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We present an original manuscript entitled “Joint probabilistic inversion of 3D magnetotelluric and seismic data: Lithospheric structure and melting dynamics in southeast Australia” by M.C. Manassero¹, S. Özaydin¹, J. C. Afonso¹,², J. Shea¹,³, S. Thiel⁴,⁵, A. Kirkby⁶, I. Fomin¹ and K. Czarnota⁷.

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Yours Sincerely,

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Joint probabilistic inversion of 3D magnetotelluric and seismic data: Lithospheric structure and melting dynamics in southeast Australia

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Key Points:

\begin{itemize}
  \item We apply a novel approach for joint probabilistic inversions of 3D magnetotelluric and seismic data.
  \item We use the new method to image the lithosphere-asthenosphere system beneath southeastern Australia.
  \item The imaged lithosphere correlates with the location of volcanic centers and provides insights on the melt production in the region.
\end{itemize}

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Abstract

The thermochemical structure of the lithosphere exerts control on melting mechanisms in the mantle as well as the location of volcanic eruptions and ore deposits. Imaging the complex interactions between the lithosphere and asthenospheric mantle require the joint inversion of multiple data sets and their uncertainties. In particular, the combination of temperature and electrical conductivity with data proxies for bulk composition and elusive minor phases is a crucial step towards fully understanding large-scale lithospheric structure and melting. We apply a novel probabilistic approach for joint inversions of 3D magnetotelluric and seismic data to image the lithosphere beneath southeast Australia. Results show a highly heterogeneous lithospheric structure that correlates with the location of Cenozoic volcanism and deep conductivity anomalies. In regions where the conductivities have been at odds with sub-lithospheric temperatures and seismic velocities, we observe that the joint inversion provides conductivity values consistent with other observations. The results reveal a strong relationship between metasomatized regions in the mantle and i) the limits of geological provinces in the crust, which elucidates the subduction-accretion process in the region; ii) distribution of leucitite and basaltic magmatism; iii) independent geochemical data, and iv) a series of lithospheric steps which constitute areas prone to generating small-scale instabilities in the asthenosphere. This scenario suggests that shear-driven upwelling and edge-driven convection are the dominant melting mechanisms in eastern Australia rather than mantle plume activity, as conventionally conceived. Our study offers an integrated lithospheric model for southeastern Australia and provides insights into the feedback mechanism driving surface processes.

Plain Language Summary

The lithosphere is the outermost rigid layer of the Earth and the focus of important geological processes such as earthquakes (seismic activity), volcanism, and mineralization. The location of these seismically active zones, magma intrusion/production, and ore deposits often coincide with deep discontinuities in the lithospheric structure. Imaging the structure of the lithosphere using geophysical techniques is then crucial to fully understand the nature of these processes. Obtaining the most reliable images of the lithospheric structure requires the joint analysis of two or more geophysical data sets. In particular, the combination of magnetotellurics (an electromagnetic technique) and seismic data holds great potential due to their complementary sensitivity to the Earth’s properties. Combining a joint analysis with a probabilistic approach help us understand the variability of the lithospheric structure better since they provide a large number of models that can explain the data. Given the good data coverage in southeast Australia, we use a new probabilistic approach for the joint analysis of magnetotelluric and seismic data to image the lithosphere structure beneath this region. Our results show a complex lithospheric structure in line with the location of volcanic centers and the tectonic history of the region. Lithospheric composition derived from the models also provides significant insights into how melt production in the area might have occurred.

1 Introduction

Magnetotellurics (MT) has great potential for investigating metasomatism and tectonomagmatic evolution in the lithosphere (e.g., Wannamaker et al., 2008; Comeau et al., 2015; Aivazpourporgou et al., 2015; Wannamaker et al., 2014; Bedrosian, 2016; Blatter et al., 2022; Özaydın & Selway, 2022; Cordell et al., 2022). Imaging the deep thermochemical structure of the lithosphere is crucial due to its connection with surface expressions of fluid and melt pathways, such as the location of ore deposits (e.g., Griffin et al., 2013; Heinson et al., 2018; Kirkby et al., 2022) and volcanic centers (Davies et al., 2015). One of the main limitations of the MT method lies, however, in constraining deep conduc-
tivity structures, particularly beneath shallow conductive features. This is due to the 
the diffusive behaviour of electromagnetic waves and the high sensitivity of MT to con-
ductors (Jones, 1999). The MT method is also ambiguous in discerning the different fac-
tors that affect electrical conductivity, such as temperature, water/melt content and com-
position. Unlocking the full potential of the MT method requires the development of method-
ologies that can assign meaningful physical interpretations to conductivity anomalies and 
discriminate between their causes (Selway, 2014).

A widely adopted methodology to reduce feature ambiguity is combining MT with 
other geophysical data sets via joint inversions (e.g., Khan et al., 2006; Gallardo & Meju, 
2007; Jegen et al., 2009; Moorkamp et al., 2010; Afonso, Rawlinson, et al., 2016; Jones 
et al., 2017). By exploiting the complementary sensitivities of different data sets to the 
properties of interest, joint inversions minimize the range of acceptable models consist-
tent with the available data and increase model resolution (cf. Moorkamp et al., 2007; 
Afonso et al., 2013a; Afonso, Rawlinson, et al., 2016; Afonso, Moorkamp, & Fullea, 2016). 
For example, in the case of MT and seismic data, both data sets are sensitive (to dif-
cerent degrees) to the background thermal and compositional structure of the lithosphere. 
However, only MT is strongly sensitive to minor conductive phases (e.g., hydrous min-
erals and graphite), hydrogen content or small-scale melt/fluid pathways (Karato, 1990, 
2006; Evans, 2012; Yoshino, 2010; Khan, 2016; Selway, 2014; Manassero et al., 2021). In 
this way, joint MT+seismic inversions hold great potential for improving the resolution 
of conductivity structures (e.g., Moorkamp et al., 2007, 2010; Gallardo & Meju, 2007; 
Afonso, Moorkamp, & Fullea, 2016), detecting the presence of partial melt (cf., Selway 
& O’Donnell, 2019; Evans et al., 2019) and fluid pathways (cf., García-Yeguas et al., 2017; 
Bennington et al., 2015) in the lithosphere, and understanding the relationship of these 
lithospheric features with the location of ore deposits (e.g., Takam Takougang et al., 2015) 
and metasomatized lithologies (e.g., Snyder et al., 2014).

In addition to the benefits of joint inversions, valuable information about model 
uncertainties (Tarantola, 2005; Afonso, Rawlinson, et al., 2016; Manassero et al., 2021) 
can be obtained via probabilistic approaches. These approaches provide a range of mod-
els and their probability in explaining the observations by sampling millions of possible 
models and computing a forward solution for each of them. Thus, probabilistic inver-
sions naturally address the non-uniqueness problem in geophysics (particularly in MT) 
and quantify model ambiguity (Tarantola, 2005; Gregory, 2005; Rosas-Carbajal et al., 
2013). Given their intrinsic computational cost, simulation-based probabilistic approaches 
are limited to problems where fast forward operators are available. In the case of 3D MT 
inversions, fully probabilistic methods have been infeasible due to the large CPU time 
required by the associated forward problem (Miensopust et al., 2013). In order to ad-
dress this limitation, Manassero et al. (2020) developed a novel strategy, referred to as 
RB+MCMC, that allows obtaining fast and accurate approximations of the forward so-
lution and performing joint probabilistic inversions of 3D MT data with other data sets 
(Manassero et al., 2021). Potential applications and the efficiency of the method to solve 
the joint inverse problem of MT and seismic data were previously demonstrated with whole-
lithosphere synthetic examples in our previous paper (Manassero et al., 2021).

In this work, we apply the new method to the dense MT and seismic data sets in 
south-eastern Australia to provide new constraints on its complex lithospheric structure. 
This region has undergone several episodes of accretion-subduction and extension pro-
cesses (Glen, 2005), lithospheric deformation (Moresi et al., 2014) and contains part of 
the mafic Eastern Australian Volcanic Province (EAVP), one of the most voluminous 
intraplate volcanic regions in the world (Johnson et al., 1989; Sutherland et al., 2012). 
The EAVP is somewhat unique since half its volcanism is age-progressive and commonly 
linked to a hot mantle plume (e.g., Sutherland et al., 2012; Davies et al., 2015), whereas 
the remaining half is not age-progressive with no obvious melting mechanism (Wellman 
& McDougall, 1974). To further exacerbate this issue, lava compositions throughout the
EAVP (including both age-progressive and non-age-progressive volcanic centers) argue for low-temperature melting of metasomatized mantle source lithologies. In contradiction to the presence of a hot mantle plume, these compositions suggest a source driven by mild perturbations in mantle temperatures (Shea et al., 2022), such as shear-driven upwelling and edge-driven convection (Davies & Rawlinson, 2014; Demidjuk et al., 2007). Since thermo-physical variations in the lithosphere exert control on melt composition and eruption locations (Davies et al., 2015), reliable models of the lithospheric first-order structure are necessary to explain these controversies.

Independent results from conventional ambient noise and teleseismic tomography (Rawlinson et al., 2016; Davies et al., 2015; Young et al., 2013), xenolith thermobarometry (e.g., Lu et al., 2018), thermal modeling (e.g., Tesauro et al., 2020) and a recent 3D conductivity model (Kirkby et al., 2020) have yielded important information about the lithospheric structure beneath southeast Australia. However, some discrepancies are observed between these models despite the fact that they all describe the same region. Beneath the Eastern Volcanics, for example, the sub-lithospheric mantle conductivities in the model of Kirkby et al. (2020) are at odds with the mantle temperatures inferred by Tesauro et al. (2020) and the low velocities imaged by Rawlinson et al. (2016) and Davies et al. (2015). In this work, we attempt to bridge the results of independent studies and obtain reliable lithospheric models that are compatible with independent geophysical, geological, and geochemical evidence in south-eastern Australia. In particular, we aim to assess how the lithospheric structure may have influenced melting mechanisms for the EAVP and to better understand overall melting dynamics in intraplate settings.

