Passive source reverse time migration based on the spectral element method

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Abstract

Increasing deployment of dense arrays has facilitated detailed structure imaging for tectonic investigation, hazard assessment and resource exploration. Strong velocity heterogeneity and topographic changes have to be considered during passive source imaging. However, it is quite challenging for ray-based methods, such as Kirchhoff migration or the widely used teleseismic receiver function, to handle these problems. In this study, we propose a 3-D passive source reverse time migration strategy based on the spectral element method. It is realized by decomposing the time reversal full elastic wavefield into amplitude-preserved vector P and S wavefields by solving the corresponding weak-form solutions, followed by a dot-product imaging condition to get images for the subsurface structures. It enables us to use regional 3-D migration velocity models and take topographic variations into account, helping us to locate reflectors at more accurate positions than traditional 1-D model-based methods, like teleseismic receiver functions. Two synthetic tests are used to demonstrate the advantages of the proposed method to handle topographic variations and complex velocity heterogeneities. Furthermore, applications to the Laramie array data using both teleseismic P and S waves enable us to identify several south-dipping structures beneath the Laramie basin in southeast Wyoming, which are interpreted as the Cheyenne Belt suture zone and agree with, and improve upon previous geological interpretations.

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¹⁰ Key Points:

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11	• A 3-D passive source reverse time migration based on the spectral element method
12	is proposed to image complex structures.
13	• Amplitude-preserved vector P and S wavefields are accurately decomposed by solv-

- ¹⁴ ing corresponding weak-form solutions.
- Several south-dipping structures can be identified from P and S migration results
 beneath the Laramie basin, which are interpreted as the Cheyenne Belt suture zone.

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17 Abstract

Increasing deployment of dense arrays has facilitated detailed structure imaging 18 for tectonic investigation, hazard assessment and resource exploration. Strong velocity 19 heterogeneity and topographic changes have to be considered during passive source imag-20 ing. However, it is quite challenging for rav-based methods, such as Kirchhoff migration 21 or the widely used teleseismic receiver function, to handle these problems. In this study, 22 we propose a 3-D passive source reverse time migration strategy based on the spectral 23 element method. It is realized by decomposing the time reversal full elastic wavefield into 24 amplitude-preserved vector P and S wavefields by solving the corresponding weak-form 25 solutions, followed by a dot-product imaging condition to get images for the subsurface 26 structures. It enables us to use regional 3-D migration velocity models and take topo-27 graphic variations into account, helping us to locate reflectors at more accurate positions 28 than traditional 1-D model-based methods, like teleseismic receiver functions. Two syn-29 thetic tests are used to demonstrate the advantages of the proposed method to handle 30 topographic variations and complex velocity heterogeneities. Furthermore, applications 31 to the Laramie array data using both teleseismic P and S waves enable us to identify sev-32 eral south-dipping structures beneath the Laramie basin in southeast Wyoming, which 33 are interpreted as the Cheyenne Belt suture zone and agree with, and improve upon pre-34 vious geological interpretations. 35

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Plain Language Summary

Increasing deployment of dense arrays has allowed detailed structure imaging for 37 tectonic investigation, hazard assessment and resource exploration. However, traditional 38 ray-based migration methods or 1-D velocity model-based receiver function methods may 39 greatly degrade the imaging quality without considering the full wavefield propagation 40 effect and velocity heterogeneities. Therefore, in this study, we develop a 3-D passive source 41 migration based on the spectral element method, which is capable of handling topographic 42 variations as well as complex velocity heterogeneities for real data applications. Several 43 synthetic tests and applications to the Laramie array data using both teleseismic P and 44 S waves are used to demonstrate the capability of our method to image complex struc-45 tures, such as subduction and suture zones. 46

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47 **1** Introduction

In past decades, teleseismic receiver functions have been widely used for imaging 48 discontinuities within the Earth's crust and upper mantle (Burdick & Langston, 1977; 49 Langston & Corvallis, 1977; Vinnik, 1977; Ai et al., 2007; Kawakatsu & Watada, 2007; 50 Kind et al., 2012; H. Zhang et al., 2016; Long et al., 2019). Based on the assumption of 51 a 1-D Earth model where discontinuities are horizontal, common convert point stack-52 ing (Dueker & Sheehan, 1997; L. Zhu, 2000; Gilbert et al., 2003) has generally been used 53 to improve the image quality by mapping the receiver functions to depths. In the case 54 of a single station, like the SEIS on Mars (Banerdt et al., 2020; Knapmeyer-Endrun et 55 al., 2020; Lognonné et al., 2020) or sparse station distributions such as the USArray (Meltzer 56 et al., 1999), such simplifications are natural and useful. However, it indeed degrades the 57 image quality of geologically complex structures, such as subsurface environments with 58 steep faults and laterally discontinuous interfaces, without considering the effects of wave-59 field diffraction and scattering caused by lateral variations of impedance contracts and 60 velocities (L. Chen et al., 2005; Shang et al., 2012). In addition, increasing deployment 61 of dense and nodal arrays has facilitated detailed structure imaging for tectonic inves-62 tigation, hazard assessment and resource exploration (Okada et al., 2004; Zheng et al., 63 2008; Nábělek et al., 2009; H. Zhang et al., 2016; X. Yang et al., 2017; X. Chen et al., 64 2018; Li et al., 2018; Clayton et al., 2019; X. Wang et al., 2021; Onyango et al., 2022; 65 Wu et al., 2023). Therefore, it calls for advanced imaging methods, such as seismic mi-66 gration/inversion, which exploit the full complexity of recorded wavefields and rely less 67 on a priori information about the Earth's structures, to get accurate images of complex 68 subsurface discontinuities (Sheehan et al., 2000; L. Chen et al., 2005; Shang et al., 2012; 69 Cheng et al., 2016; Li et al., 2018; Millet et al., 2019). 70

