $\xi$  &  $V_{PH}$ .
- Fundamental & higher mode (up to 5th overtone) R&L dispersion curves at 40-200s period extracted from phase velocity maps (Durand et al., 2015) at 4 cratonic locations (Figs. 2&3) for inversion in Section 4.

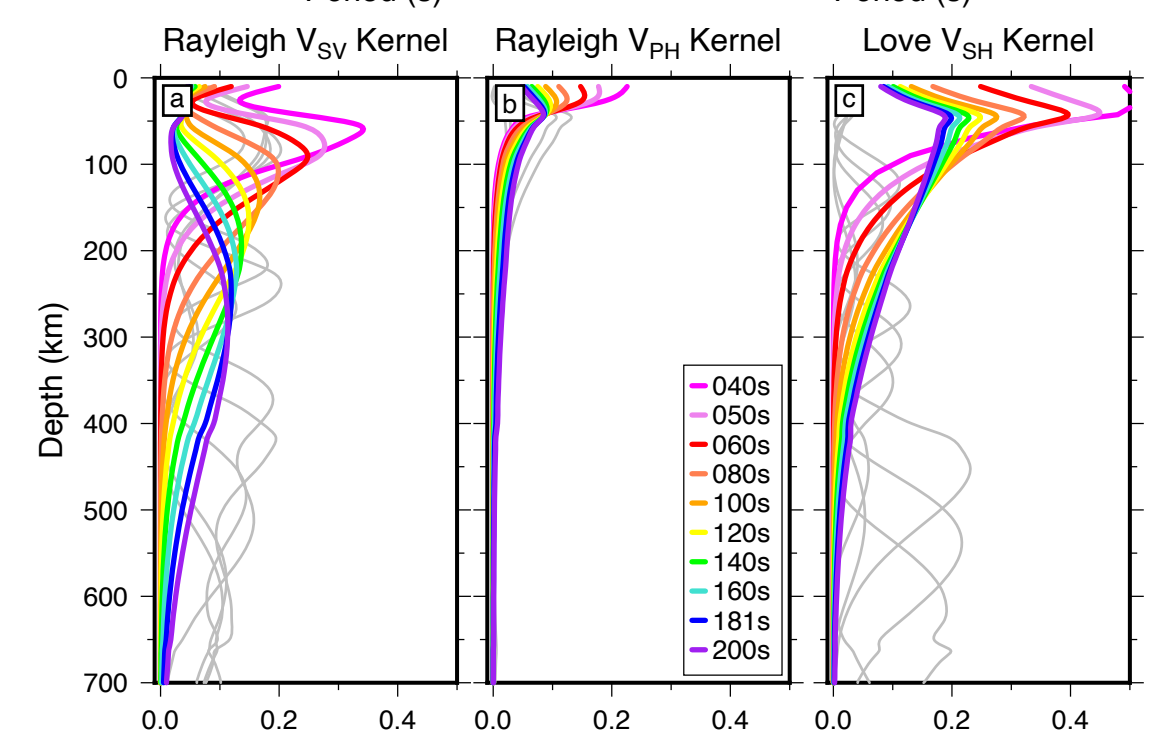
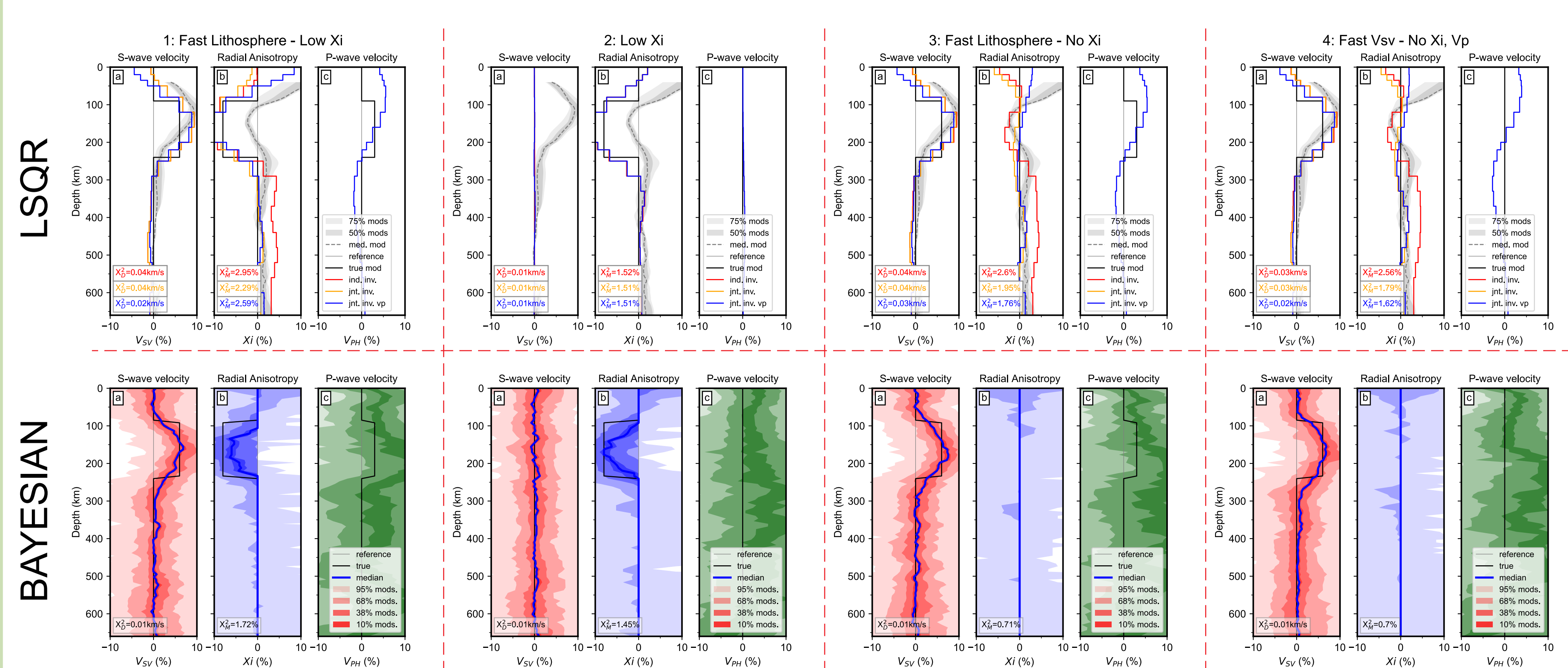