To achieve these goals, we use our probabilistic approach to obtain thermal, seismic velocity, and electrical conductivity models (with their uncertainties) by jointly inverting 3D magnetotelluric data and seismic velocities from a tomographic model (Rawlinson et al., 2016). Bulk water content maps are derived from the inversion results to further understand the connection between geodynamic processes and metasomatism.

2 Geological background

The Tasmanides in eastern Australia are a complex orogenic system that developed from west to east through repetitive cycles of subduction and accretion along the eastern margin of Gondwana (Glen, 2005, 2013; Champion et al., 2016; Rosenbaum, 2018). This region is broadly divided into the Delamerian Orogen in the west (early-Palaeozoic) and the younger (mid-Paleozoic) Lachlan Orogen in the southeast (Figure 1.a). Much of the geological complexity in the area can be explained by a geodynamic model of a micro-continent collision and later development of an orocline, referred to as the Lachlan Orocline model (Cayley, 2011; Cayley & Musgrave, 2015; Moresi et al., 2014; Musgrave, 2015). The major structures described in this model are curved crustal geometries with an eastward rotation that persists below the base of the crust (Musgrave, 2015), which have been imaged by gravity, magnetic, and potential field data (e.g., Musgrave & Rawlinson, 2010; Nakamura & Milligan, 2015; Nakamura, 2016; see Figure 1.c-d.); ambient noise tomography in the crust (e.g., Young et al., 2013; Pilia et al., 2015, see Figure 10.d); 2D MT conductivity models (Aivazpourporgou et al., 2015) and a recent 3D MT conductivity model (Kirkby et al., 2020; Heinson et al., 2021). This latter 3D conductivity model shows, for the first time, that some of those crustal structures persist below the Moho, providing new insights about the lithospheric architecture and geodynamic history of the region.

Throughout the late Mesozoic and the entire Cenozoic, eastern Australia has been consistently subjected to voluminous mafic intraplate volcanism, which formed the extensive Eastern Australian Volcanic Province (EAVP, Johnson et al., 1989; Sutherland et al., 2012; Shea et al., 2022). Several regions throughout northern and south-eastern Australia contain recent eruptions; in northern Australia, the Kinrara vent contains lavas...
The EAVP comprises 67 separate volcanic centers with two dominant volcanic center compositions: basalt and potassic leucitite (Figure 1.b). While basaltic volcanics erupted through thinner lithosphere (<110 km) along the eastern and south-eastern seaboard, the leucitite volcanic centers lie on thick lithosphere (>125 km) in central New South Wales and central Victoria (Davies & Rawlinson, 2014; Rawlinson et al., 2017). This leucitite suite represents the most petrologically atypical and extraordinarily enriched melt compositions reported for mafic melts in eastern Australia (Cundari, 1973; Birch, 1978). Particularly, they represent melts from the most pervasively metasomatized source assemblages, likely a Ti-bearing oxide phlogopite websterite ± apatite (Shea et al., 2022), which deviate from anhydrous peridotites. A review of the intra-continental volcanic centers in the EAVP and their source assemblages is presented in Shea et al. (2022). The lack of anhydrous peridotite and the abundance of hydrous minerals in their mantle source assemblages is of particular importance to this work, indicating widespread mantle metasomatism beneath eastern Australia.

3 Methods and data sets

3.1 Data

The data used in our joint probabilistic inversion include magnetotelluric (MT) data from the AusLAMP array (Australian Lithospheric Architecture Magnetotelluric Project) in southeast Australia and the P-wave velocity model of Rawlinson et al. (2016) as seismic data. The long-period MT data were acquired at 298 AusLAMP stations (blue triangles in Figure 1.f) across a ~55 km spaced array covering an area of 950 × 950 km. Details about the data acquisition and processing are given in Kirkby et al. (2020). The MT data are the full impedance tensor for periods between 6.4 to 40,000s. Error floors are set to 5% of max(|Zxx|, |Zxy|) for the components Zxx and Zxy and 5% of max(|Zyy|, |Zyx|) for the components Zyy and Zyx. We assume uncorrelated data errors that follow a double exponential distribution (e.g., Farquharson & Oldenburg (1998); Rosas-Carbajal et al. (2013); Manassero et al. (2021)).

The P-wave velocity model used in this study (Rawlinson et al., 2016) was constructed from teleseismic tomography using data from the mainland component of the WOMBAT transportable seismic array (Rawlinson et al., 2015). In order to account for the unresolved crustal component of the teleseismic arrival time residuals, the model includes a detailed crustal model from ambient noise tomography (Young et al., 2013) and the Moho from AuSREM (Kennett & Salmon, 2012) in the starting model. Using this model, we obtain seismic velocities on a data-point grid of 50×50 km at the surface (shown in red dots in Figure 1.f) and 24 points between the surface and 340 km depth. The data errors are assumed to be uncorrelated and normally distributed with a standard deviation of 1% of the velocity. Examples of data and data fits for MT data and seismic velocities are shown in Figures 2 and 3, respectively. Additional figures can be found in the Supplementary Material.

3.2 Bayesian inversion and model parameterization

In the Bayesian or probabilistic approach to the inverse problem, inference about the model parameters m, given observed data d, is based on the so-called posterior probability density function (PDF):

\[
P(m|d) = \frac{P(d|m)P(m)}{P(d)} \propto L(m)P(m) \propto \exp(\phi)P(m),
\]
**Figure 1.** Maps showing (a) orogens that comprise the Tasmanides of southeastern Australia with grey outline denoting geological provinces (Raymond et al., 2018); (b) Mesozoic to Cenozoic sedimentary basins after Raymond et al. (2012), leucitites volcanoes (orange) and basaltic volcanics (pink) after Shea et al. (2022). The basaltic in NV are highlighted in purple. (c) Total magnetic intensity map (TMI) which includes airborne-derived TMI data for onshore and near-offshore continental areas (Nakamura & Milligan, 2015). (d) Isostatic residual anomaly map (Nakamura, 2016). (e) Moho depth from the AusREM model (Kennett & Salmon, 2012) where 5km-contour lines are shown in dashed-grey. (f) Elevation map of southeast Australia including the AusLAMP MT stations (blue triangles) and the location of the velocity data (red dots). Panel (c) and (d) show major tectonic boundaries are outlined in grey. White triangles indicate stations where data fits are shown.
Figure 2. Posterior PDFs (refer to next section) of MT data for station M5. Field data and error bars are plotted in green and the computed data for the initial model is plotted in blue. Panels (a), (b), (c) and (d): Posterior PDFs of the real and imaginary parts of the off-diagonal components (Zxy and Zyx). Panels (e), (f), (g) and (h): Posterior PDFs of the real and imaginary parts of the diagonal components (Zxx and Zyy). The data has been scaled by the square-root of the period (T) in all panels. The location of the station is shown in Figure 1.f
Figure 3. Posterior PDFs (refer to next section) of P-wave velocity data for stations (a) ST15 located at 139.71 E, 32.74 S and (b) ST182 at 144.4 E, 33.20 S. P-wave velocity data and error bars are plotted in green and the computed data for the initial model is plotted in blue. For those locations, the LAB depths corresponding to the mean, lower and upper bound of the 68% CI models are shown in solid and dashed grey lines, respectively.

where \( P(m) \) denotes the prior PDF describing all the information on the model’s parameters prior to the inversion (e.g., prior geological or petrological knowledge in the area of study). \( L(m) \) is the likelihood function, which is specified by the statistical distribution of the data errors, and \( \phi \) is the misfit of model \( m \). In the case of MT, the data misfit is given by (Tarantola, 2005):

\[
\phi = -\sum_{i=1}^{N} \frac{|g_i(m) - d_i(m)|}{s_i},
\]

(2)

whereas the misfit for the seismic data takes the following form:

\[
\phi = -\frac{1}{2} \sum_{i=1}^{N} \left( \frac{g_i(m) - d_i(m)}{s_i} \right)^2.
\]

(3)

For each data set, \( g \) is the solution of a particular forward problem for model \( m \), \( N \) is the total number of data and \( s_i \) denotes the standard deviation for the \( i \)-th data error.

The posterior PDF over data and parameters is commonly approximated using sampling-based Markov chain Monte Carlo (MCMC) algorithms (Gilks et al., 1995). In our joint inversions of independent data sets, we use the Delayed Rejection Adaptive Metropolis (DRAM) scheme of Haario et al. (2006) in combination with the Cascaded Metropolis (CM) approach (Tarantola, 2005; Hassani & Renaudin, 2013; Manassero et al., 2021). Details about the general inversion framework (RB+MCMC) are given in Manassero et al. (2021) while particular details about the sampling strategy, prior information and initial model used for the current inversion are given in Appendix A.

In order to define the model parameters of the inversion, we treat the conductivity and seismic velocities in the mantle as a superposition of two contributions: background properties related to the long-wavelength thermo-physical state of the dry mantle and anomalous conductivity features associated with the presence of water content, hydrous minerals, grain-boundary graphite films, interconnected sulfides, melts or metasomatized regions (Manassero et al., 2021). A particular realization of the background is defined...
using a column-based parameterization which includes the depth to the thermal lithosphere-
asthenosphere boundary (LAB) and thermal nodes placed in the sub-lithospheric mantle of individual columns. This parameterization is used for both the seismic and MT forward solution. We also use a second parameterization, within the MT problem only, based on electrical conductivity nodes to account for the conductivity in the crust and those conductivity anomalies over the background (see details in Appendix C and in Manassero et al., 2021).