Among many seismic migration approaches, Kirchhoff migration/inversion (Schneider, 71 1978; Gray & May, 1994) has been widely used in the oil and gas industry to image shal-72 low sedimentary structures owing to its high efficiency and simplicity. It was later in-73 troduced to global seismology for imaging large-scale structures using multi-component 74 teleseismic data recorded by local dense arrays (Burridge et al., 1998; Bostock, 1998; Bo-75 stock & Rondenay, 1999; Rondenay et al., 2000; Bostock et al., 2001; Bostock, 2002; Levan-76 der et al., 2005; Rondenay, 2009; Liu & Levander, 2013; Cheng et al., 2016; Millet et al., 77 2019). Teleseismic Kirchhoff migration/inversion is implemented by weighting and stack-78 ing data along diffraction hyperbola for every possible scattering point in a regular grid 79

(Schneider, 1978; Gray & May, 1994; Bostock et al., 2001; Cheng et al., 2016; Millet et 80 al., 2019). To date, most Kirchhoff migration/inversion approaches are based on the in-81 finite high-frequency assumption by neglecting the finite frequency property of seismic 82 wave propagation, which enables us to efficiently get images by using ray tracing (Richards 83 & Aki, 1980) or solving the Eikonal equation (Gray & May, 1994; Cheng et al., 2016). 84 However, It is difficult for ray-tracing to handle multiple arrivals, shadow zones, and even 85 chaotic rays in complex subsurface environments (Audebert et al., 1997; L. Chen et al., 86 2005). Although the Eikonal equation-based Kirchhoff migration is capable of dealing 87 with complex structures, it is not easy to handle the multipathing challenge (H. Zhao, 88 2005; Waheed et al., 2015; Tong, 2021). Furthermore, Kirchhoff migration requires more 89 efforts to correct image amplitudes, and therefore, the wave-equation Kirchhoff migra-90 tion has been developed to balance accuracy and efficiency (Andrade et al., n.d.; Jin & 91 Etgen, n.d.). To mitigate the finite-frequency, multi-arrivals and inaccurate amplitude 92 problems, Gaussian beam migration (Hill, 1990, 2001) has been proposed to apply lo-93 cal slant stacks by using complex-valued traveltimes and amplitudes. These complex quan-94 tities come from the approximation of seismic wave propagation with a sum of Gaussian 95 beams, which are finite-frequency, ray-theoretical approximations to the wave equation 96 (Notfors et al., n.d.). It is further extended to the wave-equation-based two-way beam 97 wave method, which has comparable accuracy and efficiency in comparison to the wave-98 equation-based methods (J. Yang et al., 2022). 99

On the other hand, wave equation-based migration methods can essentially avoid 100 most difficulties from the ray-based method by using the full wavefields (Claerbout, 1985). 101 Among many different types of wave equation-based migration methods, synthetic and 102 field experiments have shown that reverse-time migration (RTM) (Baysal et al., 1983; 103 Whitmore, 1983; McMechan, 1983) has better performance for imaging complex struc-104 tures. Active source RTM includes a forward wavefield modeling from the source and 105 a backward wavefield modeling from the receivers, followed by applying imaging condi-106 tions to these two wavefields. In contrast, passive source RTM only requires one back-107 ward wavefield modeling for each earthquake. Therefore, inaccuracies due to source lo-108 cations and velocity variations between earthquake locations and local study regions could 109 be neglected. During the wavefield back-propagation, P and S wavefields are usually sep-110 arated by polarization decomposition (Shang et al., 2012; H. Zhu, 2017; J. Yang et al., 111 2018). Then, the imaging condition is applied to cross-correlate separated P and S wave-112

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fields, followed by the summation of individual events. As pointed out by Shang et al. 113 (2012), passive source RTM of converted waves differs fundamentally from single station 114 receiver function analysis, and also in several important ways from more traditional re-115 ceiver function migration (L. Chen et al., 2005). However, both current 2-D and 3-D pi-116 oneer works are based on the finite difference method, which requires projecting station 117 locations on the model grids, and it is not easy to handle complex topography in real 118 applications (Rajasekaran & McMechan, 1995; Bevc, 1997; Shragge, 2014; Shang et al., 119 2017; Yi et al., 2019). In addition, there are few real data applications for 3-D passive 120 source RTM to image crustal and uppermost mantle structures. 121

