

# P and S wave travel time tomography of the SE Asia-Australia collision zone

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## Abstract

The South-East (SE) Asia - Australia collision zone is one of the most tectonically active and seismogenic regions in the world. Here, we present new 3-D P- and S-wave velocity models of the crust and upper mantle by applying regional earthquake travel-time tomography to global catalogue data. We first re-locate earthquakes provided by the standard ISC-Reviewed and ISC-EHB catalogues using a non-linear oct-tree scheme. A machine learning algorithm that clusters earthquakes depending on their spatiotemporal density was then applied to significantly improve the consistency of ~~travel-time~~ travel-time picks. We used the Fast Marching Tomography software package to retrieve ~~3D~~ 3-D velocity and interface structures from starting 1-D velocity and Moho models. Synthetic resolution and sensitivity tests demonstrate that the final models are robust, with P-wave speed variations ( $\sim 130$  km horizontal resolution) generally recovered more robustly than S-wave speed variations ( $\sim 220$  km horizontal resolution). The retrieved crust and mantle anomalies offer a new perspective on the broad-scale tectonic setting and underlying mantle architecture of SE Asia. While we observe clear evidence of subducted slabs as high velocity anomalies penetrating into the mantle along the Sunda arc, Banda arc and Halmahera arc, we also see evidence for slab gaps or holes in the vicinity of east Java. Furthermore, a high-velocity region in the mantle lithosphere connects northern Australia with Timor and West Papua. The S-wave model shows broad-scale features similar to those of the P-wave model, with mantle earthquakes generally distributed within high-velocity slabs. The high velocity mantle connection between northern Australia and the eastern margin of the Sunda Arc is also present in the S-wave model. While the S-wave model has a lower resolution than the P-wave model due to the availability of fewer paths, it nonetheless provides new and complementary insights into the structure of the upper mantle

beneath southeast Asia.

*Keywords:* tomography, travel-time, body-wave, SE Asia, earthquakes

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## 1. Introduction

The southern region of the Eurasian plate comprises the continental core of southeast Asia, and is bounded by the Indo-Australian, Pacific and Philippines plates (Figure 1). Significant relative plate motions in the region have created a complex and dynamic setting that encompasses processes such as orogenesis, subduction, crustal accretion, rapid exhumation, megathrust earthquakes and volcanism (Figures 1, 2). Subduction has been the dominant plate-tectonic process in the region since the Mesozoic (Hall, 1997, 2012), with thousands of kilometres of lithosphere subducted into the mantle (e.g. the Tethyan Ocean) while the Australian and Pacific plates moved in northward and westward directions, respectively (Hall and Spakman, 2015). These complex subduction processes are the cause of intense seismicity, which can be used to image the seismic structure of the region in detail.

[Figure 1 Boundaries]

Seismic tomography has been widely used to better understand the lithospheric structure and tectonic evolution of southeast Asia and the surrounding region (Hamilton, 1974, 1979; Fukao et al., 1992; Puspito et al., 1993; Widiyantoro and van der Hilst, 1997; Hafkenscheid et al., 2001; Lebedev and Nolet, 2003; Replumaz et al., 2004; Amaru, 2007; Pesicek et al., 2008, 2010; Hall and Spakman, 2015). While these models are broadly similar, they can lose consistency in regions of small-scale heterogeneity such as subduction zones. The most recent regional interpretation for SE Asia was given by Hall and Spakman (2015). They used the global P-wave model UU-P07 from Amaru (2007) where the most prominent subduction zone was imaged along the Sunda-Banda arc (Sumatra, Java and Banda islands). They suggested that the simplest subduction segment spans the region between West to East Java (Figure 1) and has a relative convergence of 7cm/year and a gap between the trench and volcanic-arc of about 300 km. The initial angle of the slab is estimated

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24 to be around 20° along the trench-volcanic arc before the slab dips more steeply, at around 60°-70°  
25 at a depth of about 150 km.

26 In contrast, subduction beneath the region east of Java appears to be more complicated due to  
27 the presence of a seismicity gap between 250 and 500 km depth, the origin of which is debated.  
28 [Widiyantoro et al. \(2011\)](#) propose that the gap is a hole in the subducted slab with an along-strike  
29 length of about 400 km. [Hall and Spakman \(2015\)](#) note that the aseismic region of the slab is  
30 structural, e.g., thinning of lithospheric mantle, either created during or prior to the subduction.  
31 Alternatively, the seismic gap in the subducting slab might have originated from a strong com-  
32 positional heterogeneity which reduced the rigidity of the lithosphere, making ~~the~~ earthquakes  
33 less likely ([Hall and Spakman, 2015](#)); however, there is little evidence to support this hypothesis.  
34 [Widiyantoro et al. \(2011\)](#) also interpret a second (but smaller) hole located east of the first gap  
35 (between about 200 and 400 km depths, with an along-strike length of about 150 km) as a feature  
36 caused by slab necking. [Hall \(2009\)](#) proposed that these holes resulted from a buoyant thickened  
37 oceanic crust, like the Roo Rise ([Kopp et al., 2006](#)), which arrived at the subduction trench at East  
38 Java from the south. This buoyant object entered the trench producing a slab tear and subducted  
39 together with the rest of the lithosphere, thus producing a hole in the slab highlighted by [Widiyan-  
40 toro et al. \(2011\)](#). However, [Hall and Spakman \(2015\)](#) describe an alternative procedure for the  
41 creation of the tear based on a buoyant object locally blocking the subduction process resulting in  
42 a disconnected slab subducting on either side of it. It has been speculated that this object was the  
43 Roo Rise ([Simandjuntak and Barber, 1996](#)); however, [Hall and Spakman \(2015\)](#) disagree due to  
44 the dimension and position of the hole which implies that it was created about 8 Ma ago.

45 The northwest segment of the Sunda arc near Sumatra is parallel to the relative motion between  
46 India and SE Asia. The subduction beneath Sumatra is partitioned into two segments with a  
47 possible slab tear or fold between them. [Hall and Spakman \(2015\)](#) interpret the low-velocity  
48 anomalies in the middle of Sumatra as a slab tear, while [Pesicek et al. \(2008, 2010\)](#) interpret the  
49 slab in this region to have the form of a NNE- to NE-plunging fold. The tear or fold divides the  
50 trench-normal subduction and the trench-parallel movement in Sumatra. The [Pesicek et al. \(2008\)](#)  
51 and [Hall and Spakman \(2015\)](#) tomographic models are similar in this region, but the interpretation  
52 has one notable difference in the Benioff zone contours. [Pesicek et al. \(2008\)](#) proposed that the

53 Benioff zone extends from Sumatra beneath the Malay peninsula while [Hall and Spakman \(2015\)](#)  
54 limits its extension to the Malaysian coast. Essentially, the interpretation of [Pesicek et al. \(2008\)](#)  
55 states that the Benioff zone contours extend further NE compared to the interpretation of [Hall and](#)  
56 [Spakman \(2015\)](#). [Hall and Spakman \(2015\)](#) observed that the slab dips more strongly north of the  
57 tear or fold compared to the south. The NNE-trending tear or fold thus separates subduction in the  
58 west which penetrates into the lower mantle (down to 800 km) from subduction in the east, which  
59 penetrates into the transition zone (down to 550 km).

60 Seismicity in the region of the Banda Arc exhibits a strongly curved Benioff zone. Two major  
61 contrasting explanations are often given for the shape of the Benioff zone; one suggests a sin-  
62 gle curved subducting slab ([Spakman and Hall, 2010](#); [Hall and Spakman, 2015](#)) while the other  
63 suggests the presence of double subduction from north and south ([Cardwell and Isacks, 1978](#);  
64 [Das, 2004](#)). Based on tomographic images ([Hall and Spakman, 2015](#)) it is observed that the slab  
65 is totally confined to the upper mantle. [Widiyantoro and van der Hilst \(1997\)](#) describe the slab  
66 geometry as spoon-shaped. The slab has the form of a lithospheric fold which bends west and  
67 has a flat-lying portion which sits at the bottom of the upper-mantle. Some authors support the  
68 two-slab model ([Cardwell and Isacks, 1978](#); [Das, 2004](#)) basing their interpretation on the complex  
69 spatial variation of the focal mechanisms. [Hamilton \(1979\)](#); [Charlton \(2000\)](#); [Milsom \(2001\)](#);  
70 [Spakman and Hall \(2010\)](#); [Hall and Spakman \(2015\)](#) suggest that the subduction is due to a bent  
71 and deformed single slab. [Hall and Spakman \(2015\)](#) propose that the Banda slab is caused by roll-  
72 back into the Banda embayment which is part of the Australian continent rather an extension of  
73 a single long-lived subduction zone north of Australia. [The strong seismic activity in the region's](#)  
74 [upper mantle indicates a folded surface. For Spakman and Hall \(2010\), the rollback of the Banda](#)  
75 [embayment is based on the measured age of the backarc basins, which show evidence of a young](#)  
76 [subduction history entirely confined to the upper mantle. The areas surrounding the Banda Sea](#)  
77 [\(Sunda, Sulawesi\) feature subduction which penetrates into the lower mantle, which points to a](#)  
78 [different evolution mechanism. As the Australian plate moved northward at high speed \(7 cm/yr\),](#)  
79 [the increasing resistance of the mantle to plate motion may have progressively folded the slab,](#)  
80 [causing this curved subduction zone \(Spakman and Hall, 2010\).](#)

81 The Molucca Sea is of particular interest because it is the only active arc - arc collision in

82 southeast Asia. The inverted U-shape formed by the two subducting slabs has been recognised for  
83 many years (e.g. [Puspito et al. \(1993\)](#)). The collision comprises two subducting slabs dipping east  
84 (Halmahera) and west (Sangihe) below the two respective volcanic arcs. Both earthquake location  
85 and tomographic studies indicate that the west-dipping slab reaches the bottom of the upper-mantle  
86 (down to about 650 km) and has an overall dip angle of  $45^\circ$  ([Hatherton and Dickinson, 1969](#);  
87 [Puspito et al., 1993](#)). On the other hand, the Halmahera slab only appears to penetrate to a depth  
88 of 400 km ([Hall and Spakman, 2015](#)).

89 In the P-wave ~~travel-time~~ travel-time tomography study of the region by [Hall and Spakman](#)  
90 ([2015](#)) the authors invoke thermal processes at subduction zones to explain many of the velocity  
91 anomalies present in their model. It is well known that the recovery of both P- and S-velocities  
92 better constrains the thermal state of the lithosphere compared to P-velocity alone ([Goes et al.,](#)  
93 [2000](#)). While high temperatures lower both types of velocities, P- and S-waves have different  
94 sensitivities to temperature and composition ([Trampert et al., 2001](#)). Therefore, having both types  
95 of velocity available improves the likelihood of untangling their relative contributions, which in  
96 turn may provide more insight into subduction and other plate tectonic processes.