We incorporate one conductivity cell below each MT station as an extra parameter of the inversion to account for the galvanic distortion effect produced by near-surface inhomogeneities that are below the resolution of our model (Jones, 2011; Chave & Jones, 2012; Avdeeva et al., 2015). Similarly to the methodology used in ModEM (Kelbert et al., 2014), these cells are placed in the first (thin) layer of the numerical mesh used to solve the MT forward problem.

The crustal structure used within the seismic models consists of three crustal layers per column (upper crust/sediments, middle crust, and lower crust). Each layer has a fixed thickness and its own set of physical properties: coefficient of thermal expansion, isothermal compressibility, thermal conductivity, bulk density, volumetric radiogenic heat production (RHP), thickness, P-wave ($V_p$) and S-wave ($V_s$) velocities, and $V_p/V_s$ ratio. Whilst all of these can be included as parameters of the inversion, in this work, the unknowns are the $V_p$ of each crustal layer. The layers’ densities are computed from $V_p$ assuming Brocher’s empirical law (Brocher, 2005) while $V_s$ is obtained from the sampled $V_p$ and the prior information on $V_p/V_s$ ratios.

Information about the model parameters (i.e., LAB depths, temperature nodes, $V_p$ velocity in the crust and conductivity nodes) and their uncertainties can be obtained by exploring the posterior PDF. For example, we can estimate the mean model and the models corresponding with the lower and upper bounds of the 68% confidence interval (CI) of the posterior PDF. This interval corresponds with the range of models that fall within one standard deviation from the mean, and is commonly used as an indicator of model uncertainty. This complete information is later used to obtain models in terms of the thermo-physical properties of the whole lithosphere, such as seismic velocities, electrical conductivity and water content.

3.3 Forward Problems and Model Discretization

Most of the forward problems solved during the probabilistic inversion (MT forward in 3D, heat transfer and surface wave dispersion curves) have been described in detail in Manassero et al. (2020, 2021) and in Afonso et al. (2013a, 2013b); Afonso, Rawlinson, et al. (2016). Subsequently, we focus on the model discretization and derivation of the seismic velocity and background conductivity models from the model parameters.

The study area is subdivided into 441 columns of size $0.45^\circ \times 0.45^\circ \times 410$ km, where each column is discretized at three different scales:

1. The finest discretization scale makes up the fine mesh (2 km) to solve the seismic forward operators (surface waves and seismic models) and the steady-state conductive geotherm in the lithosphere.
2. The intermediate discretization comprises the finite elements (FE) used to solve the MT forward problem. In this case, we use $63 \times 36 \times 36$ FE elements of size $17 \times 17$ km in the horizontal and variable vertical size with depth. The air comprises four FE cells and a total thickness of 106 km.
3. The thermal nodes constitute the coarser discretization used to obtain mantle-stable mineral assemblages and the corresponding physical properties (e.g., seismic velocities, bulk density). In this study, the thermal nodes are placed every 50 km in the vertical direction.
The thermo-physical properties at the thermal nodes are obtained by interpolating the properties from pre-computed tables at the nodes’ specific composition, sampled temperature and pressure (computed using a quadratic lithostatic-type approximation, see Appendix B). In order to compute these tables, we use components of the software PerpleX (Connolly, 2009; Afonso et al., 2013b) to solve the Gibbs free-energy minimization problem together with the database and thermodynamic formalism of Stixrude & Lithgow-Bertelloni (2011) within the CFMAS system \((\text{CaO} – \text{FeO} – \text{MgO} – \text{Al}_2\text{O}_3 – \text{SiO}_2)\). All thermophysical properties computed at the thermodynamic nodes are linearly interpolated to the fine mesh for the computation of the seismic forward operator and to the FE mesh for the computation of the MT forward solution (see details in Appendix B).

3.4 Mantle composition

The equilibrium assemblages are computed using a mean bulk mantle composition (i.e., specific CFMAS compositions) of 44.3 wt% \(\text{SiO}_2\), 2.8 wt% \(\text{Al}_2\text{O}_3\), 8.5 wt% \(\text{Fe}_2\text{O}_3\), 39.3 wt% \(\text{MgO}\) and 2.7 wt% \(\text{CaO}\). We estimate this mean composition by averaging eight spinel lherzolites xenoliths (see Table S1 in Supplementary material) that were entrained in EAVP lavas. We use major element compositions from Irving (1980), O’Reilly & Griffin (1987), Griffin et al. (1987) and unreported samples from Bokhara River (J. Shea, personal communications), which cover the area of interest. Since this is the most recent volcanism in eastern Australia, these xenoliths are the most representative samples of current mantle compositions available.

The vast majority of xenoliths in eastern Australia equilibrated at pressures < 2 GPa (O’Reilly & Griffin, 1985; Pearson et al., 1991; Sutherland et al., 1994). However, the CFMAS compositions used here are close to the continental mantle (McDonough, 1990): fertile peridotites (Boyd, 1989b) and pyrolite (Ringwood, 1962). Suggesting these xenoliths have not been modally metasomatized, and their CFMAS concentrations are equilibrated with upper mantle peridotites, which allows for their average to be used as an estimate for current mantle compositions beneath eastern Australia. We note that in cratonic regions, the average mantle compositions may resemble harzburgitic or dunitic compositions (Boyd, 1989b).

The use of an average mantle composition is justified by the fact that the seismic velocities and electrical conductivity have second-order sensitivity to dry bulk mantle composition (see Figure S1, S2 and Özaydın & Selway, 2020; Trampert et al., 2001; Goes et al., 2000). We also assume a dry mantle composition for the background properties. The reason for this is that, firstly, the seismic velocities are not affected by small amounts of water commonly observed in mantle samples (Yu et al., 2011, and references therein) as the actual phase compositions are insensitive to water contents bound to minerals (Cline II et al., 2018). Secondly, since the conductivity nodes represent any conductivity value that departs from the background (dry resistive mantle), this choice allows us to sample only positive anomalies and reduce the number of parameters by two (see Manassero et al., 2021).

3.5 Mantle water content

Using outputs of the joint probabilistic inversion (thermal structure and conductivity models) and the mean mantle composition described above, we obtain estimations of the bulk water content in the mantle (i.e. hydroxyl or \(\text{OH}^{-}\) bound to nominally anhydrous minerals) as a proxy for mantle metasomatism. These estimations are made using the software MATE (Özaydın & Selway, 2020), which includes several experimental models for electrical conductivity, water partitioning and solubility (based on petrological studies). In particular, we used the electrical conductivity models of Gardés et
al. (2014a), Dai & Karato (2009a), Liu et al. (2019), and Dai & Karato (2009c) for olivine, orthopyroxene, clinopyroxene and garnet, respectively.

The solutions for the water content lie between the bounds defined by the dry lithosphere (i.e., 0 ppm) and the maximum bulk water content calculated using the olivine water solubility model of Padrón-Navarta & Hermann (2017). The experimental coefficients used in the water partitioning are: \( D_{\text{opx/ol}}^{\text{OH}} = 5.6 \), \( D_{\text{cpx/opx}}^{\text{OH}} = 1.9 \) of Demouchy et al. (2017) and \( D_{\text{gt/ol}}^{\text{OH}} = 0.8 \) of Novella et al. (2014); which reflect the sub-solidus conditions found in the continental lithospheric mantle in southeastern Australia. Since we aim to portray variations of water content in the mantle rather than fitting the real water content seen in xenoliths, the choice of the experimental parameters is adequate for our calculations. All water calculations are made using the calibration of Withers et al. (2012) for olivine, and the calibration of Bell et al. (1995) for pyroxenes and garnet.

The electrical conductivity of each individual phase is turned into bulk conductivity through the Generalised Archie’s Law (Glover, 2010) with cementation components \( m \) of \( m = 2 \) for orthopyroxenes, \( m = 4 \) for clinopyroxenes and garnet, and \( m < 1 \) for olivine (perfectly connected). The Generalised Archie’s Law is preferred over the conservative estimates of Hashin-Shtrikman lower-bound since it allows us to incorporate the effects of specific minerals in the conductivity values, such as highly-interconnected phlogopites. The main cementation components used here, however, provide similar values to the Hashin-Shtrikman lower-bound for a lherzolitic matrix (Özaydın & Selway, 2020).

4 Results

4.1 Thermal structure of Southeast Australia

The depth to the thermal LAB (1250°C, Afonso, Rawlinson, et al., 2016) obtained from the joint probabilistic inversion of seismic velocities and MT data is shown in Figure 4, while the complete 3D temperature structure is shown via depth slices in Figure 5 (first three columns). Figure 4 also includes a recent LAB model obtained from a 1D joint probabilistic inversion of elevation, surface heat flow, Rayleigh wave dispersion curves, and geoid anomalies using LitMod1D (Haynes et al., 2020; Afonso et al., 2013b); and the estimated LAB depths from two recent seismic tomography models in eastern Australia (Davies et al., 2015; Rawlinson et al., 2017). These results reveal a highly heterogeneous lithospheric structure beneath the Tasmanides: while shallow LAB depths (< 100 km) are imaged on the eastern and south edges of the model, deeper LAB depths (> 250 km) are found beneath the Curnamona Province (CP) and the northern part of the Delamerian Orogen. We observe that the lithospheric structure is also in agreement with the locations of recent volcanism (e.g., Figure 4.a): the leucitite volcanic centers correlate with regions of an intermediate lithospheric thickness (125-160km) while the basaltic volcanoes are located in regions where the LAB depth is less than 120 km. The most distinctive features are step-like changes in the LAB depth from the CP to the southeast corner of the model.