In this study, we combine passive source RTM with the spectral-element method 122 (SEM) (Komatitsch & Vilotte, 1998; Fichtner et al., 2009). Owning to its advantages 123 in handling complex topographic variations, wavefield coupling between different media, 124 flexibility to handle station distributions and so on, the SEM has been widely applied 125 to construct crustal and mantle velocity models at regional and global scales (Tromp et 126 al., 2005; Fichtner et al., 2009; Tape et al., 2009; Fichtner, 2010; Liu & Levander, 2013; 127 H. Zhu et al., 2012; Y. Wang et al., 2016; K. Wang et al., 2020, 2021; Tromp, 2020; H. Zhu 128 et al., 2020; Maguire et al., 2022). Here, we use it for passive source RTM. Because the 129 wave-equation is solved in a weak-form fashion in SEM, directly taking strong-form spa-130 tial derivatives for wavefield decomposition results in significant discontinuities between 131 element boundaries. To solve this problem, we propose to solve the corresponding weak-132 form solutions for P and S wavefield decomposition. This paper is organized into the fol-133 lowing parts, we first briefly review passive source RTM based on the wavefield decom-134 position, followed by the theory of weak-form solutions for wavefield decomposition, and 135 then several synthetic tests are used to demonstrate the advantages of the proposed method. 136 Finally, we apply it to the Laramie array for suture zone imaging. 137

- 138 2 Methodology
- 130

2.1 Passive source RTM based on vector wavefield decomposition

Based on the principle that all back-propagating direct P and S receiver wavefields $(\mathbf{u}^P \text{ and } \mathbf{u}^S)$ should coincide at "source" locations, wave-equation-based passive source imaging seeks strong energy in space as the source locations (McMechan, 1982; Nakata & Beroza, 2016; J. Yang et al., 2020; Duan et al., 2021). Similarly, the vector \mathbf{u}^P and u^S wavefields constructed by back-propagating multichannel direct P and Ps converted
 (or direct S and Sp converted) seismograms, should coincide at the scattering locations
 in depths, such as the Moho (Shang et al., 2012; Li et al., 2018). The imaging condition
 can be given as

$$I(\mathbf{x}) = \sum_{i_e=1}^{Ne} \int_{t=0}^{T} \boldsymbol{u}(\mathbf{x})_{i_e}^{P}(t) \cdot \boldsymbol{u}(\mathbf{x})_{i_e}^{S}(t) dt \quad ,$$
(1)

where $I(\mathbf{x})$ is the image, $\boldsymbol{u}(\mathbf{x})_{i_e}^P$ and $\boldsymbol{u}(\mathbf{x})_{i_e}^S$ are the constructed vector P and S wavefields for i_e th event at the spatial coordinate \mathbf{x} . N_e is the total number of events for imaging. t represents the current time and T denotes the total time for wavefield back-propagation. " \cdot " denotes the dot product between two vector fields. The final image is obtained by summing over all the time steps during wavefield back-propagation and all seismic events.

Because the vector \mathbf{u}^P and \mathbf{u}^S wavefields are coupled during elastic wavefield simulations, one may consider back-propagate multichannel P and Ps (or S and Sp) seismograms independently by solving two acoustic wave-equations (Sun et al., 2004; Duan et al., 2021) or decouple them during the wavefield back-propagation. H. Zhu (2017) proposed an amplitude-preserved wavefield decomposition method to decouple the vector \mathbf{u}^P and \mathbf{u}^S from elastic wavefield \mathbf{u} by introducing a vector \mathbf{w} in the form of

$$\nabla^2 \mathbf{w} = \mathbf{u} = \mathbf{u}^P + \mathbf{u}^S \quad , \tag{2}$$

¹⁵⁹ where vector P and S wavefields can be obtained according to

$$\mathbf{u}^P = \nabla (\nabla \cdot \mathbf{w}), \quad \mathbf{u}^S = -\nabla \times \nabla \times \mathbf{w} \quad . \tag{3}$$

Here ∇ , ∇ and ∇ are the gradient, divergence and curl operators, respectively. How-

ever, solving Equation 3 requires explicitly solving the Poisson's equation in Equation

¹⁶² 2. To reduce the computational costs, J. Yang et al. (2018) further extended this method

to avoid solving Poisson's equation under the isotropic media assumption with

$$\mathbf{u}^P = \nabla (\nabla \cdot \alpha^2 \hat{\mathbf{u}}), \quad \mathbf{u}^S = -\nabla \times (\nabla \times \beta^2 \hat{\mathbf{u}}) \quad , \tag{4}$$

where $\hat{\mathbf{u}}$ denotes the extrapolated wavefield excited by a double integral of the original seismograms, α and β are the local P and S velocities, respectively.

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2.2 Strong- and weak-form solutions for the wavefield decomposition

¹⁶⁷ In this section, we introduce H. Zhu (2017) and J. Yang et al. (2018)'s method to ¹⁶⁸ the SEM, so that it is more convenient to handle irregular surface topography and sta¹⁶⁹tion locations. Different from the finite difference (FD) method, SEM seeks a weak so-

- 170 lution $\bar{\boldsymbol{u}}(\mathbf{x},t)$ to the equation of motion in the Galerkin sense (Komatitsch & Vilotte,
- ¹⁷¹ 1998; Fichtner, 2010). Lagrange polynomials collocated at the Gauss–Lobatto–Legendre(GLL)
- points are used to interpolate the wavefield at any spatial locations in the form of

$$\boldsymbol{u}(\mathbf{x},t) \approx \bar{\boldsymbol{u}}(\mathbf{x},t) = \sum_{i,j,k=1}^{N+1} \boldsymbol{u}^{ijk}(t) L_{ijk}(\mathbf{x}) \quad , \tag{5}$$

where $\boldsymbol{u}(\mathbf{x},t)$ and $\bar{\boldsymbol{u}}(\mathbf{x},t)$ represent the true and weak-form solutions of the wave-equation, respectively. \mathbf{x} represents the spatial coordinate and t represents time. N denotes the degree of the Lagrange polynomials, and therefore we have N + 1 GLL points for one element along each direction. $L_{ijk} = l_i(x)l_j(y)l_k(z)$ represents the basis function chosen as the product of three Lagrange polynomials l(x), l(y) and l(z) along the x, y and z directions collocated at the corresponding GLL points.