97 In this paper, we construct high-resolution ( $\sim 130$  km  $\sim 220$  km for P- and S-wave respectively)  
98 3-D seismic models of the crust and upper mantle beneath the SE Asia - Australia collision zone  
99 by inverting both P- and S-wave arrival times. The models use all available arrival times from  
100 earthquake-station pairs in the region of interest that have been archived in ~~international seismic~~  
101 ~~data repositories~~ the International Seismological Centre (ISC) over the last 35 years. A data-  
102 processing strategy based on unsupervised machine learning ([Ester et al., 1996](#); [Pedregosa et al.,](#)  
103 [2012](#)) selects and weights the arrival times used in the inversion for seismic velocities. We applied  
104 a non-linear location method (NLL) ([Lomax et al., 2009](#)) to obtain reliable source locations for  
105 subsequent use in the Fast Marching Method (FMM) ([Sethian and Popovici, 1999](#); [Rawlinson and](#)  
106 [Sambridge, 2004a](#); [de Kool et al., 2006](#)), which is used to solve the forward problem. An iterative  
107 non-linear inversion scheme is applied to constrain velocities and interfaces in the crust and mantle  
108 beneath SE-Asia. The models obtained with NLL are tested against those obtained using the ISC-  
109 Reviewed and ISC-EHB datasets. We also test the influence of using the 3-D crust1.0 ([Laske et al.,](#)  
110 [2013](#)) velocity model for the crust as a starting model, compared to the 1-D reference model ak135

111 (Kennett et al., 1995). Finally, we examine our results in light of previous seismic wavespeed  
112 models of the region.

## 113 2. Data

114 Frequent high-magnitude earthquakes in the study area underpin the high-quality body-wave  
115 arrival-time data available from the catalogues. The arrival times and source locations were down-  
116 loaded from the International Seismological Centre (ISC). The ISC Bulletin is an ideal source of  
117 global arrival times as it comprises the largest collection of freely available seismic data. In this  
118 study, we have used both the ISC-Reviewed (Engdahl and Gunst, 1966) dataset and the updated  
119 ISC-EHB dataset, which is a groomed version of the ISC Bulletin, containing seismic events from  
120 1960 to 2013. The review procedure for the ISC-Reviewed dataset checks that the hypocentre  
121 is in the seismic region for the reported arrivals and reports missing data, magnitude, phase-time  
122 residuals and outliers. The improved dataset produced by Engdahl et al. (1998) and Weston et al.  
123 (2018) benefits from phase re-identification of ISC arrivals and source relocation based on the 1-D  
124 ak135 velocity model (Kennett et al., 1995). By using NLL (Lomax and Curtis, 2001; Lomax  
125 et al., 2001, 2009) on both datasets we obtained two more datasets, making four in total, which  
126 from now on will be referred to as NLL-ISC-Reviewed and NLL-ISC-EHB since they are based  
127 on the ISC-Reviewed and ISC-EHB datasets respectively. However, these datasets are not entirely  
128 independent, since they share many of the same picked arrival times.

129 We used the selection criteria devised by Amaru (2007) to refine our dataset. For P-wave  
130 arrivals we set maximum residuals of  $\pm 7.5$  s and  $\pm 3.5$  s for epicentral distances of less and  
131 more than  $25^\circ$ , respectively. For S-wave arrivals we set our maximum residual to  $\pm 7.5$  s irre-  
132 spective of epicentral distance. The selection criteria are sourced from Bijwaard et al. (1998) in  
133 which they compute the density of travel-time residuals versus the epicentral distance and state that  
134 they no longer display the well-known dependence of ISC delay times on epicentral distance that  
135 indicates deviations of the reference model velocities from the layered averaged real Earth. The  
136 precision of the P phases was estimated following the method of Gudmundsson et al. (1990). The  
137 aforementioned criteria eliminate approximately 7.2%, which amounts to 444514 arrivals from

138 [an original pool of 478822](#). The resultant number of picks obtained using these thresholds are  
139 summarised in Table 1.

140 [Figure 2 Earthquakes]

141 [Figure 3 Stations]

142 Overall, 12 (eight P-wave and four S-wave) tomographic models were produced and compared  
143 to investigate the robustness of our results. The differences between the various models are sum-  
144 marised in Table 1. Crustal phases Pb/Sg and Pg/Sg were incorporated in models P\_F and S\_C  
145 to determine whether the constraints they provide on crustal structure have any influence on the  
146 recovery of the mantle structure. In particular, we test whether our mantle model features any  
147 significant change if we jointly invert for crust and mantle velocity structure or simply invert for  
148 mantle velocities alone. The starting models crust1.0 and ak135 were produced by [Laske et al.](#)  
149 (2013) and [Kennett et al. \(1995\)](#), respectively.

150 [Table 1 Models]

151 As seen in Table 1, the number of picks for NLL datasets are more than the ISC-Reviewed-R  
152 and ISC-EHB-R datasets for the P\_A and P\_B models. This is because the reduced ISC-Reviewed-  
153 R and ISC-EHB-R catalogues are a subset of the initial ISC-Reviewed and ISC-EHB datasets  
154 respectively which only include picks contained in the final NLL solution, subject to the selection  
155 criteria described previously (threshold on residual). For P\_E, P\_F, P\_G and S-wave models we  
156 only use the ISC-EHB dataset since in these cases we would like to include as many data as  
157 possible. [The number of events and receivers ranges from 16490-30360 and 511-665 respectively](#)  
158 [and the specific number of picks per model can be found in Table 1.](#)

### 159 3. Method

#### 160 3.1. Non-linear earthquake location

161 A Non-Linear Location method with Oct-Tree importance sampling ([Lomax and Curtis, 2001](#);  
162 [Lomax et al., 2001](#)) was used for earthquake location prior to tomographic inversion. The Oct-  
163 Tree algorithm provides the maximum likelihood location from the non-linear posterior Proba-  
164 bility Density Function (PDF) of the events. The PDF can also be used to define the spatial

165 uncertainty on the location ([see Figure S1](#)). The ISC-Reviewed and ISC-EHB datasets were re-  
166 located using NLL. For the final event locations we used the ‘‘Global Mode’’ of NLL ([Lomax](#)  
167 [et al., 2009](#)), which is in spherical coordinates and uses a minimum of 40 arrivals for each event.  
168 The maximum hypocentre Root-Mean-Square (RMS) is set to 10s, which restricts the accepted  
169 relocated events to those which have their phase arrival RMS below 10 seconds. These criteria  
170 resulted in a reduction to 41250 from 49206 events for the ISC-EHB catalogue and to 61358 from  
171 322922 events for the ISC-Reviewed catalogue. This is expected since the ISC-EHB catalogue is a  
172 groomed version of the data ~~rather than compared to~~ [the ISC-Reviewed catalogue](#), which includes  
173 lower quality data from 1900 onward with, in some cases, only a few inaccurate arrivals associ-  
174 ated with each event. [Compared to the ISC catalogues, our use of NLL to determine hypocenters](#)  
175 [and origin times is completely automated i.e. there is no manual intervention by an analyst. As](#)  
176 [such, we found that a minimum threshold of at least 40 arrivals per event was necessary to ensure](#)  
177 [stable relocations. Below that number, we found that the location uncertainty of some events](#)  
178 [was very high. By generating an ensemble of tomographic models using a number different but](#)  
179 [arguably robust datasets \(defined by different source locations, different numbers of arrivals, but](#)  
180 [overlapping arrival time picks\), we have a means of assessing the robustness of features that we](#)  
181 [choose to interpret.](#)

### 182 3.2. Improving Signal to Noise ratio (S/N) using unsupervised machine learning

183 In order to improve the S/N ratio and reduce redundancy in the data and compute time required  
184 for the inversion [we automatically determine ray bundles which are then used to form summary](#)  
185 [rays. To do so](#), we modified the approach of [Bijwaard et al. \(1998\)](#) using unsupervised machine  
186 learning. In their study, [Bijwaard et al. \(1998\)](#) formed ray bundles for similar raypaths and applied  
187 a weighting factor to each bundle using the formula

$$W_{rb}^{-1} = \sqrt{\frac{\sum_{i=1}^N (\overline{dt} - dt_i)^2}{N}} \quad (1)$$

188 where  $W_{rb}$  represents the ray bundle weight,  $dt_i$  is the delay of ray  $i$ ,  $\overline{dt}$  is the average delay time  
189 of the ray bundle and  $N$  is the total number of rays in the bundle. Hence, similar raypaths are



190 de-clustered by grouping them together, leaving only one raypath for a specified source-receiver  
191 pair. The final raypath is assigned a weighting factor,  $W_{rb}$  based on the ~~travel-time~~ travel-time  
192 residuals of the raypaths within the cluster. The weighting factors were restricted to vary by less  
193 than one order of magnitude. We ~~also~~-modified the cell division for the ray bundles ~~, following of~~  
194 Amaru (2007), who used a  $0.3^\circ \times 0.3^\circ \times dz$  cell size for the ray bundles with  $dz$  increasing from  
195 15 km at the surface to 40 km at 660 km depth. ~~Unsupervised~~-To determine the ray bundles or  
196 clusters, unsupervised machine learning was implemented using the scikit-learn tool (Pedregosa  
197 et al., 2012) and the Density-Based SCANning (DBSCAN) algorithm of Ester et al. (1996). After  
198 transforming geographic into Cartesian coordinates, we initially set (1) maximum permitted dis-  
199 tance between the clusters to  $0.3^\circ$  and (2) minimum number of events in each cluster to two. The  
200 Cartesian coordinates and the ~~travel-times~~ travel-times, which are known as “features” in machine  
201 learning (Ester et al., 1996) are then used for clustering events. These features are calibrated to  
202 cluster events whose location and ~~travel-time~~ travel-time should be consistent with each other. In  
203 other words, events which are found in a similar location in space with similar arrival times at the  
204 same receiver are grouped together. The new clusters include sources that are detected as “close  
205 sources” for the same receiver. We therefore follow a de-clustering approach (Pyrzcz and Deutsch,  
206 2002) where the identified “close sources” form the ray bundle associated with a weighting uncer-  
207 tainty.

208 ~~This~~-The purpose of unsupervised machine learning is to detect hidden patterns in the dataset  
209 without the need for any training algorithm. The new DBSCAN machine learning approach en-  
210 hances the quality of the generated ray bundles. It takes into consideration a number of attributes  
211 of the data, in particular the euclidean distances across all four dimensions ~~of all the data points~~  
212 ~~(x,y,z location and the travel-time). This is an objective and quantitative measure of how close~~  
213 ~~the points are to each other, with DBSCAN clustering together points that are close together in~~  
214 ~~space. Since we are initially unsure about the number of ray bundles which exist in our dataset,~~  
215 ~~DBSCAN can identify this number using a density-based approach rather than grouping events~~  
216 ~~within a fixed distance to each other~~ travel-time). Here, we include the spatiotemporal location of  
217 the events which we are unsure about their (location and travel-time) exact relationship. DBSCAN  
218 identifies the outliers in a dataset, which in our case are the raypaths which do not form a ray bun-

219 dle. If a group of events are found in a similar position and one has an anomalous travel-time,  
220 then this particular event gets the highest uncertainty and the others gets the uncertainty based  
221 on the formula 1. More importantly, raypaths which carry similar information, i.e., found to  
222 have a similar source-station pair and ~~travel time~~travel-time, do not contribute explicitly to our  
223 model, but do reduce the data noise (assuming that it is random) and time required for inversion  
224 (see Figure ~~S1-S2~~ for a demonstration of the effectiveness of this approach). By doing so, we  
225 don't have to employ low-quality single rays in our model and thus, the ray bundles are of higher  
226 quality. The DBSCAN algorithm is used in a variety of scientific areas from participatory sensing  
227 (Zenonos et al., 2018) to astronomy (Daruru et al., 2010) to obtain hidden patterns in the dataset.  
228 DBSCAN does not previously require to pre-set the number of existing clusters in our dataset. It  
229 can identify this number using a density-based approach rather than grouping events within a fixed  
230 distance, as done by Bijwaard et al. (1998). On the other hand, DBSCAN requires the minimum  
231 number of raypaths needed to form a ray bundle (cluster) and the maximum permitted distance  
232 between the raypaths to form a cluster. We present the DBSCAN approach as a method alternative  
233 to Bijwaard et al. (1998) for grouping similar raypaths together and it is not necessarily a better  
234 method.