The LAB model is in good agreement with the mean LAB obtained using LitMod1D (Afonso et al., 2013a, 2013b; Haynes et al., 2020), even though the data sets used in each inversion are completely different. The first-order LAB structure is also in agreement with the LAB depths derived from seismic velocity models (Davies et al., 2015; Rawlinson et al., 2017). In particular, we observe similar LAB depths beneath the basaltic volcanoes and west of 146°E, where a wedge-like structure follows the curvature of the Stawell Zone (SZ, see Figure 3.1). There is a clear discrepancy in the center of the models where our inversion yields shallower mean LAB depths (< 150 km). The LAB of Davies et al. (2015) falls, however, within the model uncertainties of our inversion. Even though we use the model of Rawlinson et al. (2016) as data, the reason for this discrepancy is that our in-
version directly samples temperature and favors dynamic features (instead of a thick-conductive LAB) in order to fit all constraining data sets simultaneously.

The main difference between our LAB depths and the models of (Davies et al., 2015; Rawlinson et al., 2017) is that we observe deeper LAB depths beneath CP. At first, this difference could be attributed to the fact that the composition in this cratonic area is more depleted than the average bulk mantle composition used here (see Section 3.3) and deeper LAB depths are needed to fit the fast seismic velocities in this region. However, the seismic velocities computed for a cratonic composition (abyssal peridotite after Boyd, 1989a) are comparable to those computed with the average bulk composition used in the lithosphere (Figure S3a, Supplementary Material). Given the good data fit for the seismic velocities in the region (Figure S3b), we note that the LAB depths found beneath CP are consistent with the Paleoproterozoic-Archean origins of this cratonic region (Page et al., 2005; Hand et al., 2008).

The depths to the thermal LABs obtained after an RB+MCMC probabilistic inversion using MT data only are shown in Figures 6. Compared to the results from the joint inversion, these figures show large variability and lack of structure in the LAB models. This comparison elucidates the fact that MT alone provides low sensitivity to discriminate the temperature from other factors controlling the conductivity. Given the poor constraints provided by MT, it becomes evident that other types of data sets (e.g., seismic) are necessary to image the thermal lithospheric structure properly.

4.2 Seismic velocity structure

Depth slices of the P-wave velocity structure predicted by our model are shown in Figure 5. The P-wave velocity model of Rawlinson et al. (2016) is also included as a reference. In all cases, the velocities are plotted relative to the AusREM model at 34.4°S, 145°E (Figure S4 in Supplementary Material). We observe that the inversion succeeded in reproducing the $V_p$ structure found in the model of Rawlinson et al. (2016). In particular, the mean P-wave velocity down to 100 km is practically identical in both models. Interestingly, the Newer Volcanic province stands out as a low-velocity anomaly at depths between 60 and 80 km. On the other hand, the basaltic volcanoes in the middle of the Eastern Province (∼149°E, 34°S) correlate well with deep low-velocity anomalies.

Some minor discrepancies are observed at depths below 100 km between our results and the model of Rawlinson et al. (2016). For instance, we obtain slightly higher seismic velocities (0.6% higher on average) at depths from 100 to 180 km at the eastern end of the model. Whilst most of these differences fall within the model uncertainties, they can be mostly attributed to the constraints imposed by the MT in the joint inversion. Similarly, we obtain slightly slower velocities throughout the whole model at 200-220 km depth (see Figure 5). At these depths, the local discrepancies are simply explained by the use of different physical parameterizations.

4.3 Electrical conductivity structure

The conductivity models for the crust and mantle predicted by the joint inversion are shown in Figures 7, 8 and 9. For comparison, these figures include the results obtained from a recent deterministic inversion of MT data (Kirkby et al., 2020), using the ModEM software (Kelbert et al., 2014). The main structures observed in the conductivity models are comparable (within model uncertainties) to those in the model of Kirkby et al. (2020) at all depths.
Figure 4. Depth of the thermal LAB. (a) Mean model after the joint probabilistic inversion; (b) mean model obtained after a 1D joint probabilistic inversion (Haynes et al., 2020; Afonso et al., 2013b); (c) and (d) lower and upper bounds of the 68% confidence interval (1 standard deviation from the mean), respectively. (e) and (f) depth of the LAB after Rawlinson et al. (2017) and Davies et al. (2015), respectively. The location of leucitite-bearing volcanism are shown in blue and standard basaltic volcanoes in grey. The 140 km-contour of the LAB depth in shown in dashed-grey line and the outline of the tectonic provinces in solid grey lines. The location of the Stawell Zone (SZ) is marked in panel (a).
Figure 5. Columns (1)-(3): depth slices from the (1) mean model and those models corresponding to (2) the lower and (3) upper bound of the 68% CI of the posterior PDF for the temperature. Columns (4)-(6): depth slices from the models corresponding to (4) the lower and (5) upper bound of the 68% CI, and (6) mean of the posterior PDF for the P-wave velocity; Column (7): P-wave velocity model of (Rawlinson et al., 2016). Selected depths are shown on the left of the figure. In all cases, velocities are plotted relative to 1-D reference model AusREM at 34.4°S, 145°E shown in Figure S4 of the Supplementary Material.
4.3.1 Crust

Figures 10 indicate the agreement between the conductivity structure in the crust and independent information: sedimentary basins (Raymond et al., 2012), magnetic anomalies (Nakamura & Milligan, 2015) and a shear velocity model (Pilia et al., 2015). In particular, Panel (a) shows that the extent of the Paleozoic to Cenozoic sedimentary basins in the region is well outlined by the mean conductivity model at 2 km depth. A visual comparison between Panels (c,e,f) and Panel (b) shows the correlation between total magnetic anomalies and conductivity features at different depths. Examples of these are (A) a conductor in the CP; (B) a SW-NE linear structure close to the NW limit of Murray Basin; conductors (C) and (D) in the Tabberabbera Zone; (E) a N-S conductor aligned with the western border of the Eastern Province; (F) two resistive structures west of the Sydney Basin; (G) the Sydney Basin; (H) a conductive region aligned with the north limit of the Northwest and Central NSW provinces; (I) a highly resistive region in the Stawell Zone near the NSW-VIC border; (J) a circular structure in the middle of the model; (K) a high-conductivity anomaly; and (L) a conductor east of CP.

The conductor (A) correlates well with the conductor seen in the MT study of Robertson et al. (2016), using data from 74 AusLAMP stations placed in the Ikara-Flinders Ranges and CP, and in the recent study of Kay et al. (2022), using a densely-spaced MT modeling scheme. Comparing Panels (c) and (d), we observe that the structures (A)-(G) correlate well with low and high-velocity regions imaged by the shear-wave velocity model of Pilia et al. (2015). We note that the concentric geometries at 29 km depth, such as conductor (J) and structures on the west of the model, resemble the features of the Lachlan Orocline model revealed by potential field and passive seismic data (c.f. Kirkby et al., 2020).

4.3.2 Mantle

The similarities between the mantle conductivity models at ~40-80 km depth and features found in the gravity anomalies (Nakamura, 2016) are shown in Figures 11. These features are: the conductor (A); a conductor (M) in the northwest of the Tabberabbera Zone; two conductivity lineaments (N and O); and a linear conductor (P) placed at the border between the Stawell Zone and the Delamerian Orogen.

The conductivity models between 80-250 km depth (Figures 8 and 9) largely resemble the ModEM model of Kirkby et al. (2020). In particular, we observe a similar
north-eastward orientation of the conductors in the middle of the model ($C_1$, $C_2$, $C_3$, $C_4$ and $C_5$ in Figure 9). Comparing the mantle conductivity with the mean LAB structure in (Figures 12), our models suggest that there is a good alignment between the LAB topography and these mantle conductors (cf. Kirkby et al., 2020). Notably, $C_5$ aligns well with the LAB wedge northwest of the model, $C_3N$ and $C_4S$ follow the 140 km LAB depth iso-surface while $C_1$, $C_2$ and $C_3S$ align with LAB depths < 120 km. We also observe a deep high-conductivity structure beneath the CP ($C_6$), which agrees well with the structure imaged by previous MT studies (e.g., Robertson et al., 2016; Thiel & Heinson, 2013). A high-conductivity region ($C_4$) is observed below the central-leucitite volcanoes. In our models, the extent of this region is larger and more connected than in the ModEM model.

The main difference we observe between our conductivity model and the model Kirkby et al. (2020) is that the sub-lithospheric conductivities along the south-east coast and in the middle of the model are higher than those values found in the model of Kirkby et al. (2020) ($R_1$ and $R_2$ in Figures 9) and in the models from a probabilistic inversion of MT data only (Figure S5, Supplementary Material). The high resistivity ($> 10^4$Ωm) in these regions have been at odds with the mantle resistivity range obtained for sub-lithospheric temperatures and pressures (Fullea et al., 2011; Naif et al., 2021). We observe that, due to the favorable constraint of the seismic data to the thermal structure in the joint probabilistic inversion, these high resistivity values are not permissive anymore.

Another example of the favorable constraint imposed by the seismic data in the conductivity models is shown in Figures 12. At 140 km depth, we observe that the conductors in the east ($C_1$, $C_2$ and $C_3\text{S}$ in Figure 12.c) are located within a region defined by a 1250°C-contour (Figure 12.b). These mantle conductivity structures correlate well with both the location of the eastern basaltic volcanics and a stripe of low P-wave seismic velocities (Figure 12.d). At the same time, the stripe of high seismic velocities beneath the east coast at a depth of 140 km is a clear example of the constraint imposed by the MT data in the velocity models. This stripe is not seen in the models of Rawlinson et al. (2016) and correlates with a relatively cold and highly resistive mantle (Figures 12.b-c).