Owing to the continuous prosperities of the Lagrange polynomials, a straightforward solution of Equation 4 based on SEM, noted as the strong-form solution, can be realized by

$$\mathbf{u}^{P} = \nabla [\nabla \cdot \alpha^{2} \sum_{i,j,k=1}^{N+1} \boldsymbol{u}^{ijk}(t) L_{ijk}(\mathbf{x})], \quad \mathbf{u}^{S} = -\nabla \times [(\nabla \times \beta^{2} \sum_{i,j,k=1}^{N+1} \boldsymbol{u}^{ijk}(t) L_{ijk}(\mathbf{x})] \quad , \quad (6)$$

which is fairly easy to implement because the spatial derivatives from the gradient, di-182 vergence and curl operators are taken from the Lagrange polynomials. However, we ex-183 pect quite strong artifacts in the separated vector wavefields, which could contaminate 184 the final imaging result. As illustrated in Figure 1, strong S mode artifacts appear in the 185 P mode. This is because the solution of SEM is a weak-form approximation, and although 186 the wavefield is continuous cross elements, the continuity of spatial derivatives is not guar-187 anteed. To mitigate this leakage during wavefield decomposition, we use the weak so-188 lutions for Equation 4 as follows: 189

$$\int_{G_e} \mathbf{\Phi} \cdot \mathbf{u}^P d^3 \mathbf{x} = \int_{G_e} \mathbf{\Phi} \cdot \nabla (\nabla \cdot \alpha^2 \mathbf{u}) d^3 \mathbf{x} = -\int_{G_e} [\nabla \cdot \mathbf{\Phi}] [\nabla \cdot (\alpha^2 \mathbf{u})] \quad , \tag{7}$$

$$\int_{G_e} \mathbf{\Phi} \cdot \mathbf{u}^S d^3 \mathbf{x} = -\int_{G_e} \mathbf{\Phi} \cdot \nabla \times (\nabla \times \beta^2 \mathbf{u}) d^3 \mathbf{x} = \int_{G_e} [\nabla \times \mathbf{\Phi}] \cdot [\nabla \times \beta^2 \mathbf{u}] d^3 \mathbf{x} \quad , \quad (8)$$

where Φ denotes any arbitrary, differentiable, time-independent test function, which will be chosen as the basis functions similar to the way we get SEM solutions. As shown in Figure 2, both P and S modes are clearly separated without significant leakages. This indicates that the weak-form solutions can significantly improve the quality of wavefield decomposition, which is the basis for the following imaging tests. The deviations of Equations 7 and 8 can be found in the supplementary material. 197

2.3 The workflow for passive source RTM

The implementation of passive source RTM can be summarised in the following four steps:

200	1.	Teleseimic data preprocessing. During this step, we estimate source time functions
201		for each teleseismic event based on the principle component analysis for the aligned
202		incident waves (Bostock & Rondenay, 1999; Shang et al., 2017), such as P, S or
203		SKS waves. For this purpose, the incident waves are aligned according to the on-
204		set time of the incident phases estimated using the TauP package (Crotwell et al.,
205		1999). Then, an iterative time deconvolution (Kikuchi & Kanamori, 1982) is ap-
206		plied to each component of original (unaligned) data to remove the source time func
207		tion, followed by bandpass filtering for later finite frequency simulations.
208	2.	Injecting the preprocessed teleseismic data for the SEM solver as adjoint sources
209		to the acceleration vector. Given the linear relation between the source time func-
210		tion and wavefield, this step further avoids obtaining the double integral of the
211		teleseismic data, which should be injected to construct the second time integral
212		of the displacement wavefield indicated by Equation 4.
213	3.	During the reverse time wavefield propagation, solving the weak-form solutions
214		shown in Equations 7 and 8 to get vector wavefields $\mathbf{u}_{i_e}^P$ and $\mathbf{u}_{i_e}^S$, followed by ap-
215		plying the zero-lag cross-correlation imaging condition (Equation 1), noted as $\mathbf{I}_{ie}^{PS}.$
216		To save computational costs, one may consider performing wavefield separation
217		for every several time steps according to the Nyquist rule. We also apply the imag-
218		ing condition to $\mathbf{u}_{i_e}^P$ with $\mathbf{u}_{i_e}^P$ to get the PP image as $\mathbf{I}_{i_e}^{PP}$.