### 235 *3.3. Iterative non-linear tomographic inversion*

236 We performed an iterative non-linear tomographic inversion for Vp and Vs variations by ap-  
237 plying the software package FMTOMO (Rawlinson and Sambridge, 2004a,b). FMTOMO uses the  
238 Fast Marching Method (Sethian, 1996; Sethian and Popovici, 1999) for the forward step of ~~travel~~  
239 ~~time~~travel-time prediction in which the eikonal equation is solved on a grid of points. The main  
240 advantages of this method are robustness in the presence of extreme heterogeneity and computa-  
241 tional efficiency, particularly when the ratio of sources to receivers is  $\gg 1$  or  $\ll 1$  (Rawlinson  
242 et al., 2008). Due to the large size of our southeast Asian dataset, the forward step was executed in  
243 parallel mode on a cluster computer in order to reduce the computing time. FMTOMO uses a sub-  
244 space inversion scheme (a gradient-based technique) (Kennett et al., 1988) to solve the linearised  
245 inversion step which includes damping and smoothing regularisation. The ~~iterative sequential~~  
246 application of the forward and inversion steps iteratively solves the non-linear problem ~~since ray~~

247 path geometries are updated after each application of the subspace inversion scheme. The objective  
 248 function that is minimised has the form:

$$S(\mathbf{m}) = (\mathbf{g}(\mathbf{m}) - \mathbf{d}_{obs})^T \mathbf{C}_d^{-1} (\mathbf{g}(\mathbf{m}) - \mathbf{d}_{obs}) + \epsilon (\mathbf{m} - \mathbf{m}_0)^T \mathbf{C}_m^{-1} (\mathbf{m} - \mathbf{m}_0) + \eta \mathbf{m}^T \mathbf{D}^T \mathbf{D} \mathbf{m} \quad (2)$$

249 where,  $\mathbf{g}(\mathbf{m})$  are the predicted residuals,  $\mathbf{d}_{obs}$  are the observed residuals,  $\mathbf{C}_d$  is the a priori covariance  
 250 matrix,  $\mathbf{m}_0$  is the reference model,  $\mathbf{C}_m$  is the a priori model covariance matrix,  $\mathbf{D}$  is the second  
 251 derivative smoothing operator and  $\epsilon$  is referred to as the *damping* factor and  $\eta$  as the *smoothing*  
 252 factor (Rawlinson et al., 2006). According to the subspace scheme, the perturbation  $\delta\mathbf{m}$  required  
 253 to minimise the objective function defined in Equation 2 is:

$$\delta\mathbf{m} = -\mathbf{A} [\mathbf{A}^T (\mathbf{G}^T \mathbf{C}_d^{-1} \mathbf{G} + \epsilon \mathbf{C}_m^{-1} + \eta \mathbf{D}^T \mathbf{D}) \mathbf{A}]^{-1} \mathbf{A}^T \hat{\gamma} \quad (3)$$

254 where,  $\mathbf{A} = [a^j]$  is the  $M \times n$  projection matrix (built from the gradient vector and its rates of  
 255 change),  $\mathbf{G}$  is the matrix of Fréchet derivatives and  $\hat{\gamma}$  is the gradient vector ( $\hat{\gamma} = \frac{\partial S}{\partial \mathbf{m}}$ ). For more  
 256 details on the subspace inversion scheme, see Rawlinson and Sambridge (2003); Rawlinson et al. (2006).

257 In order to represent structure, FMTOMO uses cubic B-splines to describe continuous velocity  
 258 variations from the 3-D model parameter grid; similarly, continuous interfaces such as the Moho  
 259 are described by applying cubic B-splines to a 2-D interface grid. Further details on FMTOMO  
 260 can be found in de Kool et al. (2006) and Rawlinson et al. (2006).

261 We choose to describe our model in terms of a crust and a mantle layer separated by the Moho  
 262 interface, which is defined by the crust1.0 global model of Laske et al. (2013). The 1-D ak135  
 263 reference model was not found to be ideal for a starting model in the mantle, since it produced  
 264 largely positive velocity models below 300 km depth following the inversion. Instead, we have  
 265 produced a reference 1-D model using FMTOMO in what is effectively a 1-D inversion mode (see  
 266 Figure S2S3). The grid spacing of the final 3D-3-D models was set to 1.2° and-in latitude and  
 267 longitude for P-wave model, 2° horizontally-and-in latitude and longitude for S-wave model and  
 268 55 km vertically for both P- and S-wave models respectively. The crustal part of the model was

269 more densely parametrised, leading to a ~~resolution~~ minimum permitted structural scale-length of  
270 approximately  $0.5^\circ$  horizontally and 16 km vertically. We obtained different 3-D models of the  
271 region based on the four sets of earthquake locations, as described above. For P-wave tomography,  
272 we compared results obtained from models P\_A, P\_B, P\_C and P\_D (Table 1). The ~~3D~~ 3-D P- and  
273 S-wave tomographic models obtained with the ISC-EHB dataset (Figures 6, 10) show a ~~reduction~~  
274 ~~of data variance~~ data variance reduction of 63.1% for  $V_p$  and 42.8% for  $V_s$ , which equates to a  
275 final RMS misfit of 681 ms for  $V_p$  and 704 ms for  $V_s$ .

## 276 4. Results

### 277 4.1. Stability and resolution of the results

278 Our dataset of regional ~~travel-times~~ travel-times was inverted for both velocity variations and  
279 interface depth. We analysed the relationship between data variance, model variance and model  
280 roughness to obtain the best damping ( $\epsilon$ ) and smoothing ( $\eta$ ) parameters. We encountered the same  
281 issue as [Pesicek et al. \(2010\)](#), who observed that determining the regularising parameters solely  
282 from synthetic tests produces models that are underdamped. This underdamping occurs because  
283 synthetic tests cannot represent the true noise in the dataset. We instead use a hybrid approach  
284 based on both model results (as sometimes used in global tomography([Pesicek et al., 2010](#)), e.g.  
285 [Pesicek et al. \(2010\)](#)) and trade-off curves (from real and synthetic data) to choose the optimum  
286 smoothing and damping (See Figures ~~S3~~, S4, S5, S6, S7, S8, S9 ~~and~~, S10 and S11). In some  
287 applications of global tomography the optimum regularisation parameters are obtained when the  
288 model best recovers the velocity anomalies that are associated with known geological features  
289 (e.g. subduction zones), while with trade-off tests, optimum regularisation occurs at or near the  
290 point of maximum curvature.

291 We performed a synthetic checkerboard test that includes the addition of Gaussian noise with a  
292 standard deviation of 0.5 s to the synthetic ~~travel-times~~ travel-times in order to simulate the arrival  
293 time picking uncertainty. The new ~~travel-times~~ travel-times serve as the synthetic observables  
294 in the inversion. The test is done for three different checkerboard sizes (noting that the S-wave  
295 checkerboard anomalies are larger than the corresponding P-wave checkerboard anomalies in each

of the three cases owing to much reduced data coverage). The checkerboard sizes for P-wave are: 2° horizontally and 80 km vertically for the coarse grid, 1.2° horizontally and 53 km vertically for the medium grid and 0.75° and 40 km vertically for the fine grid and for S-wave: 3° horizontally 80 km for the coarse grid, 2.2° horizontally and 80 km for the medium grid and 1.2° horizontally and 53 km vertically for the fine grid. The robustness of the solution depends on path coverage and data noise. Checkerboard anomalies were recovered using the same source-receiver paths corresponding to the observables and the same input parameters. The best checkerboard recovery occurs in the Philippines, Sulawesi and along the Sunda arc (see Figure 4). A gap in station distribution across the South China Sea reduces resolution in this region of the model. In general, the coarser checkerboard is better resolved-recovered over a larger region compared to the finer checkerboard, which is to be expected. Cross sections through the checkerboard are shown in Figures S11, S12, S13, S14, S15 and reveal that good resolution can be achieved down to about 800 km depth, although the well-recovered areas tend to decrease as depth increases due to the raypath geometry. The S-wave checkerboard tests, in general, indicate that the minimum size of the recovered anomalies is larger compared to the P-wave results, which is to be expected due to the much smaller size of the than for P-waves; this is expected, as the S-wave dataset (see Figure 5) dataset is significantly smaller. While we do not account for finite frequency effects in our tomography, it is also worth noting that S-waves are typically comprised of longer-period signals, increasing the width of the corresponding sensitivity kernels (Tian et al., 2009). Typically longer wavelength structure ought to be better recovered compared to shorter wavelength structure in the presence of sparse ray coverage.

[Figure 4 Checkerboard P]

[Figure 5 Checkerboard S]

#### 4.2. Tomographic models with different datasets

We seek to test the robustness of our results with respect to the input observables. Different catalogues have different sources, different source locations and different picks. If we test a variety of different datasets and find that similar features emerge, then we can be more confident that they are not artefacts resulting from a particular choice of catalogue. We also test the effect of including

324 crustal phases and different crustal models. All these models are summarised in Table 1. The P-  
325 wave tomographic models (P\_A, P\_B, P\_C, P\_D) can be compared in Figures 6 and 7 at depths  
326 of 200 km and 300 km, respectively. Figure 6 reveals some clear differences between the four  
327 models in a number of regions including the Sunda-Banda arc and Sulawesi region (Figure 1).  
328 In particular, the aforementioned regions appear to be better reconstructed when using NLL-ISC-  
329 EHB (P\_B) and NLL-ISC-Reviewed (P\_D) because these models more closely resemble the results  
330 from previous studies and reveal known geological features. At greater depths (see Figure 7) the  
331 differences between the models are less pronounced in the well-resolved regions.