### 4.4 Joint assessment of bulk water content and temperature maps

One of the key benefits of our inversion is that we can dissociate the effects of the temperature and other factors controlling the conductivity structures, such as water content. The bulk water content maps derived from the mantle conductivity models are shown in Figures 13 and 14. We observe that most of the localized conductive anomalies above the background are likely to contain high percentages of bulk water content. This is the case for the following structures depicted in Figure 14: $C_1$, $C_2$ and $C_3$ beneath the eastern basaltic volcanoes; $C_4$ below the central leucitites; $C_5$ on the eastern boundary of Delamerian Orogen; $C_6$ beneath CP; and the deep localized conductor $C_7$ at $\sim −30.5°N, 147°E$.

Figures 15-17 and Figure S6 show vertical slices of the conductivity, water content and temperature along four transects depicted in Panel (b) of Figure 12. The transects in Figure 15 cross most of the geological provinces on the west and demonstrate a striking correlation between the geological boundaries and the alternation between wet/dry portions of the lithosphere. The joint assessment of these transects clearly shows that the lithospheric mantle beneath CP ($C_6$) corresponds with a highly conductive, hydrated, and cold region. We observe a high-conductivity anomaly ($C_5$) below the Stawell Zone that crosses the LAB. While the high temperatures found in this region ($T_3$ in Figure 15c) can partially explain the conductivity of this structure, Figure 15b indicates that a large part of its conductivity is related to the presence of water content. Similarly, the high-conductivities observed in the region $C_{3N}$ (beneath Tabberaberra Zone) and the conductor $C_{NV1}$ (at $\sim 90$ km depth beneath the NV) are entirely explained by a large amount of water content.
Figure 7. Conductivity in the crust from the joint probabilistic inversion. Columns (1)-(3): depth slices from the (1) lower, (2) upper bound of the 68% percentile and (3) mean conductivity models of the posterior PDF. Column (4): conductivity model of (Kirkby et al., 2020). Selected depths are shown on the left of the figure and the boundaries of geological provinces are shown in grey lines.
**Figure 8.** Mantle conductivity from the joint probabilistic inversion. Columns (1)-(3): depth slices from the (1) lower, (2) upper bound of the 68% percentile and (3) mean conductivity models of the posterior PDF. Column (4): conductivity model of Kirkby et al. (2020). The location of leucitite-bearing volcanism are shown in blue and standard basaltic volcanoes in grey. Selected depths are shown on the left of the figure.
Figure 9. Mantle conductivity from the joint probabilistic inversion. Columns (1)-(3): depth slices from the (1) the lower, (2) upper bound of the 68% percentile and (3) mean conductivity models of the posterior PDF. Column (4): conductivity model of Kirkby et al. (2020). The location of leucite-bearing volcanism are shown in blue and standard basaltic volcanoes in grey. Selected depths are shown on the left of the figure. Dashed-black lines highlight conductors in the mean model and resistors in the ModEM model at 123 km depth.
Figure 10. (a) Sedimentary basins overlying mean conductivity model at 2 km depth and 200 $\Omega$m-resistivity contour in dash lines. (b) Total magnetic anomalies after Nakamura & Milligan (2015) (c) Mean conductivity and (d) shear wave velocity model after Pilia et al. (2015) at 2 km depth. (e-f) Mean conductivity models at 12 and 29 km depth. We refer the reader to the main text for a description of structures A-L. Boundaries of geological provinces are shown in grey lines.
Figure 11. (a), (c) and (d) Mean conductivity models at 46, 61 and 81 km depth, respectively. (b) Isostatic gravity anomalies after Nakamura (2016). We refer the reader to the main text for a description of structures M-P. Boundaries of geological provinces are shown in grey lines.
Figure 12. (a) Mean LAB depth. Contours of the LAB depth every 20 km are shown in grey-dashed line. Mean models at 140 km of (b) temperature (c) electrical conductivity and (d) P-wave velocity relative to 1-D reference model AusREM at 34.4°S, 145°E. The 1250°C-contour (corresponding with the thermal LAB) is plotted in dashed-black in (b-d). Panel (c) shows the location of the geological provinces and conductors in dashed blue. The location of leucitite volcanoes are shown in blue triangles and the surface outcrop of basaltic volcanics are shown in grey in all panels. Panel (b) shows five transects which are discussed in section 5.1.
**Table 1.** Bulk water content and mantle conductivity models from the joint probabilistic inversion. Columns (1)-(3): water content maps obtained from the (1) the lower, (2) upper bound of the 68% CI and (3) mean conductivity models. Column (4): depth slices from mean conductivity models of the posterior PDF. The location of leucitite-bearing volcanism and basaltic volcanoes are shown in orange and turquoise in (3); and blue and in grey in (4). Selected depths are shown on the left of the figure.

<table>
<thead>
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<th>Depth</th>
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<th>Upper bound 68% CI</th>
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<tr>
<td>81km</td>
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<tr>
<td>102km</td>
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![Figure 13.](image-url)
**Figure 14.** Bulk water content and mantle conductivity models from the joint probabilistic inversion. Columns (1)-(3): water content maps obtained from the (1) lower, (2) upper bound of the 68% CI and (3) mean conductivity models. Column (4): depth slices from mean conductivity models of the posterior PDF. The location of leucitite-bearing volcanism and basaltic volcanoes are shown in orange and turquoise in (3); and orange and in grey in (4). We refer to the main text for an explanation of structures C1-C7 Selected depths are shown on the left of the figure.
The north-south transects in Figure 16 show a bulge and two defined steps in the lithospheric structure at \( \sim 50 \), \( \sim 300 \) and \( \sim 650 \) km close to where the northern and central leucitites, and basaltic volcanoes erupted on the surface, respectively. These features in the LAB correlate with the location of sub-lithospheric high-conductivity regions \((C_7, C_4 \text{ and } C_{AS})\). Comparing the conductivity, bulk water, and thermal structure, we observe that while the water content at \( C_7 \) and \( C_4 \) is likely to be large, \( C_{AS} \) correlates with a semi-hydrated to dry mantle region. We also observe that the conductivity of \( C_{AS} \) can be explained by the high anomalous temperatures found in that region \((T_1)\).

Figures 17, with transects across the eastern basaltic volcanoes and the NV, show the continuation of the structure \( C_{AS} \) beneath the NV. A high-conductivity and wet region \((C_1)\) is observed in the sub-lithospheric mantle below the Eastern Province. This deep, wet structure is also seen in Figure S6 (Supplementary material) and correlates with a high-temperature anomaly \((T_3)\). A discussion regarding the relationship between water content, metasomatism and interpretation of the lithospheric structures described in this section is given below.
Figure 16. Vertical slices along transect D-A’ (crossing the leucitite volcanics and basaltic volcanics in the south) of a) the mean conductivity model, b) bulk water content derived from the mean conductivity model and c) temperature. The Moho and LAB depths along that transect are shown in dashed lines in all panels. d) Intermediately connected ($m = 2.5$ in blue) and poorly connected ($m = 5$ in green) phlogopite in a dry lherzolitic matrix that fit the observed conductivities along the transect. The elevation and location of the volcanics and the NV are shown at the top of the figure.
Figure 17. Vertical slices along transect B’-B’-B (across the basaltic volcanoes on the east and the NV) of a) the mean conductivity model, b) bulk water content derived from the mean conductivity model and c) temperature. The Moho and LAB depths along that transect are shown in dashed lines in all panels. The LAB depth along that transect is shown in dashed-black line in all panels. The elevation and location of the volcanics and the NV are shown at the top of the figure.
5 Discussion

5.1 Mantle metasomatism and lithospheric structure

Mantle metasomatism occurs when incipient melts or fluids react with mantle rocks. These reactions can impart chemical enrichments or alter the modal mineralogy of a mantle domain. The latter i) affect the modal proportions of peridotites, ii) introduce new volatile-bearing phases (phlogopite, amphibole, apatite, and carbonates) and, in some pervasive cases, iii) create new lithological domains, such as pyroxenite ± volatile-bearing phase lithologies (e.g. O’Reilly & Griffin, 1987). The generation of volatile-bearing phases reduces the solidus temperature (Foley et al., 2009; Pintér et al., 2021a) and increases the electrical conductivity of the mantle domain (Selway, 2014).

The bulk water content we report in this study acts as a general proxy for metasomatism or mantle fertility, i.e., the inclusion of phases (metasomes) that increase the electrical conductivity of the mantle. This proxy indicates, for example, i) the presence of water and/or phases such as graphite or sulphides for depths above 75-120 km (Selway, 2014; Özyaydın & Selway, 2020); ii) co-existing water and phlogopite in cold mantle below 75-120 km depth; or iii) presence of melt in regions of elevated temperatures (via the joint assessment of bulk water content and temperature models).

5.1.1 Metasomatism across Tasmanides

Figures 15.a-b demonstrate consistent southeastward dipping lithospheric-scale structures from CP to Bendigo Zone, which correlate well with the limits of the geological provinces. We observe a successive alternation of conductive/wet and resistive/dry lithospheres that resemble the west-to-east subduction-accretion process in eastern Australia (Glen, 2005; Shea et al., 2022). The joint assessment of the transects in Figure 15 suggests that the hydrated and cold region C6 is likely subjected to pervasive mantle metasomatism. This result is consistent with the accretion process in the area, whereby successive subduction and orogenic events introduced material into the mantle to act as metasomatic agents and, overtime, preferentially metasomatized the old, thick lithosphere.