- 4. Sum all corresponding event images to obtain \mathbf{I}_{PS} and \mathbf{I}_{PP} , and then get the image with $\mathbf{I}_{PS} = \frac{\mathbf{I}_{PS}}{\mathbf{I}_{PP}+\epsilon}$ to compensate for energies at great depths. Here, ϵ is a small damping value used to avoid dividing by zeros.
- 222 **3** Applications

In this section, we use two synthetic tests and one real data example to validate our method based on linear array recordings although we are using 3-D modeling and migration. It is well-known that station spacing together with the data frequency are important factors that may result in spatial aliasing during passive source RTM (Gray, 2013; Shang et al., 2012). Therefore, most teleseismic migration methods used to image

lithospheric structures rely on dense arrays (station spacing is ~ 5 km) (Bostock, 1998; 228 Rondenay et al., 2000; Rondenay, 2009; Shang et al., 2017). Therefore, all synthetic sim-229 ulations in our tests are implemented using a 3-D plane-wave injection method, FK-SEM 230 (Tong, Komatitsch, et al., 2014; Tong, Chen, et al., 2014) and recorded by a dense lin-231 ear array. Then, passive source RTM is implemented based on 3-D migration. We note 232 here, there are mainly three advantages to implement 3-D rather than 2-D migration even 233 using recordings from a linear array: 1) no 3-D/2-D transformation is needed ("Line-source 234 simulation for shallow-seismic data. Part 1: Theoretical background, author=Forbriger, 235 Thomas and Groos, Lisa and Schäfer, Martin", 2014; Schäfer et al., 2014; C. Zhang et 236 al., 2018), which is important to correct phase and amplitude differences between 3-D 237 and 2-D simulations; 2) more teleseismic events can be chosen for real data applications; 238 3) no needs to project stations onto a line, which is required for 2-D migration even with 239 linear arrays because their locations are not critically located on 2-D linear mesh grid 240 (Shang et al., 2017). 241

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3.1 Synthetic test 1: effects of topographic variations

The first synthetic test is used to highlight the effects of topographic variations on 243 passive source migration when using high-frequency teleseismic data. The input model 244 is a two-layer model with an interface at 30 km depth. The density, Vp and Vs for the 245 first layer between 0 and 30 km depth are 2.72 g/cm^3 , 5.8 km/s and 3.46 km/s, which 246 are 3.423 g/cm^3 , 8.06 km/s and 4.53 km/s between 30 and 60 km depths. These values 247 are taken from the AK135 model (Kennett et al., 1995). The simulation domain spans 248 -90 to 90 km along the X (longitude) direction, -60 to 60 km along the Y (latitude) di-249 rection and 0 to 60 km along the depth direction. The element size is 1.0 km, which al-250 lows for the wavefield simulation with the highest frequency around 3.0 Hz. The Stacey 251 absorption boundary condition is used for each face of the model except for the top, which 252 is a free-surface boundary condition for our simulations (Mahrer, 1990). A 1.5-D topo-253 graphic change (no change along the Y direction) following the solid line shown in Fig-254 ure 4b is designed to validate its effects on passive source RTM. We note here, the model 255 will be stretched in depth direction within the SPECFEM package to handle the topo-256 graphic changes. 161 stations with a horizontal station spacing of 1.0 km are distributed 257 on the surface of the model. We use a Ricker source wavelet with a peak frequency of 258 1.0 Hz to initialize the incident plane waves. 12 P-wave plane waves, with incident an-259

gles ranging from $12^{\circ}-27^{\circ}$, are injected into the local simulation domain to simulate the 260 teleseismic waves. The back-azimuthal angles for events 1, 3, 7, 9 and 11 are 90° and 270° 261 for the rest events. This makes a symmetric coverage of plane waves propagating through 262 the model. The corresponding teleseismic data are recorded by the station for each event. 263 Seismograms for the first teleseismic event with an incident angle of 27° and the back-264 azimuthal angle of 90° are shown in Figure 3. By using the short-time average over long-265 time average (STA/LTA) algorithm (Withers et al., 1998), it is feasible to detect the on-266 set of the incident wavefield. The magenta lines in Figure 3 denote the time windows used 267 in our imaging, which are 5 seconds before and 6 seconds after the onset of the incident 268 P waves. It helps us to isolate multiple reflections that may distort our imaging results. 269 We speculate that the onset of incident P waves in both Z and X components shows the 270 imprint of the topographic changes. Without considering the topographic changes, i.e., 271 injecting teleseismic data at 0 km elevation of each station, a significant imprint of to-272 pography can be observed in the imaging result in Figure 4a, which looks like a mirror 273 of the topography. On the contrary, the reflector can be accurately imaged by injecting 274 the teleseismic data at the right locations by considering the topographic changes (Fig-275 ure 4b). 276

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3.2 Synthetic test 2: imaging subduction slabs

In this synthetic test, we validate the capability of our method to image complex 278 structures in subduction zones. The background model is a two-layer crust and mantle 279 structure, with density, Vp and Vs, the same as the previous example. A 3-D subduct-280 ing slab (Figure 6a-b) is designed between -60 to 60 km along the X (Longitude) direc-281 tion, -30 to 30 km along the Y (Latitude) direction and 40 to 140 km in depth with a 282 thickness of 50 km. The Moho depth is 35 km. The model parameters (density, Vp and 283 Vs) of the subducting slab are 12% greater than the background model. Our simulation 284 domain spans from -120 to 120 km along the X direction, -48 to 48 km along the Y di-285 rection and 0 to 160 km along the depth direction. We use 201 stations with 1.0 km sta-286 tion spacing evenly distributed from -100.0 km to 100.0 km. 12 teleseismic P wave events 287 with the same back-azimuth and incident angles as the previous example are injected 288 into the local study region, starting from an initial depth at 240 km. The rest param-289 eters are the same as the previous synthetic test. Synthetic Z and X component seismo-290 grams for the first teleseismic event are illustrated in Figures 5a and b. Because the back-291

azimuth angle of this event is 90°, the X components are equal to the radial components,
which could be used to investigate Ps-converted waves. Due to the complexity of the velocity model, we speculate several kinds of multiple reflections (Moho and slab-related)
for the P waves in Z components. The corresponding Ps-converted waves can be investigated from X components in Figure 5b. Interestingly, despite being weaker than Mohorelated converted waves (S), slab-related as well as multiple reflection-related S-converted
waves can also be seen.