FIGURE 3 (upper): Rayleigh & Love (a,b) fundamental mode and overtone (1-5) dispersion curve data extracted from phase velocity maps (Durand et al., 2015) at 40-200s period at four cratonic locations (colored see Section 4) and 'tectonically active' locations (grey). FIGURE 3 (lower): Fundamental mode (colored) and 3rd overtone (grey) sensitivity kernels for Rayleigh ( $V_{SV}$  &  $V_{PH}$ , a,b) and Love ( $V_{SH}$ , c) waves at a range of periods for modified PREM reference model.

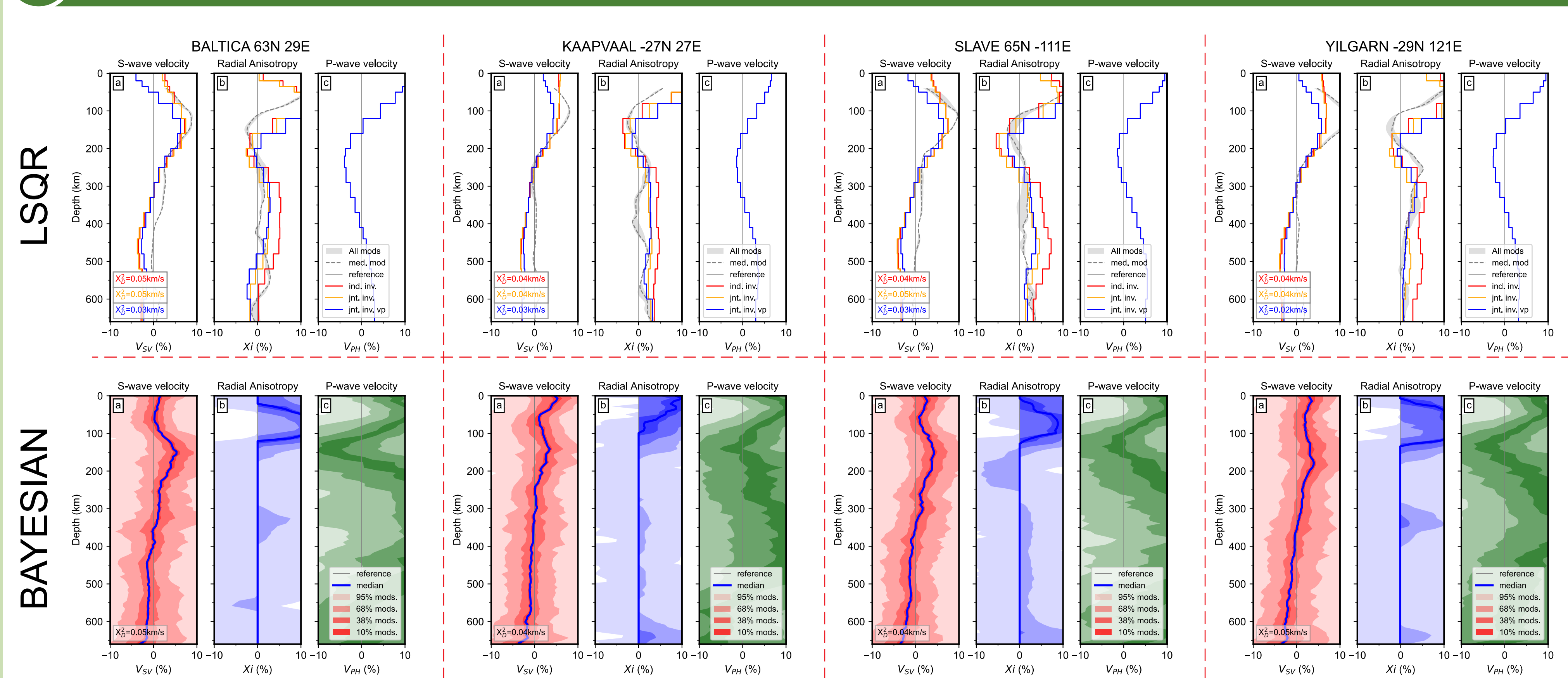
## 3 Synthetic Data Inversions



FIGURES 5-8 (Above, Upper): Four synthetic models inverted using variably parameterized LSQR algorithm.  $V_{SV}$ ,  $\xi$  &  $V_{PH}$  shown in percent deviation from reference model (no anisotropy). Red curve: Independent inversion for  $V_{SV}$  &  $V_{SH}$ . Orange curve: Joint inversion for  $V_{SV}$  &  $V_{SH}$ . Blue curve: Joint inversion for  $V_{SV}$ ,  $V_{SH}$  &  $V_{PH}$ . Chi squared data ( $X^2_{\chi}$ ) and model ( $X^2_{\text{mod}}$  for  $V_{SV}$  &  $\xi$ ) fits shown. Grey line: Reference model. Black line: True model. Grey shaded regions and dashed curve (med. mod) show distribution of  $V_{SV}$  &  $\xi$  profiles from CAM2016 model (Priestley et al., 2020) extracted at cratonic locations explored in Section 4.