The crustal conductor (A) below CP (described in Section 4.3.1) can be seen in Figure 15a. Kay et al. (2022) interpreted this region as deposition of interconnected graphite, which either stems from a shallow/crustal biotic source from tectonic imbrication or an abiotic source related to ascending fluids from a metasomatized mantle. We observe that the deep metasomatized mantle below CP shown in our results provides evidence to explain an abiotic source for the shallow crustal conductors. Moreover, an abiotic source could provide the mantle source reservoir for the unique heavy δ13C group B alluvial diamonds from the Wellington field and Bingara, New South Wales (Davies et al., 2002).

5.1.2 Metasomatism below leucitite volcanoes

The leucitite lavas have melt compositions comparable to lamproites, and were derived from atypical mantle assemblage of phlogopite bearing pyroxenite (Shea et al., 2022). Due to these lava compositions, Kirkby et al. (2020) interpreted the conductors beneath the central leucitites as phlogopite stored in the mantle, suggesting a metasomatized and volatile-rich mantle region. Given the high conductivities (< 100Ω m) and bulk water content (~ 200 ppm) observed around C4 (Figures 16), our results indicate a high probability for the presence of volatile-bearing minerals, likely introduced via modal mantle metasomatism.

Using the water calculation setup described in Section 3.5 and the phlogopite conductivity model of Li et al. (2017), we calculated the electrical conductivities of lherzolite with 5 and 10 % vol. of 0.52 w.t. fluorine-bearing phlogopite (average fluorine value
of mantle rocks, Özaydın et al., 2022) for both perfectly connected (\(m = 1.1\), Modified Archie’s Law) and sparsely populated phlogopites (\(m = 6\), Modified Archie’s Law). The results show that perfectly connected cases are 2.5 orders of magnitude more conductive than the observed conductivities in the region, while the conductivity for sparsely populated/disconnected cases lay near the lower bound of the observed conductivities (Figure S7). These results suggest that a lherzolite with 5-10 % vol of intermediately connected (\(6 < m < 1.1\)) phlogopite explains the conductivities in \(C_4\). Furthermore, leucitite melting from a hydrous pyroxenite source could exhaust phlogopites in the assemblage (Foley et al., 2022). If past melts have destroyed phlogopites from the source region, the residue would be reminiscent of an enriched and hydrated lherzolite (Green, 2015). Since the water content for a residual hydrous lherzolite (Green, 2015) matches with the observed bulk water content in region \(C_4\), we conclude that the most applicable explanation for the high conductivities observed in \(C_4\) is a hydrous lherzolite with small percentages of intermediately interconnected to sparsely populated phlogopite. These results are also illustrated in Figure 16.d, which shows the percentage of intermediately connected (\(m=2.5\)) phlogopite that can explain the observed conductivities across the leucitites volcanoes. Since large volume of phlogopite would drastically lower the seismic velocities (e.g., Selway et al., 2015) and this is not seen in our models, we note that the large percentages (\(\geq 15\%\)) of poorly connected (\(m=5\)) phlogopites are unlikely to explain the conductivities in the region.

The northern leucitites sit above a high-conductivity and metasomatized region below the LAB (\(C_7\)). The ultrapotassic compositions of these lavas suggest lower-degree partial melting (Cundari, 1973) which is consistent with the colder temperatures found in the region. Furthermore, potassium-rich basalts are produced by melting of a metasomatized mantle source that has been enriched in phlogopite (Xu et al., 2017; Förster et al., 2019). We calculated the effect of phlogopites in this region and found similar results to those for the central leucitite (Figures 16.d and S6) favoring intermediately connected phlogopites. The higher conductivities and colder temperature in this region compared to the central leucitites provide a favorable scenario for the presence of existing phlogopites that survived previous melting events.

### 5.1.3 Metasomatism underneath Newer Volcanics

The lithospheric step at \(\sim 650\) km in Figures 16 sits below the basaltic volcanoes. This step is also seen in Figure 15 and correlates with the sub-lithospheric conductor \(C_{2S}\) and a high temperature anomaly of \(\sim 1350-1400^\circ\text{C} (T_1)\). From the low-velocities observed at 60-80 km beneath the NV (Figure 5), Rawlinson et al. (2017) interpreted this temperature anomaly as a mantle upwelling and the source of the NV (see also Rawlinson et al., 2015). The existence of shallow mantle upwellings beneath lava field volcanism provides a favorable setting for decompression melting and mantle metasomatism (Aivazpour-porgou et al., 2015).

In order to assess the existence of current melting in the region, we calculated water-depressed solidus curves (Hirschmann et al., 2009) for four cases of melt with the observed conductivities, temperatures and bulk water content in the lithospheric column beneath NV (Figure S8 in Supplementary Material). These four cases have melt mass fractions (\(\Phi\)) and Generalised Archie’s Law cementation component (\(m\)) of: (I) \(\Phi = 1\%\), \(m = 6\); (II) \(\Phi = 1\%\), \(m = 1.1\); (III) \(\Phi = 10\%\), \(m = 6\); (IV) \(\Phi = 10\%\), \(m = 6\). For the calculations we used the melt electrical conductivity model of Sifré et al. (2014), water-partitioning coefficients between melt and minerals for garnets from Novella et al. (2014) and the coefficients from Hirschmann et al. (2009) for olivine and pyroxenes. The results indicate that the geotherm in the region is not hot enough to maintain melt generation with the water content observed at \(C_{2S}\) only. However, melt can be maintained through the assimilation of carbon-bearing species and an oxidised solidus (Pintér et al., 2021b). Carbonate-bearing xenoliths entrained in lavas from NV show strong evidence for carbonatite meta-
somatization in the source (Yaxley et al., 1991, 1998) and could indicate presence of melt
and incipient metasomatic agents (Frey et al., 1978) in the region.

We interpret that conductive metasomatic agents that percolated from \( C_{3S} \) and
then \( C_{NV1} \) are a feasible explanation for the high conductivity values observed in the
conductors \( C_{NV2} \) and \( C_{NV3} \) found at \( \sim 20-75 \) km depth beneath the basaltic volcanoes
(Figure 15). This finding is consistent with the work of Shea et al. (2022) which shows
that the compositions of these basalts cannot be produced by melting a garnet lherzo-
lite mantle source, instead requiring modally metasomatized mantle sources of enriched
mantle, likely pyroxenites \( \pm \) hydrous phases (Frey et al., 1978; O’Reilly & Zhang, 1995;
Zhang et al., 2001; Zhang & O’Reilly, 1997). This interpretation is further supported
by extensive sampling of metasomatized xenoliths, including carbonate-, phlogopite-, pargasite-
, and apatite-bearing samples found throughout the Newer Volcanics (Frey & Green, 1974;
Yaxley et al., 1991, 1998; Bonadiman et al., 2021; Lu et al., 2020).

5.1.4 Metasomatism beneath Eastern basaltic volcanics

Beneath the volcanics of the Eastern Province, we observe highly-metasomatized
sub-lithospheric regions \( (C_1 \text{ and } C_3) \) and shallow conductors \( (C_{EP1}, C_{EP2}, C_{EP3}) \) at \( \sim 
50 \) km depth (Figures 17 and S6). The conductor \( C_1 \) correlates with a deep temperature
anomaly \( T_3 \) and a low-velocity anomaly (Figures 5). Rawlinson et al. (2015) has inter-
preted this velocity anomaly as the mantle source for the eastern basaltic volcanoes. Fol-
lowing the procedure describe above, we evaluate the existence of melt in this region (Fig-
ures S9) for melt mass fractions \( \phi = 0.25\% \) and \( \phi = 1\% \) using highly interconnected (Tubes
model, Ten Grotenhuis et al., 2005) and unconnected (H-S bounds, Glover et al., 2000)
melt models. The results indicate that the temperatures in \( T_3 \) are not hot enough to main-
tain melt only with water in the system. However, the geotherm might be just hot enough
to produce some percentage of melt in combination with water and carbon-reduced solidus.
If melting exist, it could be low-degree volatile-rich incipient melts that are acting as meta-
somatic agents; which, may be producing metasomatized mantle domains to act as sources
for future major melting events (Shea et al., 2022).

Figure 18 demonstrates the relationship between the location of volcanics in the
Eastern Province, the distribution of shallow conductors (average conductance 20-50 km)
and mantle metasomatism (through average water content near the LAB). According
to these results, basalt fields tend to be associated with shallow conductors (Figure 18a)
and a dryer lithosphere (Figure 18b). Basalts also tend to surround the most meta-
somatized/conductive regions of the lithosphere, similar to what has been observed with
kimberlites worldwide (Özaydın & Selway, 2022). Furthermore, the clear association with
the distribution of lavas and shallow conductors indicate that the lavas sourced from deep
mantle and traversed towards the crust. On their ascend, the lavas precipitated conduc-
tive minerals forming shallow conductors \( (C_{EP1}, C_{EP2}, C_{EP3} \text{ in Figures 17 and S6}) \). Fig-
ure 17 shows a clear conductive pathway from \( C_1 \) to \( C_{EP1} \), while the mantle beneath
the volcanics is relatively dry. This dry mantle beneath the lava fields may indicate that
the high-degree melting events may have exhausted the mantle source region in meta-
somes and water. Another possibility is that basaltic lavas were sourced from adjacent
conductive/metasomatic mantle regions \( (C_1 \text{ and } C_3) \) via oblique trans-lithospheric weak-
ness zones. Further analysis with contributions from geodynamic modeling, seismic to-
mography and melt modeling may be required to understand the full-scope behind the
genesis of melts and their structural control towards the surface in Eastern Australia.