Our RTM is implemented using a smoothed two-layer background model (Figures 299 6 c-d) to avoid artifacts arising from sharp velocity interfaces during the wavefield sep-300 aration (J. Yang et al., 2018). It is smoothed vertically using a Gaussian function with 301 a radius of 5.0 km. This means we do not include the slab in our migration velocity model, 302 which is close to the real cases. After stacking and illuminating, the interfaces of the slab 303 are well imaged as displayed in Figure 6g. It suggests that our method is capable of imag-304 ing complex structures, such as velocity anomalies with high dipping angles. Because it 305 is quite difficult to isolate multiple reflections in this case, we also see different-order mul-306 tiple imaging artifacts in our stacked result. We also show the imaging result from the 307 first four teleseismic events, which also successfully captures the main features of the slab 308 and the Moho, but with stronger artifacts due to not enough stacking. In addition, the 309 images along the Y (Latitude) direction (Figures 6f and h), which is perpendicular to 310 our stations' distribution, indicate that with only several effective stations, the migra-311 tion will map the converted waveforms along the isochrone interfaces in depths (Schneider, 312 1978). Because we only use one linear array across the slab, it is necessary to consider 313 the contributions of teleseismic events with back-azimuthal angles away from the array 314 direction. We test another three back-azimuthal angle pairs, which are $0^{\circ}/180^{\circ}$, $45^{\circ}/225^{\circ}$ 315 and $135^{\circ}/315^{\circ}$ with the same incident angle as the previous test. Figure 7 shows the mi-316 gration results from the back-azimuthal angle pairs of $0^{\circ}/180^{\circ}$ and $45^{\circ}/225^{\circ}$. The mi-317 gration result from $135^{\circ}/315^{\circ}$ azimuthal-angle pair is similar to the one from $45^{\circ}/225^{\circ}$. 318 As expected, when the azimuthal angles are away from the linear array direction, the 319 interfaces of the imaged slab and the Moho become weaker and more incoherent arti-320 facts become stronger. Interestingly, the Moho interface disappears when using teleseis-321 mic events with the $0^{\circ}/180^{\circ}$ azimuthal pair. This could be explained by the concept of 322 the Fresnel zone. Bostock (1998) pointed out that the Fresnel zone of the scattering points 323 depends on the dominant frequencies and depths of the interface. Assuming that con-324

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structive interference arises for phase differences less than a quarter period, the diam-325 eter of the Fresnel zone around the interface varies between ~ 25 km for $P_m S$ (Moho con-326 verted wave) at 0.5 Hz to ~160 km for $P_{660}S$ (generated at the 660 km discontinuity) 327 at 0.25 Hz. This means the converted waves recorded at stations that arise from deeper 328 interfaces (e.g., X=0, Y=0 and Z=660 km) could be the response to a larger range of 329 interfaces, e.g., X=[-80, 80], Y=[-80, 80], Z=[580, 740] km. We note here, the estima-330 tions of X and Y ranges are fairly rough. For teleseismic events with back-azimuthal an-331 gles away from the linear array, their Fresnel zone is also away from the structures be-332 neath the linear array. As a consequence of small Fresnel zones, shallow interfaces, such 333 as the Moho, gradually disappear beneath the linear array due to incoherence stacking 334 from different events. Therefore, reasonable good station, incident angle and azimuthal 335 angle coverages are preferred for our method if we intend to image the interfaces beneath 336 a linear array (Shang et al., 2012; Li et al., 2018). 337

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3.3 Imaging the Cheyenne Belt suture zone (CBSZ) using the Laramie array

340

3.3.1 Geological setting

The CBSZ is a tectonic suture zone between the Archean Wyoming craton to the 341 north and the Paleoproterozoic Yavapai province to the south (Karlstrom & Houston, 342 1984; Sims & Stein, 2003; Hansen & Dueker, 2009; Jones et al., 2010). It contains a set 343 of steeply south-dipping shear zones formed during the 1.78–1.75 Ga Medicine Bow orogeny 344 when the Proterozoic Green Mountain arc collided with the passive margin of the Wyoming 345 craton via south-facing subduction (Tyson et al., 2002; Hansen & Dueker, 2009). Steep 346 stretching lineations and shear-sense features indicate south-side-up motion (Tyson et 347 al., 2002; Hansen & Dueker, 2009). To better constrain the structure of the CBSZ, the 348 Laramie array was deployed within the Laramie basin across the inferred trace of the 349 Cheyenne belt (Figure 8) (Hansen & Dueker, 2009). It is a dense 80 km long linear ar-350 ray with broadband seismometers, which consists of 30 sensors spaced 2.2 km apart and 351 was deployed for a period of eight months in 2000–2001. Based on P and S receiver func-352 tions together with teleseismic P wave traveltime tomography, Hansen and Dueker (2009) 353 found an imbricated Moho north of the Cheyenne belt. It is basically consistent with the 354 interpretation of seismic results from the CDROM (Continental Dynamics Rocky Moun-355 tain) project (Tyson et al., 2002). However, either due to the limited aperture of the Laramie 356

array, or the methodologies, the CBSZ hasn't been clearly imaged as shown in Figure 357