332 [Figure 6 P 200]

333 [Figure 7 P 300]

334 The P-wave tomographic models show multiple high wave-speed anomalies marking subduct-  
335 ing slabs in the SE-Asia upper mantle (Figures 6-7). One of the best resolved high wave-speed  
336 anomalies is along the Sunda-Java arc (Figure 1), where the Indo-Australian plate subducts below  
337 the Sundaland plate. This is depicted with a high velocity anomaly which shifts towards Borneo as  
338 the depth increases (Figure 8ii). Subduction associated with the Sunda arc extends eastwards until  
339 the Banda arc. The horseshoe-shaped high-velocity anomaly beneath Banda (Puspito et al., 1993)  
340 is evident in Figure 6 while Figure 7 shows a spoon-shaped slab with a flat lying portion beneath  
341 the Banda Sea. The final models also show the Philippine trench and Sangihe-Halmahera arc-arc  
342 subduction near Sulawesi (Figure 8) as localised high velocity zones. That all of these features are  
343 largely consistent across the model ensemble is supported by the votemap we illustrate in Figure  
344 9.

345 [Figure 8 P cross]

346 [Figure 9 Votemap]

347 In the case of S-waves, due to the reduced size of the dataset compared to P-waves, we de-  
348 cided to only use ISC-EHB, for which we produced four results (S\_A, S\_B, S\_C, S\_D as shown  
349 in Table 1). The rest of the datasets were not used for different reasons. ISC-Reviewed is  
350 a poorer quality dataset compared to ISC-EHB and datasets produced by NLL would contain  
351 less data since we require more than 40 arrivals with RMS less than 10s, as described above,  
352 which may result in the loss of important arrivals. This NLL approach is entirely automatic and

353 unsupervised i.e. no manual “grooming” was done by analysts as it was the case of ISC-EHB. We  
354 don’t have complete information on these methods, which are often based on heavily-supervised  
355 approaches; as an example, for the ISC-EHB catalogue, reassign the depth of some events based on  
356 other sources of information. In particular, they plot together the newly relocated and ISC-GEM  
357 (Storchak et al., 2015) events along the subduction zones taking into account their curvature and  
358 use these plots to confirm or modify the earthquakes’ depth. This difference between the two  
359 approaches reduces the number of arrivals produced by NLL significantly enough, not to consider  
360 them for S-wave tomography. The S-wave tomographic model (S\_A), in which only mantle ve-  
361 locity structures are inverted for, is shown in Figure 10. The S-wave models are obtained using a  
362 dataset that is approximately 21% (comparing ISC-EHB datasets i.e. P\_E with S\_A) the size of the  
363 P-wave dataset (Table 1), resulting in a lower-resolution model (see Figure 10). This difference in  
364 resolution makes it difficult to compare the P- and S-wave models directly, although ostensibly the  
365 broad scale features of the two models are quite similar. Figure 11 does exhibit notable differences  
366 from Figure 8, in particular the cross-sections i, ii, iv and v. In Figure 11ii we see no evidence of  
367 continuous subduction beneath Java and the small aseismic region in the slab east of Java, which  
368 is interpreted as a hole by Hall and Spakman (2015) does not correspond to a S-wave velocity  
369 anomaly in Figure 11iv. Finally, the inverted U-shaped arc-arc collision in northeastern Sulawesi  
370 shows a high S-wavespeed anomaly in the Halmahera slab extending down to 500 km as seen in  
371 Figure 11v.

372 [Figure 9-10 S-wave] [Figure 10-11 S cross]

373 Perturbations in P-wavespeed and S-wavespeed crustal structure, relative to the crust1.0 start-  
374 ing model (P\_E, S\_B), are shown in Figure S15-S16; in this example, we have allowed crustal  
375 velocity to be constrained in addition to mantle velocity to examine to what extent variations in  
376 crustal velocities might influence the recovery of mantle structure. A negative velocity anomaly  
377 is shown beneath the Sunda arc in both models. In contrast, a positive velocity anomaly is shown  
378 beneath the Philippines in the P-wave model while a negative anomaly is shown in the S-wave  
379 model. Similar results are shown in Figure S16-S17 where the global model ak135 was used as a  
380 starting model (P\_G, S\_D). Models P\_F and S\_C include crustal phases in the inversion, but they  
381 do not show velocity differences from the models P\_E and S\_B because the number of the Pg/Sg

382 and Pb/Sb phases is small compared to the full dataset. In addition, we have investigated how  
383 the mantle velocity model is affected when inverting for both crust and mantle (see Figures 8 and  
384 ~~S17~~S19) by comparing models P\_D and P\_H. In all our models we also invert for the Moho inter-  
385 face geometry, ~~however;~~ however, we observe no notable differences from the starting interface  
386 we adopt from crust1.0.

### 387 4.3. Discussion

388 Overall, the NLL-selected dataset and ISC-EHB catalogues appear to give better results than  
389 the ISC-Reviewed catalogue. This is expected since the ISC reviewing process only requires that  
390 the depth of the event is appropriate for the region in which it occurs and checks for outliers and  
391 ~~mis-associated~~ incorrectly associated phases. On the other hand, the NLL and ISC-EHB catalogue  
392 perform a relocation of the ISC events (and dynamic phase identification ~~for~~ in the case of ISC-  
393 EHB) (Bondár and Storchak, 2011; Engdahl et al., 1998). ~~A~~ The votemaps for the models P\_A,  
394 P\_B, P\_C, P\_D are shown in Figures 9 and S18. These two figures reveal regions of the model in  
395 which we can be more confident that velocity anomalies exist. In general, all our models agree  
396 on the basic pattern of velocity anomalies. A comparison between our P\_D model and Amaru  
397 (2007) can be seen in Figure ~~S18~~S20 which shows that both models exhibit similar features.  
398 Votemaps have only been produced for models P\_A, P\_B, P\_C, P\_D which occur from different  
399 datasets. Votemaps for S-wave models were not produced since we use the same dataset which  
400 does not lead to major changes in mantle velocities. The model of Amaru (2007) is a global model  
401 constrained by 18 million P-wave picks and uses adaptive parametrisation to deal with irregular  
402 data distribution. In our models, we explicitly include a crustal layer and invert for S-wave as  
403 well as P-wave velocity anomalies using regional sources only. We have also adopted a different  
404 approach for the ~~travel time~~ travel-time prediction and inversion, based on ~~FMM~~FMTOMO, NLL  
405 and the incorporation of crust1.0 (rather than its predecessor crust2.0) - we also include inversion  
406 for crustal velocity and Moho depth. This allows specific inclusion of Pb/Sb and Pg/Sg crustal  
407 phases which further refine crustal structure. Moreover, we have adopted a new machine-learning  
408 clustering approach for similar creating summary rays from raypaths which improved the S/N ratio  
409 and reduced the number of data ~~and as a result, it removed much of the data inconsistencies;~~ as a



410 consequence, data inconsistencies have been removed or suppressed. As seen in Figure ~~S1~~ S2 the  
411 noise in the data prior to signal ~~improvement~~ enhancement reduced the quality of the results and  
412 blurred known geological features such as the arc-arc collision in the Sulawesi region. The high  
413 quality P-wave models produced are in general agreement with the Slab2 subduction geometry  
414 model (Hayes et al., 2018) especially in the Sulawesi region which features a complex subduction  
415 system. Our S-wave model is the first of its kind for southeast Asia, and therefore comparison with  
416 pre-existing S-wave models, such as the one derived from surface wave tomography by Lebedev  
417 and Nolet (2003) is not straightforward. Having both P- and S-wave tomographic models has the  
418 potential to yield greater insight into the geological structure and plate tectonic evolution of the  
419 region. Here, we interpret the two models (P-D,S-A) that were obtained using what we regard  
420 as the best datasets available, but the primary features we interpret are also present in the other  
421 models.

#### 422 4.3.1. Sundaland core

423 All P-wave and S-wave models produced in this study exhibit low velocity anomalies between  
424 100-200 km depth in the region encompassed by the Thai-Malay Peninsula, Borneo, Java Sea  
425 and Sunda Shelf, as can be seen from Figures 6, 9 and 10. This region is part of the Sundaland  
426 continent, which is mostly composed of continental fragments added to Asia during the Triassic  
427 to Cretaceous periods (Hall and Morley, 2004; Hall, 2012; Metcalfe, 2011, 2013; Hall and Spak-  
428 man, 2015). All our models thus point to the presence of the same weakened thermal continental  
429 lithosphere inferred by previous studies (Hafkenscheid et al., 2001; Lebedev and Nolet, 2003; Re-  
430 plumaz et al., 2004; Hall and Spakman, 2015). This thermal weakening might have occurred from  
431 long-term Cenozoic subduction, plumes, or be the result of the interaction between Triassic and  
432 Cretaceous continental blocks (Hall and Spakman, 2015).

#### 433 4.3.2. Sumatra and Java

434 ~~The~~ Figures 8i and 11i show that the western part of the slab under Sumatra dips north at  
435 about 20°, progressively increasing in depth while moving northeastward. The P- and S-wave  
436 models differ slightly between 300 km and 400 km depth, especially between Sumatra and Java  
437 (compare Figures 8i and 11i). Both P- and S-wave models show a major break in the high-

438 velocity slab structure; this contrasts with the P-wave velocity model (~~Hall and Spakman, 2015~~) of  
439 [Hall and Spakman \(2015\)](#) in which the high velocity slab extends continuously in depth. One pos-  
440 sible explanation that is consistent with this observation is the presence of increased temperatures  
441 at a depth of around 350 km, which may contribute to the intense volcanism affecting the region.  
442 [Fukao et al. \(1992\)](#) suggested that the Java slab dips steeply in the lower part of the upper mantle  
443 and proposed a mechanism based on thickened and buckled subducted slab forming a megalith,  
444 which penetrated into the lower mantle due to its high density. This is depicted in Figure 8ii; how-  
445 ever, we also observe a thinning of the subducted lithosphere at approximately 400 km depth. In  
446 contrast, we do not observe the slab in Figure 11ii where ~~probably~~ the thickness of the subducting  
447 slab likely drops below the resolving power of the S-wave dataset at this depth. The absence of in-  
448 tense seismicity at this depth supports an argument of slab spreading with a possible tear. Both P-  
449 and S-wave models provide evidence for serpentinisation at 100 km below central Java as seen in  
450 Figures 8ii and 11ii, with a low velocity anomaly and absence of seismicity in the mantle wedge.

451 Both P- and S-wave tomographic models show a hole in the slab in Figures 8iii and 11iii be-  
452 tween depths of 350 and 500 km in East Java. The hole observed in both the P and S-wave models  
453 points to a temperature increase influencing the velocity of the waves and supports the interpreta-  
454 tion of [Hall and Spakman \(2015\)](#). They suggested that this aseismic, low-velocity anomaly was  
455 produced by an object blocking the slab during Late Miocene subduction, thus producing a tear.  
456 However, the second smaller hole in the slab east of Java is not imaged in the S-wave model,  
457 possibly due to the size of the hole (150 km) being ~~lower~~ smaller than the resolution of the S-  
458 wave model (220 km). In Figure 11vi we show a ~~section~~ cross-section approximately parallel to  
459 ~~subduction~~ the subduction zone. Beneath Sumbawa, a low-velocity aseismic anomaly below 100  
460 km depth may be evidence of serpentinisation as seen in Figure 11iv. Serpentinisation most com-  
461 monly occurs at the plate boundaries where water is released from the descending oceanic crust  
462 and absorbed by the adjacent mantle peridotite ([Cheng et al., 2012](#)).