5.2 Implications for magma generation beneath eastern Australia

Age-progressive volcanism in the EAVP, particularly along the Cosgrove track (Davies
et al., 2015), has been widely attributed to mantle plume activity (Wellman & McDougall,
1974). However, Shea et al. (2022) shows primitive melt compositions throughout the
Figure 18. Relationship between Cenozoic Basalts and the parameters derived from the electrical conductivity model: (a) Conductance of the lower-crust (∼ 20 – 50 km), (b) water content calculated around the LAB depth (∼ 100 – 120 km).
EAVP, including all age-progressive volcanism, can only be produced by melting meta-
678
somatized mantle source assemblages such as pyroxenites ± hydrous phases (amphibole
679
and phlogopite) ± Ti-bearing oxides ± apatite. These melt compositions do not show
680
much and mostly have no input from melting of deeper peridotite assemblages. This sit-
681
uation suggests that melt generation temperatures are too low to be driven by mantle
682
plume activity.

683
The solidi for metasomatized mantle assemblages are \( \sim 300^\circ C \) lower than anhy-
684
drous peridotites, due to their high volatile concentrations stored in hydrous minerals.
In this way, melting metasomatized assemblages can produce realistic melt fractions with
685
only slight perturbations above ambient upper mantle temperatures (\( \sim 1350^\circ C \)). Fur-
686
thermore, the models of Duvernay et al. (2021a) show that shear-driven upwelling (SDU)
687
and edge-driven convection (EDC) processes that account for water content in the up-
688
per mantle at \( \sim 1350^\circ C \) can produce enough melt to explain the total melt volume in
689
the EAVP. This scenario, combined with the imaged metasomatized mantle regions, sug-
690
gest that EDC and SDU are the dominant melting mechanism in eastern Australia rather
691
than mantle plume activity (Davies & Rawlinson, 2014; Duvernay et al., 2021b; Shea et
692
al., 2022).

693
Our results show a series of steps in the LAB that correlate well with both the lo-
694
cation of basaltic and leucitites volcanoes (Figures 16 and 19). With the Australian litho-
695
spheric plate moving northeast, these steps constitute areas prone to generating sublitho-
696
spheric small-scale, EDC instabilities (e.g., Zlotnik et al., 2008; Van Wijk et al., 2010;
697
Davies & Rawlinson, 2014; Ballmer et al., 2011; Afonso et al., 2008). A 3D rendering view
698
of these LAB steps, mean conductivity and temperature models is shown in Figure 19.
699
This figure illustrates the proposed model for melt generation in southeastern Australia.

700
The motion of the Australian plate creates an asthenospheric flow towards the south-
701
east which forms an EDC-cell when it encounters a step in the LAB (above \( C_4 \) and \( C_{3S} \)).
702
This mantle flow detaches metasomatized lithologies from thick, older lithosphere and
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drags it into places where EDC occur. The metasomatic lithologies within the EDC-cell
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act reducing the solidus and contribute to primary metasomatized melt sources. The in-
705
cipient melt produced at the EDC-cell reacts with the overlying portion of the LAB to
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create the source of the leucitites volcanoes.

707
Given the continuous displacement of the Australian plate, if one assumes the EDC
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as a regenerative process the question that arises is why the EDC is not generating melt
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continuously? We argued that the metasomatism regions act as a fusible for melt gen-
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eration. Once the metasomatic lithologies (linked to the orogenic accretion) enter the
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cell they deplete and the melt generation stops. The residual of this process (hydrous
712
phases) will most likely emplace in neighbouring regions. The importance of the conduc-
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tivity and water content models obtained in the joint probabilistic inversion is that they
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are imaging the residue of these interactions rather than the actual source of the Ceno-
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zoic volcanoes. Since numerical simulations show that EDC are normally unstable and
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transitory, small perturbations in the sub-lithospheric mantle such as sudden accelera-
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tion of plate motion could produce new EDCs. The production of melt is likely to start
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when the EDC develops close to “new” metasomatized region in the lithosphere.

6 Conclusions

The results presented here demonstrate the feasibility, benefits and performance
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of our probabilistic approach for imaging the thermochemical structure of the lithosphere.
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The joint probabilistic inversion of 3D magnetotellurics (MT) and seismic data helps un-
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lock the full potential of the MT method by providing meaningful interpretations to the
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conductivity anomalies and opens up new avenues for investigating metasomatism and
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tectono-magmatic systems. In particular, our methodology:
Figure 19. A 3D rendering view of these LAB steps, mean conductivity and temperature models depicting interactions between mantle metasomatism and steps in the LAB (red surface). Dashed black arrows show the flow of the asthenosphere and shearing of enriched mantle material into EDC-cells. Hotspots along the LAB are indicated by grey blobs, and locally occur where enriched mantle material crosses its solidus in the sub-lithospheric mantle. Incipient melts (grey arrows) may also travel along the LAB from deep to shallow portions of the lithospheric mantle.
• Succeeded in imaging electrical conductivity, seismic velocity, and temperature structures also identified by independent studies (Kirkby et al., 2020; Rawlinson et al., 2017; Davies et al., 2015; Pilia et al., 2015, cf.), gravity and magnetic data in the region (Nakamura & Milligan, 2015; Nakamura, 2016);
• Deals with the non-uniqueness of the MT problem and provides quantitative information on model uncertainties via well-behaved posterior distributions. This information is crucial for constraining the factors affecting the electrical conductivity. Depending the geological setting, the uncertainty in conductivity can be linked to uncertainties in temperature, partial melt or bulk water content.
• Offers an improved integrated model of the lithosphere beneath southeast Australia, which is compatible with both MT and seismic data as well as geochemical information from xenoliths. The results show improved resolution of the lithosphere-asthenospheric boundary depths and mantle electrical conductivities due to the favorable constraint of the seismic data to the thermal structure in the joint probabilistic inversion.

This study images a highly heterogeneous lithosphere beneath eastern Australia and provides insights for geodynamic and tectono-magmatic processes across multiple scales. The main takeaways that stem from our analysis can be summarized as:

• Widespread mantle metasomatism is identified throughout the region, suggesting complex interactions in the asthenosphere-lithosphere system.
• An alternation of conductive/wet and resistive/dry lithospheres that correlates with the location of geological provinces, resembling the west-to-east subduction-accretion process in eastern Australia.
• Associations between the lithospheric structure, metasomatized regions and distribution of magmatism within the Eastern Australian Volcanic Province (EAVP) are observed. For instance, high correlations are seen between the metasomatized mantle and the location of leucitite and basaltic volcanic centers. The high conductivities observed below the leucitite volcanoes indicate a residual hydrous lherzolite with small percentages of intermittently interconnected to sparsely populated phlogopite.
• The conductivities and temperature found beneath the Newer Volcanics suggest that melt maintained through the assimilation of carbon-bearing phases can exist in the region. This interpretation is supported by samples of metasomatized xenoliths throughout the EAVP.
• A series of steps in the present-day thermal structure correlate with the location of volcanic centers and constitute areas prone to develop small-scale edge-driven convection (EDC) cells in the sub-lithospheric mantle. These results, together with the petrology and melt chemistry presented in Shea et al. (2022), suggest that localized EDC processes are likely to account for the volcanism along the age-progressive tracks in EAVP, rather than a hot mantle plume (Davies et al., 2015; Kirkby et al., 2020).

Appendix A Sampling Strategy

The sampling strategy is specifically tailored to take advantage of the differential sensitivities of the seismic and MT data sets to the background and anomalous structures of the lithosphere (see Manassero et al., 2021). With this in mind, we subdivided the MCMC simulation into three main searches. The first search uses the column-based parameterization only to constrain the background conductivity and seismic velocities associated with the first-order temperature structure and large thermal anomalies. At each MCMC step, new models are obtained by randomly choosing a column and sampling all the column-parameters from their proposal distributions (Gaussian distributions
centered in the current sample). The prior for the LABs are uniform distributions defined in a wide range (60 < LAB < 320 km) to include most of the variability that exists in continental settings (e.g. Griffin et al., 2009; Pasyanos, 2010; Fishwick, 2010; Hasterok & Chapman, 2011; Afonso et al., 2013a, and references therein). The same type of distributions are used for the thermal nodes and Vp velocities in crustal layers with bounds 400−1600°C and 2-6.5 km/s, respectively. The second search focuses on constraining the conductivity anomalies that do not depend on thermo-physical state by sampling the conductivity nodes while still allowing the sampling of the column-parameters. At each MCMC iteration, the algorithm randomly chooses a type of parameter to sample (i.e. column or nodes). If a column is selected, the sampling strategy corresponds to that of the first search, otherwise the algorithm randomly chooses n nodes at a time and assigns a conductivity value from its proposal distribution (log-normal distributions). The prior for the conductivity nodes are Gaussian distributions centered on the background conductivity value (in log-scale) and standard deviation 2 \log_{10}(S/m). This range is large enough to keep the search as general as possible and to include most of the anomalous conductivity over the background. For a new set of nodes’ values, only the 3D conductivity model is updated via interpolation using kriging (Appendix C) and its likelihood is only evaluated with the 3D MT forward solution. After these stages, the proposal distributions are adapted to multi-dimensional Gaussian and log-normal distributions for the column-parameters and the node-parameters, respectively, via the AM algorithm. At each MCMC step of the this third stage, a metropolized independence sampler randomly selects to sample Ncol columns or Nnodes nodes.