7 of Hansen and Dueker (2009). Ruigrok et al. (2010) used seismic interferometry to ex-358

tract reflection responses from the coda of the transmitted energy from distant earth-359

- quakes, where they found discontinuities in their migration images, which were interpreted 360
- as the CBSZ. Here, we use our passive source RTM method to further investigate the 361
- detailed shape of the CBSZ with converted teleseismic P and S waves. 362
- 3.3.2 Data 363

364

The teleseismic P wave dataset is constructed from 11 events at $30^{\circ}-90^{\circ}$ distance with body-wave magnitudes greater than 5.5. The S wave dataset is selected from 1 S-365 wave and 7 SKS events from $55^{\circ}-85^{\circ}$ and $85^{\circ}-120^{\circ}$ epicentral distances, respectively (Wilson 366 & Aster, 2005; Yuan et al., 2006). Detailed information about these events can be found 367 in Table 1 and displayed in Figure 9. The data is selected, downloaded and preprocessed 368 with the standing order for data package (SOD) (Owens et al., 2004). For each P wave 369 event, three-component waveforms within the time window of two minutes before and 370 three minutes after direct P arrivals predicted by the AK135 model (Kennett et al., 1995) 371 are collected. The north-south and west-east component seismograms are rotated to ra-372 dial and transverse components after removing instrument response, linear trend, and 373 mean values, followed by a bandpass filter of 1-20 s. Then, the preprocessed three-component 374 waveforms of each event are visually inspected, and only those with a signal-to-noise ra-375 tio (SNR) larger than 3.0 and 2.0 for vertical and radial components are kept. Afterward, 376 we use the open-source software AIMBAT (Lou et al., 2013) to remove bad seismograms 377 with spurious amplitudes and cross-correlation coefficients lower than 0.80 for vertical 378 and radial components. To avoid spatial aliasing, only events with more than 15 seis-379 mograms are remained. The data selection process for S and SKS events is similar ex-380 cept: (1) the time window is defined as two minutes before and three minutes after in-381 cident S arrivals; (2) seismograms with SNR larger than 2.0 and 3.0 for vertical and ra-382 dial components are kept. This is because S waves are mainly in radial components. In 383

the end, most events have more than 25 high-quality seismograms for each component. 384

Gray (2013) and Li et al. (2018) suggested that station spacing is an important fac-385 tor for spatial aliasing during passive source RTM. For example, 5 km station spacing 386 will result in slightly aliased P-wave migration with 1 Hz data, given the incident an-387 gles range from 12 ° to 27°. Therefore, we interpolate those deleted bad seismograms 388

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using 2-D cubic-spline interpolation (J. Zhang & Zheng, 2015). The 2-D cubic-spline in-389 terpolation was originally used to refine the receiver functions from sparse station dis-390 tributions. Here, we use it to interpolate the aligned vertical and radial component seis-391 mograms. It mainly includes three steps: (1) align the seismograms according to the on-392 set of incident waves (P or S), predicted by the reference model, such as AK135 (Kennett 393 et al., 1995). The reference time could be further adjusted by applying a multichannel 394 cross-correlation algorithm (VanDecar & Crosson, 1990); (2) perform 2-D interpolation 395 for each time step using the cubic-spline method for each component; (3) shift the in-396 terpolated data back according to the reference time. We note here, the absolute am-397 plitudes are kept during interpolation for each channel and component, which is differ-398 ent from J. Zhang and Zheng (2015). This is important for dealing with multi-component 399 data. As compared in Figure 9, the interpolated seismograms follow the same trend of 400 nearby stations in great details for both P and S waves. We use 10 s before and 30 s af-401 ter the onset of P and 30 s before and 30 s after the onset of S waves for the following 402 migration. 403

404

3.3.3 Migration

Our computational domain for wavefield propagation spans from -106.25° to -105.30° 405 along longitude, 40.85° to 41.75° along latitude, and -10 to 110 km in the depth direc-406 tion. We use 48, 60 and 80 elements along these three directions, yielding an average el-407 ement size of 1.5 km. Given the minimal Vp of 5.05 km/s from our migration model and 408 about two elements for each P wavelength, it allows us to use periods greater than 0.65409 s for accurate wave simulation. A 7.5 km (about 3 elements) perfect-matched layer (PML) 410 absorbing boundary condition is applied to each surface of the simulation domain to avoid 411 artificial reflections from the boundaries (Komatitsch & Tromp, 2003). As outlined in 412 section 2.3, source time functions are estimated from the aligned vertical component of 413 each P wave event (radial for S wave), followed by an iterative time deconvolution (Kikuchi 414 & Kanamori, 1982) to remove the source time function effect (Bostock, 1998; Rondenay 415 et al., 2000; Bostock et al., 2001; Bostock, 2002; Shang et al., 2017). Given an average 416 station spacing about 2-3 km, the deconvolved event data are bandpass filtered at 1-20 417 s as the adjoint sources for finite frequency wavefield modeling. The migration velocity 418 model is extracted from a 3-D regional model US2016 (Shen & Ritzwoller, 2016). We 419 smooth it using a Gaussian function with a radius of 5.0 km both vertically and hori-420

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421 422 zontally. Then, the migration is implemented for each event and summed up to get the final stacked image. In this study, we perform the migration for P and S waves separately.