463 The tear or fold ~~of in~~ the slab beneath North Sumatra is imaged in both P- and S-wave models  
464 in Figures 7 and 10, and is most obviously observed at around 300 km depth. The difference  
465 between P- and S-wave models is most evident in Figures 8v and 11v; here, the S-wave model  
466 shows no evidence of high wave-speed subducted lithosphere starting at approximately 300 km

467 depth, as seen in the western portion of the P-wave velocity cross-section. One possible reason  
468 for the lack of this deeper high velocity anomaly in the S-wave model is a lack of resolution (it is  
469 at the limit of what the S-wave checkerboard test can resolve), although it may also due be to the  
470 smearing of P-waves (although the P-wave checkerboard tests do not suggest that much smearing  
471 is present).

#### 472 4.3.3. *Banda arc*

473 [Puspito et al. \(1993\)](#) imaged the Banda arc subduction zone as a horseshoe-shaped positive  
474 wave speed anomaly. This ~~featured-is~~ feature is also visible in our P-wave model (Figure 6). The  
475 eastern portion of Figure 8vi shows west-dipping subduction with intense seismicity, supporting  
476 the observation of [Puspito et al. \(1993\)](#) of a curved subduction zone. The subducting slab in the  
477 Banda region reaches the bottom of the upper mantle (~700 km depth). [Puspito and Shimazaki](#)  
478 [\(1995\)](#) concluded that this slab does not penetrate into the lower mantle, in contrast to the western  
479 subduction zone along the Sunda-Java arc. The tomographic models of [Widiyantoro and van der](#)  
480 [Hilst \(1996\)](#) and [Widiyantoro and van der Hilst \(1997\)](#) show a laterally-continuous subduction  
481 along the Sunda-Banda arc and north under the Molucca Sea (~~Figure 6~~Figures 6, 9). In their  
482 full waveform tomography, [Fichtner et al. \(2010\)](#) show a high-velocity S-wave anomaly below  
483 Timor at 200 km depth, which is consistent with that retrieved in our S\_A model (Figure. 10).  
484 The anomaly suggests lower temperatures extending from North Australia to the Banda Sea. This  
485 positive velocity anomaly between 100 km and 200 km depth is consistent with the thickness  
486 of ~~the expected~~-Australian Precambrian lithosphere as interpreted by [Fichtner et al. \(2010\)](#) using  
487 the correlation of isotope signatures and tomographic images. The aforementioned two models  
488 agree on the geometry of the subduction zone with an almost vertical subduction of the Australian  
489 lithosphere beneath Sumba, as seen in Figures 8iv and 11iv, at least for the first 350 km.

#### 490 4.3.4. *Sulawesi and Borneo*

491 [Puspito et al. \(1993\)](#) proposed that the western limb (Sangihe slab) of the Molucca Sea plate  
492 may penetrate into the lower mantle in contrast to the eastern limb (Halmahera slab), which only  
493 reaches depths of approximately 400 km. This is confirmed from our study as shown in Figure  
494 8v, where the Sangihe slab reaches depths of approximately 700 km while the Halmahera slab

495 is shown to terminate at 400 km depth. In contrast to the P-wave modelling results, the S-wave  
496 model in Figure 11v shows that the eastern dipping slab (Halmahera) reaches depths of 500 km  
497 although this difference may be due to the more limited resolving power of the S-wave dataset.

498 In the northern part of Borneo we observe (Figures 6, 7, 8iii, 10 and 11iii) a high velocity  
499 anomaly between 100-300 km depth which is also observed by Hall and Spakman (2015). They  
500 suggest that the anomaly might be an artefact of the poor data coverage, but it is present in both  
501 P-wave and S-wave models and appears to be resolved according to our synthetic resolution tests.  
502 The absence of seismicity in the area suggests that this anomaly might be an indication of possi-  
503 ble remnant subduction; for instance, both Cottam et al. (2013) and Hall (2013) suggest that the  
504 anomaly may represent a broken off part of the slab from northerly subduction of the Celebes Sea  
505 in the mid-late Miocene, which terminated only 5 Ma. Alternatively, it could be related to subduc-  
506 tion termination of the South China Sea in the mid-Miocene when the Dangerous Grounds block  
507 collided with the Sabah-Cagayan volcanic arc (Cottam et al., 2013). The lithospheric thickness  
508 below Borneo is estimated to be around 100 km based on the depth of the P- and S-wave velocity  
509 ~~increase~~decrease observed in Figures 8iii and 11iii.

## 510 **5. Conclusions**

511 We have developed 12 tomographic models in total, with the aim of providing a robust and con-  
512 sistent picture of the upper-mid mantle beneath SE Asia. These 12 models include eight ~~P-~~P-wave  
513 and four S-wave models which were produced using an iterative non-linear inversion scheme in  
514 which FMM was used for ~~travel-time~~travel-time prediction and a subspace inversion scheme for  
515 adjusting model parameters in order to satisfy observations. This method was used to constrain the  
516 3-D seismic structure of SE Asia with four datasets and different starting models. We incorporated  
517 the crust1.0 model to minimise the downward smearing of crustal structure in order to improve the  
518 mantle model and have examined the influence of inverting for crustal structure using phases such  
519 as Pg. Moreover, we generated new S-wave tomographic models, which provide fresh insight into  
520 the subduction processes taking place in the collision zone.

521 Based on the results of this study, we conclude that using the NLL locations and ISC-EHB  
522 catalogue in the inversion produced better P-wave models compared to using the default catalogue

523 locations. ~~The~~Inversion of the S-wave dataset resulted in a model that was comparable to a  
524 smoothed version of the P-wave model although there were a number of clear differences in the  
525 Sulawesi and Java regions. In particular, all models agree that in the region of the Thai-Malay  
526 Peninsula, Borneo, Java Sea and Sunda Shelf low-velocity anomalies suggest thermal weakening  
527 of the continental lithosphere. In addition, we confirm that the subducting slab below Java dips  
528 steeply in the upper part of the lower mantle, with a possible thinning of the subducted lithosphere  
529 at approximately 400 km depth. This feature is confirmed in the S-wave model, where we can  
530 see no evidence of subduction possibly due to the reduced resolution. Moreover, both the P-wave  
531 and S-wave models show that the Sangihe slab of the Molucca Sea may penetrate into the lower  
532 mantle while the Halmahera slab reaches a depth of only 400-500 km. A hole in the slab beneath  
533 East Java is apparent in both the P-wave and S-wave models. It is likely due to an absence of  
534 cold lithosphere caused by a tear, which explains the lower wavespeeds observed in both P- and  
535 S-wavespeeds. A smaller hole is located to the east of this hole in the P-wave model but is not  
536 visible in the S-wave model. Finally, we observed a consistent high velocity anomaly beneath  
537 North Borneo, reaching 300 km depth, which may be a signature of remnant subduction related to  
538 recent subduction termination (5 Ma) in the northern Celebes Sea.

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Table 1: A summary of the characteristics of the tomographic models generated as part of this study.

<b>Model name</b>	<b>Dataset</b>	<b>No. of picks</b>	<b>Starting model (crust/mantle)</b>	<b>Inverting for</b>
P_A	ISC-EHB-R	219592	crust1.0/ak135	mantle
P_B	NLL- ISC-EHB	263662	crust1.0/ak135	mantle
P_C	ISC- Reviewed-R	266696	crust1.0/ak135	mantle
P_D	NLL-ISC- Reviewed	309964	crust1.0/ak135	mantle
P_E	ISC-EHB	444514	crust1.0/ak135	crust/mantle
P_F	ISC-EHB with Pg, Pb	446785	crust1.0/ak135	crust/mantle
P_G	ISC-EHB	444514	ak135/ak135	crust/mantle
P_H	NLL-ISC- Reviewed	309964	crust1.0/ak135	crust/mantle
S_A	ISC-EHB	93850	crust1.0/ak135	mantle
S_B	ISC-EHB	93850	crust1.0/ak135	crust/mantle
S_C	ISC-EHB with Pg, Pb, Sg, Sb	96975	crust1.0/ak135	crust/mantle
S_D	ISC-EHB	93850	ak135/ak135	crust/mantle

667

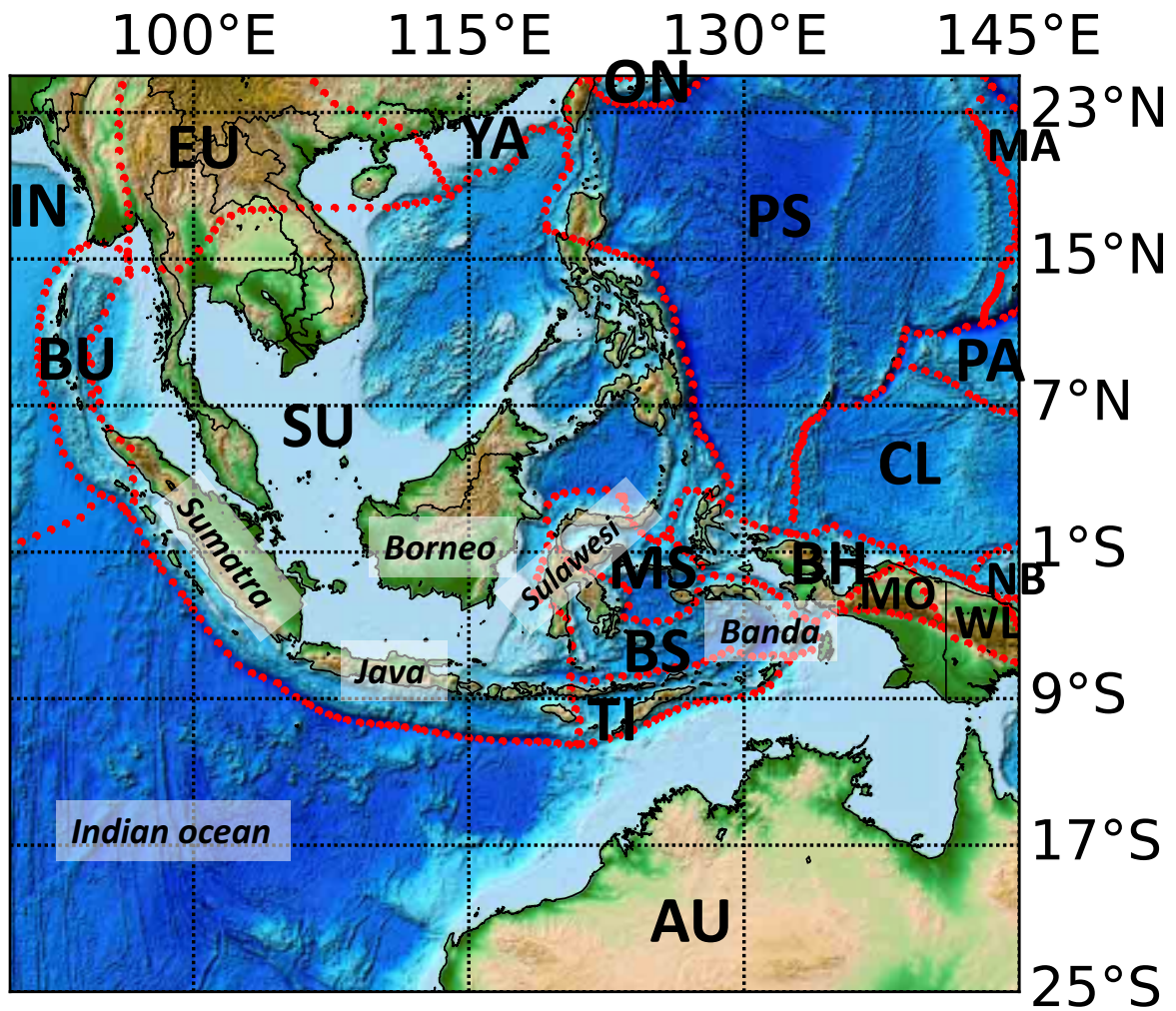


Figure 1: Plate boundaries of SE Asia as interpreted by Bird (2003). The tectonic plates which are labelled include, AU: Australia, BH: Birds Head, BS: Band Sea, BU: Burma, CL: Caroline, EU: Eurasia, IN: India, MA: Mariana, MO: Maoke, MS: Molucca Sea, NB: North Bismarck, ON: Okinawa, PA: Pacific, PS: Philippine Sea, SU: Sunda, TI: Timor, WL: Woodlark, YA: Yangtze.