Using prior information from previous inversions in eastern Australia, the initial 3D model (i.e., starting point) of our MCMC simulation is constructed by assembling the LAB depths for each 1D column from the model of Rawlinson et al. (2017) The crustal layers’ RHP are obtained with a previous 1D joint probabilistic inversion (Haynes et al., 2020; Afonso et al., 2013a, 2013b). The initial Vp for the crustal layers and Moho depths (Figure 1.e) are taken from the regional AusREM model (Australian Seismological Reference Earth Model, Kennett & Salmon, 2012) and the initial value for each thermal node is derived from an adiabat between the initial LAB and the node at 410 km depth (T_{410} = 1500°C). For the conductivity nodes, the initial values are computed as two orders of magnitude more resistive than the conductivity value at the nodes’ location given by a previous deterministic inversion (Kirkby et al., 2020). In this work, we ran a total of 2,500,000 s MCMC steps for 15 frequencies using 2 processors(Intel(R) Xeon(R) CPU E5-2680 v3 @ 2.50GHz) per frequency. Even with modest computational resources, the inversion took 61 days with an average of 2.64 s per simulation. This represents a time reduction of \sim 94% in the computation of the forward solution for this model.

Appendix B  Mapping Thermochemical Parameters to Background Electrical Conductivity

The background conductivity structure is parameterized using the depth to the LAB and temperature nodes placed in the sub-lithospheric upper mantle. For this, we first discretized the 3D numerical model in Mcol columns. Each column is made up of n_x \times n_y \times n_z FE cells (forward problem discretization) and it is characterized by its own LAB depth, temperature nodes, bulk mantle composition and radioactive heat production (RHP) in the crust. The LAB depth and RHP are used to compute a lithospheric thermal profile by solving the 1D steady-state finite-element heat transfer problem in each column (Afonso, Rawlinson, et al., 2016) with Dirichlet boundary conditions at the surface (T_0 = 10°C) and at each LAB depth (T_{LAB} = 1250°C) (cf. Afonso, Moorkamp, & Fullea, 2016). We also compute a thermal profile in the asthenosphere interpolating the temperature-nodes from the LAB to the bottom of the numerical domain (410 km) and a pressure profile in the whole model using the following quadratic lithostatic-type approximation.
Table B1: Parameters used to compute mantle conductivity

<table>
<thead>
<tr>
<th>Phase</th>
<th>$\sigma_0$</th>
<th>$\sigma_{0i}$</th>
<th>a</th>
<th>b</th>
<th>c</th>
<th>d</th>
<th>e</th>
<th>f</th>
<th>$\Delta V$</th>
<th>$\Delta H_i$</th>
</tr>
</thead>
<tbody>
<tr>
<td>Olivine</td>
<td>2.70</td>
<td>4.73</td>
<td>1.64</td>
<td>0.246</td>
<td>-4.85</td>
<td>3.26</td>
<td>0.68</td>
<td>2.31</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Opx</td>
<td>3.0</td>
<td>1.90</td>
<td>-2.77</td>
<td>2.61</td>
<td>-1.09</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Cpx</td>
<td>3.25</td>
<td>2.07</td>
<td>-2.77</td>
<td>2.61</td>
<td>-1.09</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Garnet</td>
<td>4.96</td>
<td>2.60</td>
<td>-15.33</td>
<td>80.40</td>
<td>-194.6</td>
<td>202.6</td>
<td>-75.0</td>
<td></td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

and the density in the crust:

$$P(z) = 0.99 \times (4.4773 \times 10^{-3} z^2 + 3.2206 \times 10^4 z - 1.284278 \times 10^3) \quad (B1)$$

where $P$ is pressure in Pa and $z$ is depth in meters.

At each FE cell, we obtain the mineral phases in vol% and iron content ($X_{Fe}$) for olivine, orthopyroxene, clinopyroxene and garnet as a function of temperature and pressure in the mantle from pre-computed tables. These tables contain equilibrium assemblages and associated thermophysical properties derived by free energy minimization (Afonso et al., 2013b) for different temperature, pressures and major-oxide compositions. The electrical conductivity for each mineral phase is obtained using the following Arrhenius-type equation with parameters specified in Table B1:

$$\sigma = \sigma_0 \exp \left( -\frac{\Delta H(X_{Fe}, P)}{k_B T} \right) + \sigma_{0i} \exp \left( -\frac{\Delta H_i}{k_B T} \right) + \sigma_p \quad (B2a)$$

$$\sigma_p = f(C_w) \exp \left( -\frac{\Delta H_{wet}(C_w)}{k_B T} \right), \quad (B2b)$$

$$-\Delta H(X_{Fe}, P) = a + bX_{Fe} + cX_{Fe}^2 + dX_{Fe}^3 + eX_{Fe}^4 + fX_{Fe}^5 + P\Delta V, \quad (B2c)$$

where $\sigma_0$, $\sigma_{0i}$ [S/m] and $f(C_w)$ are the small polaron, ionic and proton pre-exponential factors, respectively, $\Delta V$ [cm$^3$/mol] is the activation volume, $\Delta H$, $\Delta H_i$ [eV] and $\Delta H_{wet}$ are activation enthalpies and $X_{Fe}$ is the bulk Fe content in wt%.

The first term in the right-hand side of Equation B2a describes the contribution from small polaron conduction. Its activation enthalpy depends on iron content and pressure and its represented by a polynomial on $X_{Fe}$ (Eq. B2c) plus a term that depends on pressure (the coefficients $a, b, c, d, e, f$ are determined experimentally). The user can choose to consider the effect of the iron content in olivine and orthopyroxene or not. The second term of Equation B2a represents ionic conduction at high temperature and the third term ($\sigma_p$) represents the proton conduction due to hydrogen diffusion. $f(C_w)$ and $\Delta H_{wet}$ are functions of the water content $C_w$ [wt%] and they are obtained from laboratory experiments (see e.g., Pommier, 2014; Jones, 2014, 2016). In this study we include the proton conduction term of Gardés et al. (2014b) for olivine and the term based on Dai & Karato (2009b) for pyroxenes. However, in scenarios where the mantle composition is barely known, we choose to work with a dry mantle and let the conductivity nodes accommodate the effect of water content. The preferred model for small polaron of dry garnet is based on Dai & Karato (2009d) whereas the iron effect is taken from Romano et al. (2006). The parameters for small polaron and ionic conduction for olivine, orthopyroxene clinopyroxene and garnet used in our inversion are summarised in Table A1. Finally, the bulk electrical conductivity corresponding to each FE cell is computed using the lower bound of the Hashin–Shtrikman averaging scheme (Hashin & Shtrikman, 1962, 1963).
Appendix C Conductivity Nodes to Anomalous Electrical Conductivity

The conductivity structures related with smaller-scale features in the crust and over the background are parameterized with 7648 nodes sparsely located within the entire 3D volume. In order to define the location of the nodes, the numerical domain is first subdivided into horizontal layers (every two FE cells in the vertical direction) where their mid-points correspond to the vertical location of the nodes. Considering that bodies with dimensions smaller than the electromagnetic skin depth cannot be resolved by the MT data, the horizontal distance between nodes within each layer is chosen relative to the skin depth for the range of periods and apparent resistivities shown in the observed data. The nodal values are interpolated to each FE cell of the numerical domain via kriging (Gaussian process) interpolation (see e.g. Cressie, 1993; Rasmussen, 1997; Williams & Rasmussen, 1996; Omre, 1987; Gibbs & MacKay, 1997; Gibbs, 1998). The main idea of this method to predict the value of a function $Z$ at $m$ locations from $n$ observations by computing average spatial weights ($W$). In simple kriging, these weights are derived using a known covariance function $c$ between observations (given by the matrix $K_{obs}$) and between the observations and the $m$ estimation locations (given by the covariance matrix $K_{loc}$):

$$W = K^{-1}_{obs} \cdot K_{loc},$$

where $K_{obs} = \begin{pmatrix} c(x_{obs}^{1}, x_{obs}^{1}) & \cdots & c(x_{obs}^{n}, x_{obs}^{1}) \\ \vdots & \ddots & \vdots \\ c(x_{obs}^{n}, x_{obs}^{1}) & \cdots & c(x_{obs}^{n}, x_{obs}^{m}) \end{pmatrix}$ and $K_{loc} = \begin{pmatrix} c(x_{loc}^{1}, x_{loc}^{1}) & \cdots & c(x_{loc}^{1}, x_{loc}^{m}) \\ \vdots & \ddots & \vdots \\ c(x_{loc}^{n}, x_{loc}^{1}) & \cdots & c(x_{loc}^{n}, x_{loc}^{m}) \end{pmatrix}$.

The interpolation (or estimated value) at the $m$ locations is then given by $Z_{loc} = W \cdot Z_{obs}$, where $Z_{obs}$ is the vector containing the $n$ observations. We use a positive definite covariance function with spatially variable correlation lengths (Gibbs & MacKay, 1997; Gibbs, 1998; Manassero et al., 2021):

$$c(x_{m}, x_{n}) = \theta_{1} \prod_{i} \left( \frac{2r_{l}(x_{m})r_{l}(x_{n})}{r_{l}^{2}(x_{m}) + r_{l}^{2}(x_{n})} \right)^{1/2} \exp \left( - \sum_{i} \frac{(x_{m}^{i} - x_{n}^{i})^{2}}{r_{l}^{2}(x_{m}) + r_{l}^{2}(x_{n})} \right)$$

where $r_{l}(x)$ is an arbitrary parameterized function of position $x$ defined in $[-1, 1]^{2} \times [0, 1]$. The form of $r_{l}(x)$ as a function of the scaled coordinates $(x, y, z)$ used in the main text is shown in Procedure 1 in Supplementary Material.

Acknowledgments

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