FIGURES 9-12 (Above, Lower): Synthetic models inverted using Bayesian algorithm. Posterior distribution (credible intervals) of  $V_{SV}$ ,  $\xi$  &  $V_{PH}$  shown in percent deviation from reference model (no anisotropy). Blue curve: Median model. Chi squared data ( $X^2_{\chi}$ ) and model ( $X^2_{\text{mod}}$  for  $V_{SV}$  &  $\xi$ ) fits shown. Grey line: Reference model. Black line: True model.

## 4 Real Data Inversions



FIGURES 13-16 (Above, Upper): Four cratonic R&L dispersion curves (Fig. 3, Durand et al., 2015) inverted using variably parameterized LSQR algorithm.  $V_{SV}$ ,  $\xi$  &  $V_{PH}$  shown in percent deviation from reference model (no anisotropy). Red curve: Independent inversion for  $V_{SV}$  &  $V_{SH}$ . Orange curve: Joint inversion for  $V_{SV}$  &  $V_{SH}$ . Blue curve: Joint inversion for  $V_{SV}$ ,  $V_{SH}$  &  $V_{PH}$ . Chi squared data ( $X^2_{\chi}$ ) fits shown. Grey line: Reference model. Grey shaded regions and dashed curve (med. mod)  $V_{SV}$  &  $\xi$  profiles from CAM2016 model (Priestley et al., 2020) extracted at each location.

FIGURES 17-20 (Above, Lower): Cratonic R&L dispersion curves (Fig. 3, Durand et al., 2015) inverted using Bayesian algorithm. Posterior distribution (credible intervals) of  $V_{SV}$ ,  $\xi$  &  $V_{PH}$  shown in percent deviation (+/-10%) from reference model (no anisotropy). Blue curve: Median model. Chi squared data ( $X^2_{\chi}$ ) fits shown. Grey line: Reference model.

## 5 Methodological Verification

- Synthetic 'Complex Crustal' input model inverted in Bayesian algorithm (Fig. 21).
- Data set (40-200s period) has low sensitivity to crustal structure, and we invert to the surface, meaning artefacts are not mapped into  $V_{SV}$  &  $\xi$  in the mantle (Fig. 21).
- Further testing shows inaccurate Moho depth (>20km) in reference model produces artefacts in  $V_{SV}$  &  $\xi$  in the mantle (e.g., Meier et al., 2007; Bozdağ & Trampert, 2008; Chang and Ferreira, 2017).

FIGURE 21 (Above): Synthetic complex crustal model inverted using Bayesian algorithm. Posterior distribution (credible intervals) of  $V_{SV}$ ,  $\xi$  &  $V_{PH}$  shown in percent deviation from reference model (no anisotropy). Blue curve: Median model. Chi squared data ( $X^2_{\chi}$ ) and model ( $X^2_{\text{mod}}$  for  $V_{SV}$  &  $\xi$ ) fits shown. Grey line: Reference model. Black line: True model. Horizontal dashed grey line: 35km depth Moho in true and reference models.

- Posterior sensitivity to scaling of P-wave radial anisotropy,  $\Phi$ , to S-wave radial anisotropy,  $\xi$  ( $\Phi = 1/\xi$ ) (Figs. 22 & 23).
- Minor differences in  $V_{SV}$  profile above 300km depth, location of peak anomaly amplitude (~150km depth) is consistent between parameterizations.
- Depth extent of shallow lithospheric positive  $\xi$  anomaly consistent but peak amplitude at ~100km depth reduced from  $\xi > 10\%$  (Fig. 22) to  $\xi > 5\%$  (Fig. 23).
- Parameterization choices not likely precluding visibility of negative radial anisotropy.