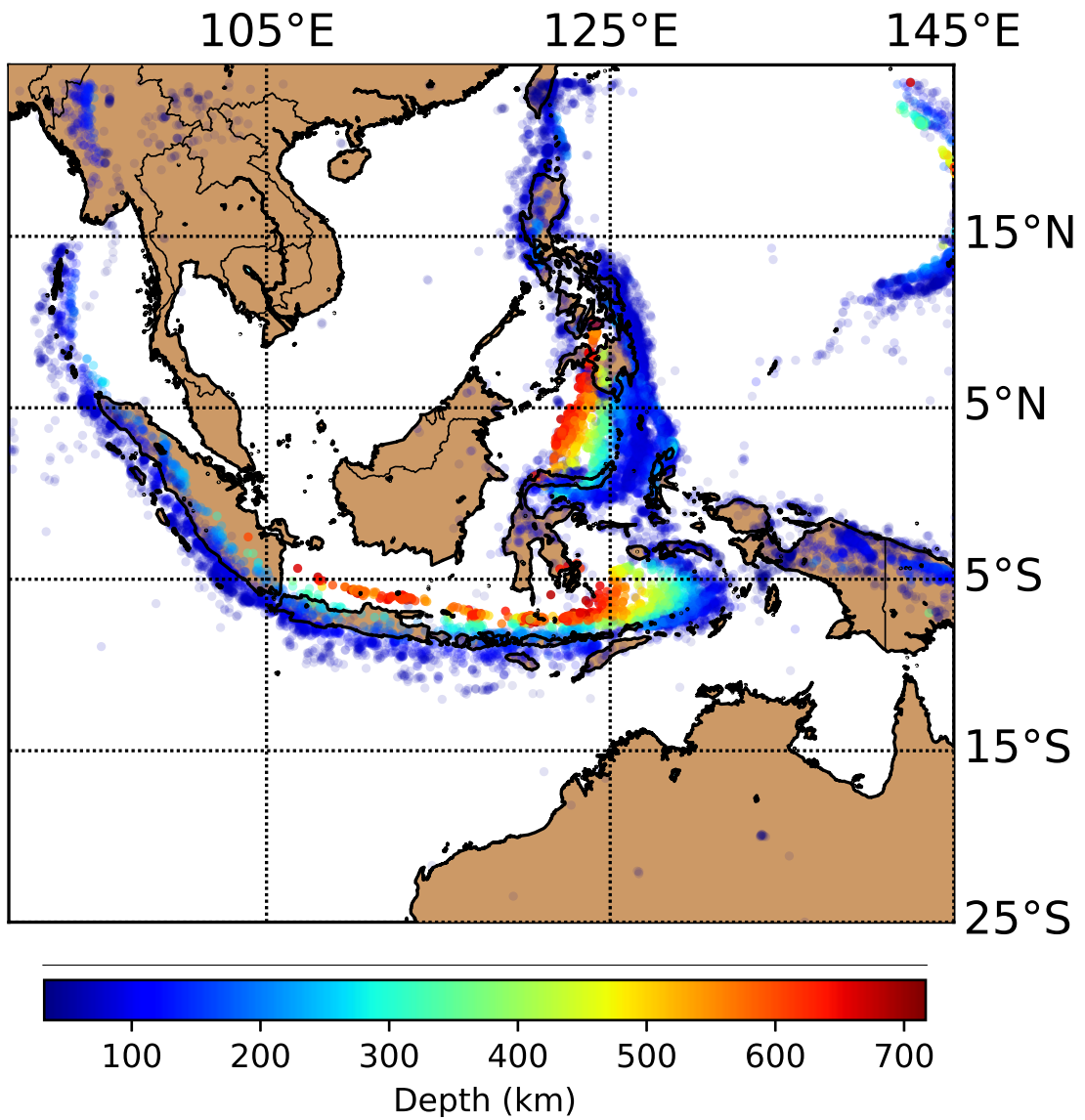


Figure 2: Regional seismicity distribution in SE Asia used for the construction of the tomographic models. For visualisation purposes greater depths are represented with greater opacity.

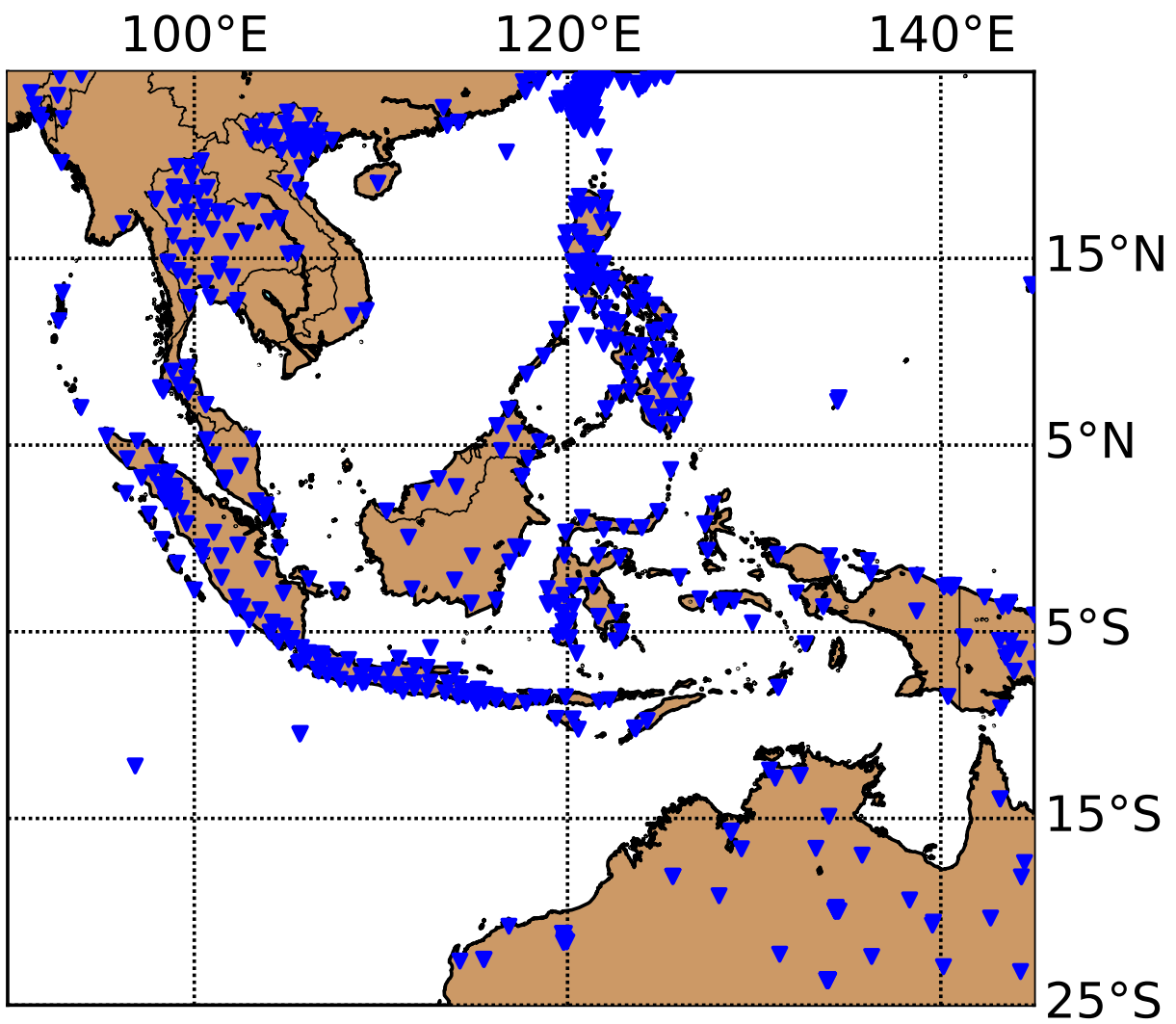


Figure 3: Map of station locations in SE Asia as used by the ISC-EHB dataset.

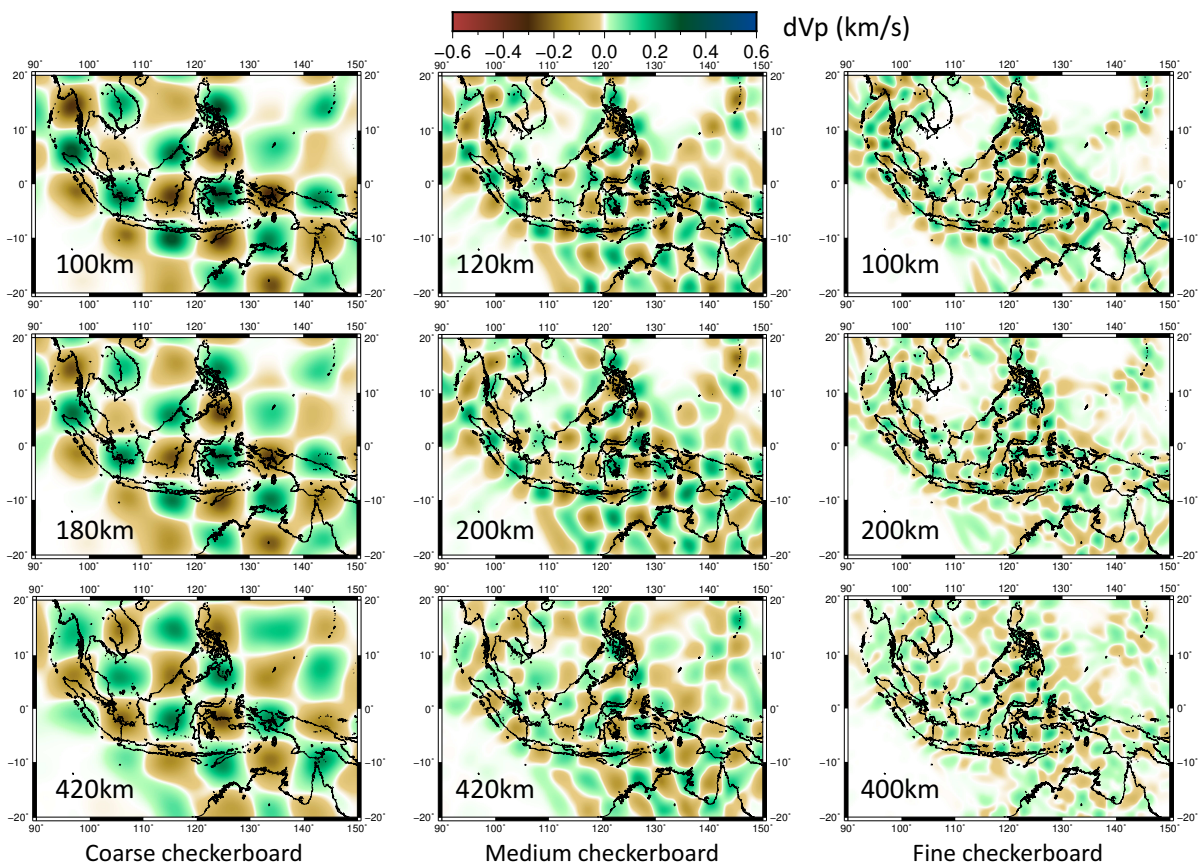


Figure 4: Results from P-wave checkerboard tests with alternating high and low velocity patterns of 0.4 km/s maximum perturbation with the P\_B model source-receiver pairs.



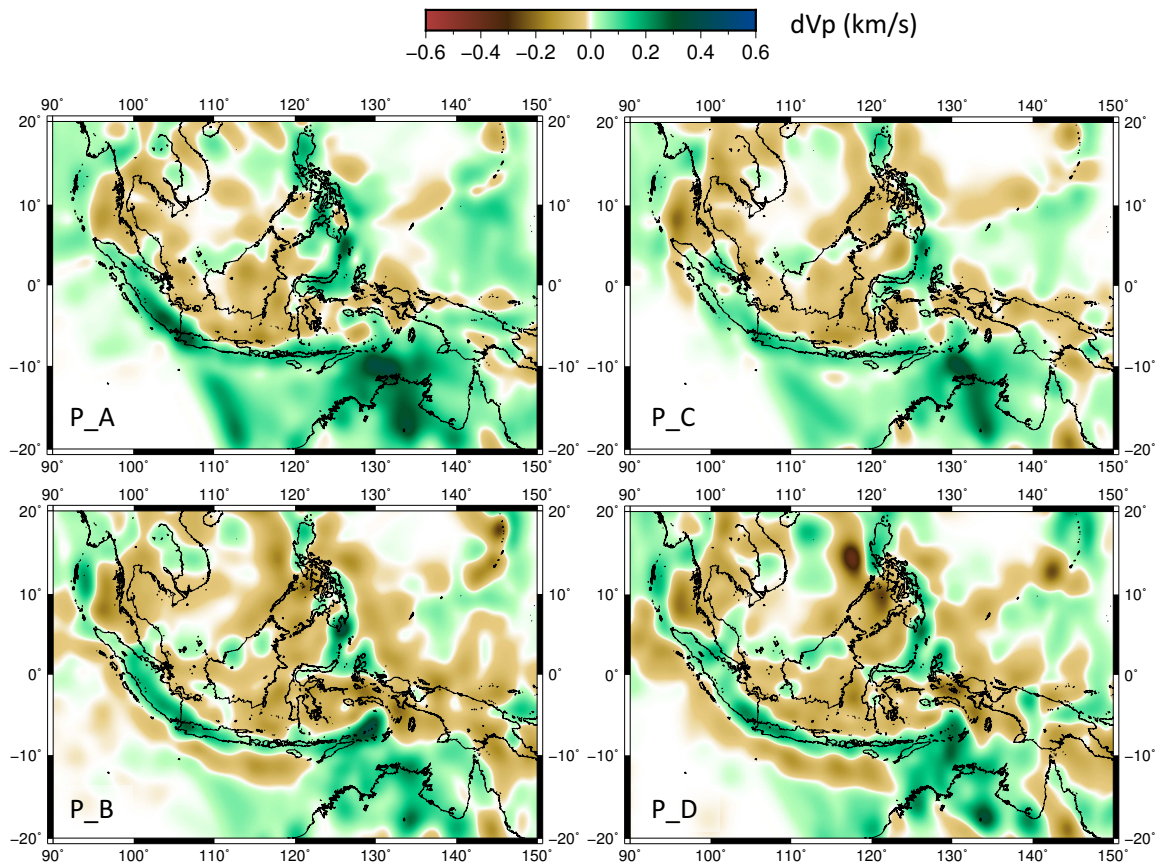


Figure 6: Comparison of the four (P\_A, P\_B, P\_C, P\_D) P-wave tomographic models using ISC-EHB-R, ISC-Reviewed-R, NLL-ISC-EHB and NLL-ISC-Reviewed datasets respectively at 200 km depth. The different input parameters used by the models are described in Table 1.



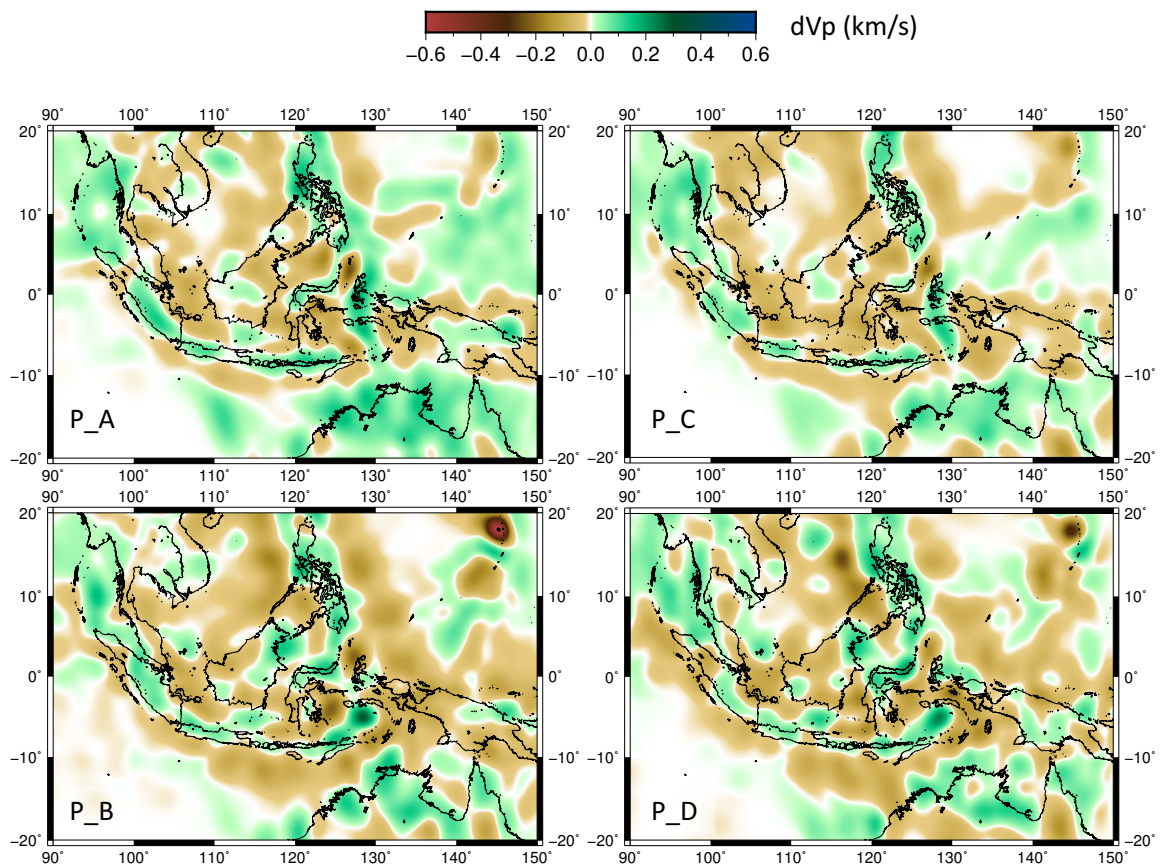


Figure 7: Same as Figure 6 but this time the model is displayed at 300 km depth. The different input parameters used by the models are described in Table 1.

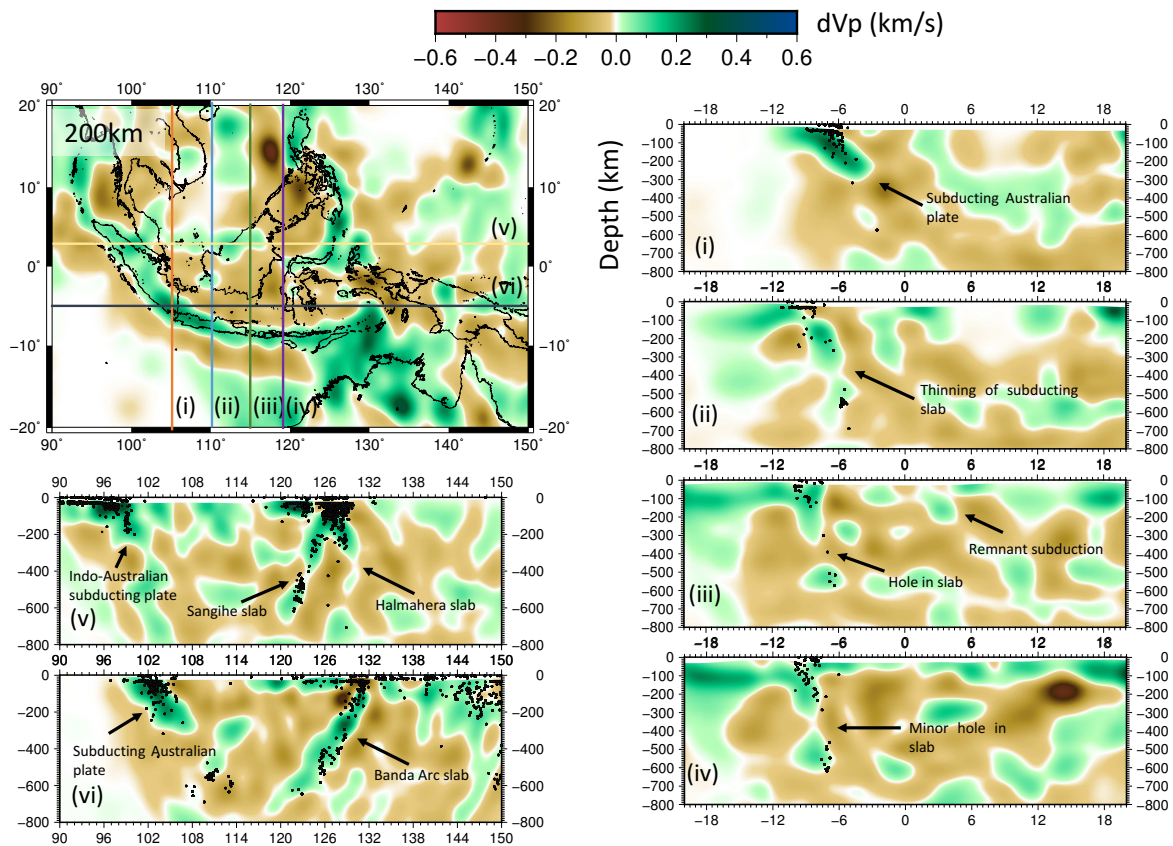


Figure 8: Cross-sections through the P-wave model using the NLL-ISC-Reviewed (P<sub>D</sub>) dataset. Features with-of interest are outlined with a purple-dashed-line-labelled arrows in the cross-sections through the model. Various slices show different features in the region: (i) subduction along Sumatra, (ii) slab in-beneath Java region, (iii) major hole in the slab below east Java, (iv) minor hole in the slab east of the major hole, (v) Sangihe and Halmahera arc-arc collision and the start of the subducting slab in northwestern Sumatra, (vi) curved subduction near Banda arc and the subducting slab in-below Sumatra near the tear (see break as discussed in section 4.3.2).