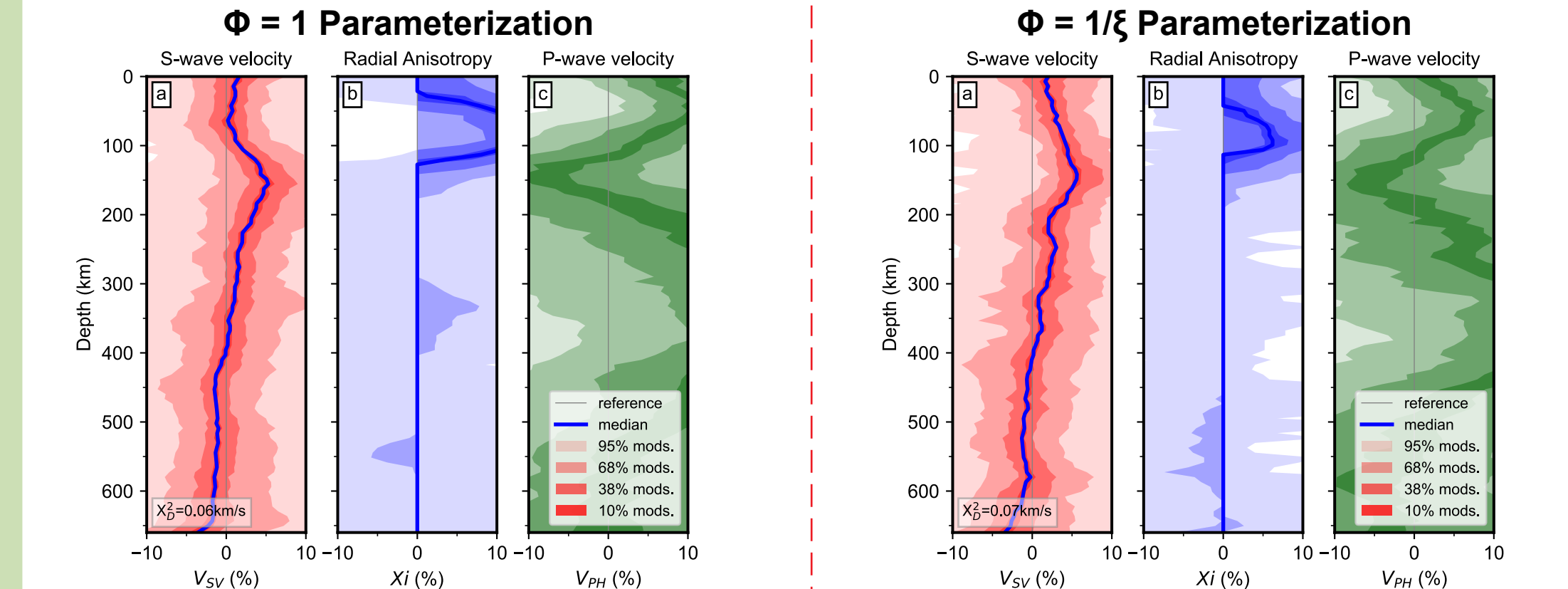


FIGURE 22 (Above, Left): As Fig. 17, Baltica cratonic R&L dispersion curves inverted using Bayesian algorithm. P-wave radial anisotropy parameter fixed:  $\Phi = 1$ . Posterior contains 25% of the number of samples in Fig. 17. Posterior distribution (credible intervals) of  $V_{SV}$ ,  $\xi$  &  $V_{PH}$  shown in percent deviation (+/-10%) from reference model (no anisotropy). Blue curve: Median model. Chi squared data ( $X^2_{\chi}$ ) fits shown. Grey line: Reference model.

FIGURE 23 (Above, Right): Same as Fig. 22 but with scaled P-wave radial anisotropy parameter:  $\Phi = 1/\xi$ .

## TAKE-HOME MESSAGES

- Negative  $\xi$  below cratons (~150km depth), reproduced using LSQR inversion, is likely an artefact...
- Bayesian inversion with free  $V_{PH}$  parameterization yields no negative radial anisotropy below cratons, only +5-10%  $\xi$  above 150km depth, similar to PREM.
- Episodic craton formation (e.g., Yuan & Romanowicz 2010; Darbyshire et al., 2013) preferred over horizontal shortening (Priestley et al., 2020).
- Set  $V_{PH}$  free in surface wave inversions, preferably with Bayesian algorithms!

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