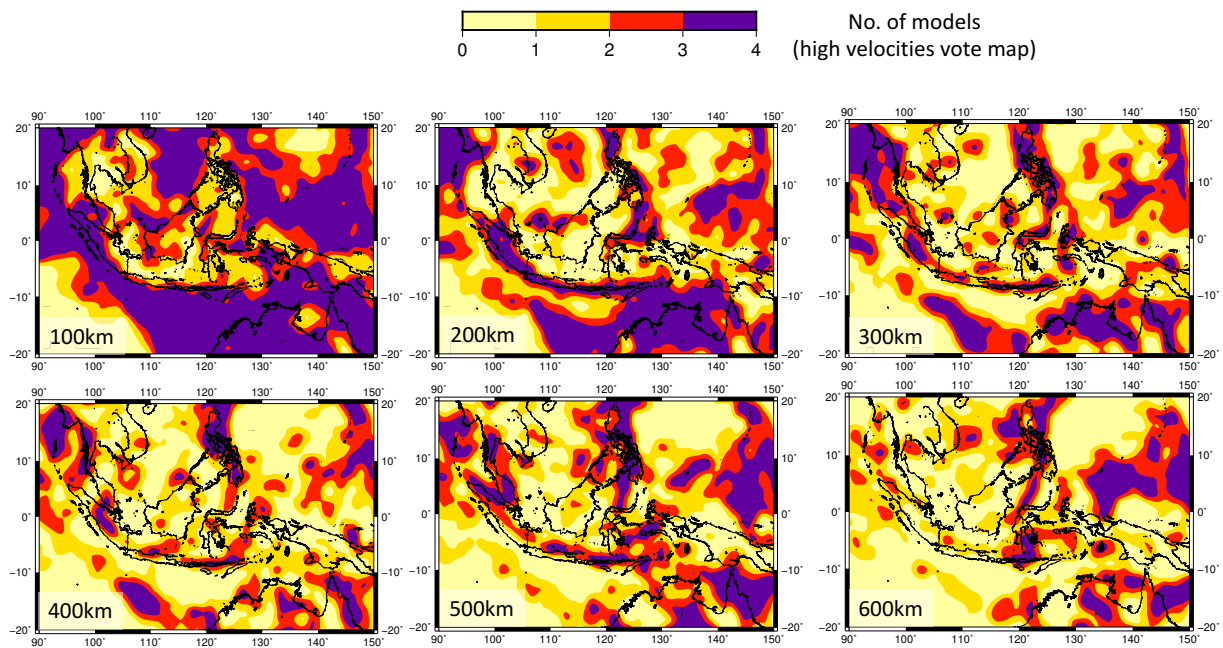


Figure 9: Votemap for all four models produced by the different P-wave datasets (P\_A, P\_B, P\_C, P\_D). High velocities ( $dv/v > 0$ ) were assigned the value 1 for each model. Thus, in places where all the models exhibit a positive perturbation, the votemap has a value of 4, and where all the models exhibit a negative perturbation, the votemap has a value of zero.

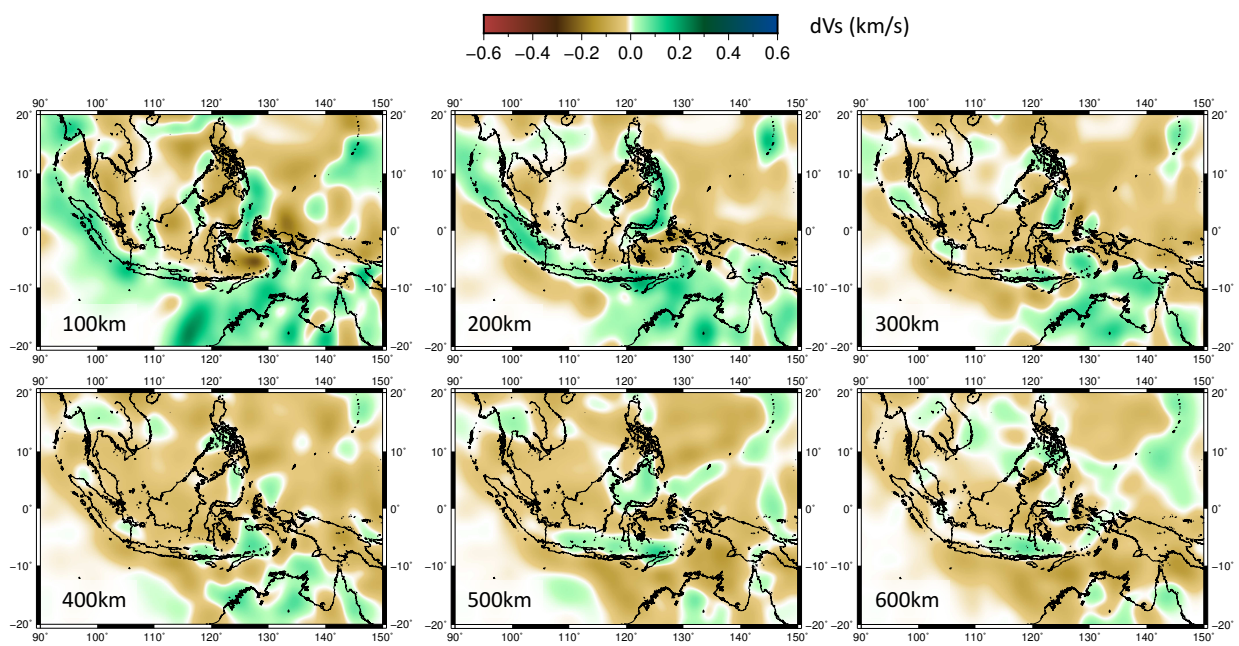


Figure 10: S-wave tomographic model using ISC-EHB (S\_A) dataset from 100-600 km depth.

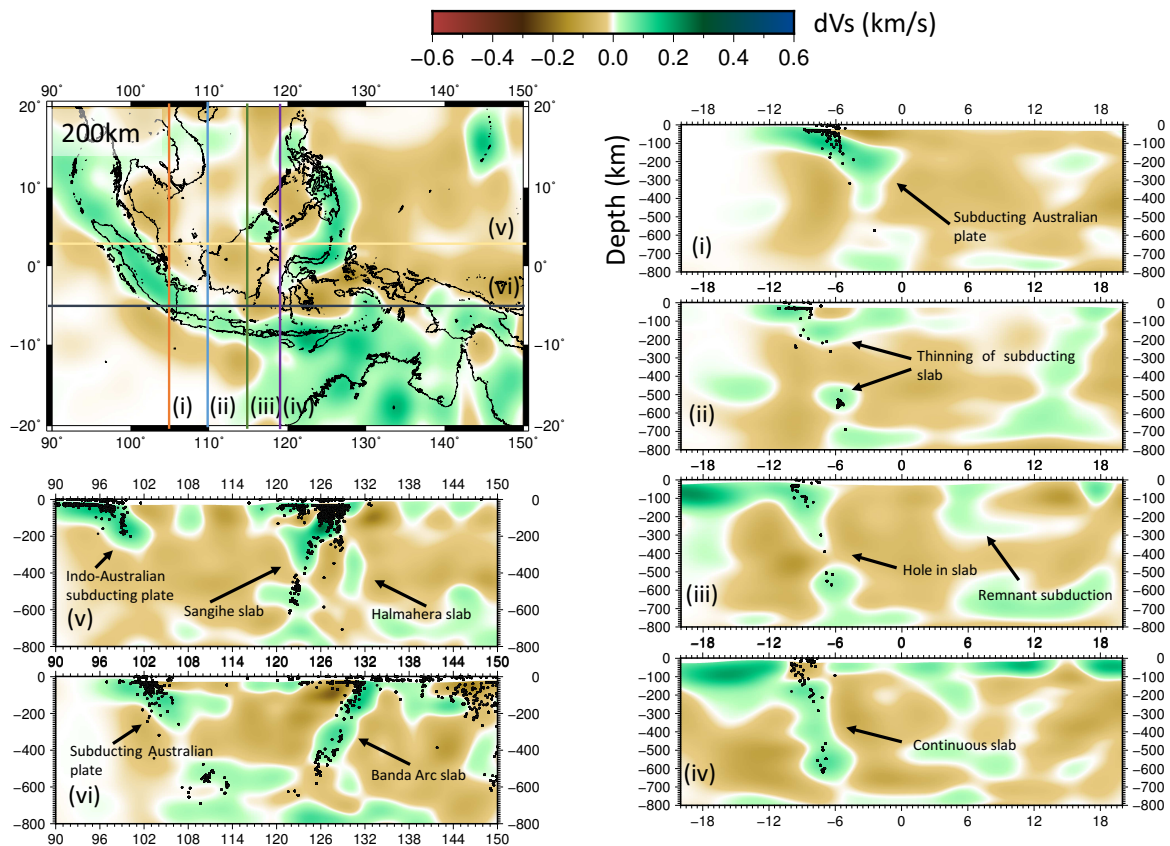


Figure 11: Cross-sections through the S-wave model based on the ISC-EHB (S\_A) dataset. The S-wave dataset has a much lower number of arrivals than P, leading to lower resolution. Features of interest are outlined-marked with an arrow and labelled in the cross-sections through the model: (i) subduction along Sumatra, (ii) slab at beneath Java region, (iii) major hole in the slab near 400 km depth, (iv) slab east of Java with no evidence of small hole, (v) Sangihe and Halmahera arc-arc collision and the start of the subducting slab at the in northwestern part of Sumatra, (vi) curved subduction near Banda arc and subduction zone in Sumatra near the tear-slab break as discussed in section 14.3.2.