423

3.3.4 Results

Due to the finite frequency property, we identify velocity increase as positive mi-424 gration phases flanked by negative ones as indicated by synthetic tests (Figures 4 and 425 6). The basement-sediment contact beneath the Laramie basin is identified in the P-wave 426 image as the positive high amplitude phases beneath the negative phases around 2-7 km 427 (Figure 10a), which becomes slightly deeper (dip to the northwest) between 30-40 km 428 in the horizon, right beneath the Laramie basin. Whereas this feature is not clear from 429 the S wave image shown in Figure 10b, possibly due to the back azimuthal angles of most 430 S events being basically perpendicular to the array direction (Figure 9d). The most promi-431 nent features in both P and S wave images are the bunches of south-dipping positive/negative 432 phases between -40 and 10 km in lateral direction and 0 and 70 km in depths. They are 433 interpreted as the CBSZ, which situated juxtaposition of accreted Proterozoic terranes 434 with the Archean Wyoming craton (Hansen & Dueker, 2009; Tyson et al., 2002). The 435 northward crustal thickening seems to be indicated by the Proterozoic Moho, highlighted 436 by dark green dots in the P wave image (Figure 10a). However, due to spatial aliasing 437 in both P and S wave images between 20 and 40 km horizontally, the Archean Moho is 438 not well constrained (Hansen & Dueker, 2009). Therefore, we try to suppress the spa-439 tial aliasing effect by refining the stations for every 1.0 km as illustrated in Figure 11. 440 The comparison of the interpolated seismograms shows a good match both in trends and 441 amplitudes of different events. We conduct P wave migration again using the refined data. 442 As shown in Figure 11c, most spatial aliasing artifacts are suppressed and both the Archean 443 Moho (~ 45 km) and the Proterozoic Moho (~ 60 km) can be clearly identified. They are 444 slightly dipping northward to the north of the CBSZ. These are consistent with previ-445 ous studies (Allmendinger et al., 1982; Prodehl et al., 1989; Snelson et al., 1998; Moro-446 zova et al., 2002). However, our Proterozoic Moho seems to be distorted with the CBSZ 447 around 15 km in the horizontal direction (south of the CBSZ), whereas Hansen and Dueker 448 (2009) found it to be continuous beneath the entire array. Nevertheless, we interpreted 449 the CBSZ as the orange shadow zones in Figure 11c. Further investigation for a detailed 450 migration velocity model will be helpful for better imaging and interpretation. 451

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452 4 Discussions

We combine full wavefield passive source RTM with the spectral element method so that it is convenient to handle topographic changes and velocity heterogeneities. Both synthetic and suture zone imaging examples demonstrate the performance of our method.

456

4.1 2-D data interpolation

Migration antialiasing due to spatial frequency components (station spacing) is a 457 longstanding problem in seismic imaging, for example, 5-D data interpolation (Trad, 2009) 458 has been developed to improve the imaging quality. Therefore, it is important to con-459 sider the effects of station spacing for passive RTM (L. Chen et al., 2005; Li et al., 2018). 460 The Nyquist-Shannon sampling theorem suggests that the station spacing is required to 461 be smaller than the apparent half-wavelength to fully construct the wavefield from recorded 462 seismograms at the surface (Gray, 2013; Li et al., 2018). Therefore, given constant crustal 463 P and S wave velocities of 5.8 and 3.2 km/s, respectively, and incident angles of 12° to 464 27°, full construction of 1.0 Hz P waves at near-surface requires a station spacing rang-465 ing from 6.4 to 14.5 km, but 3.5 to 8.0 km for 1.0 Hz S waves. This gives us an upper 466 bound limit for the station spacing. Smaller station spacing is expected because the in-467 cident angle for converted Ps waves could be larger, especially when the subsurface struc-468 tures are complex, like subduction and suture zones. For example, given a high-frequency 469 cut about 2.5 Hz, a 2.0 km station spacing would result in slight spatial aliasing as shown 470 in Figure 12a compared to the result obtained with a 1.0 km station spacing shown in 471 Figure 6g. As expected, a 4.0 km station spacing would result in even stronger spatial 472 aliasing as shown in Figure 12b. Depending on the specific imaging target, 2-D data in-473 terpolation might be necessary for migration. However, most data interpolation (regu-474 larization) strategies developed for seismic exploration, like FX domain trace interpo-475 lation (Spitz, 1991), antileakage Fourier transformation (Xu et al., 2005) or curvelet trans-476 formation (Herrmann & Hennenfent, 2008; Shang et al., 2017) require linearity of sta-477 tion distributions, which is not straightforward to handle 2-D irregularly-spaced (3-D if 478 we consider the station elevations) data. Therefore, we prefer to use the 2-D cubic-spline 479 interpolation (J. Zhang & Zheng, 2015) or the radial function-based method (Shepard, 480 n.d.), which can naturally handle irregularly-spaced data for our migration. We note here 481 that, unlike its first application for interpolating 2-D receiver functions (J. Zhang & Zheng, 482 2015), we need to align the multi-channel data prior to interpolation for each component, 483

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and then shift the interpolated data back according to the onsets of incident waves formigration.

486

4.2 Source time function estimations

It is straightforward to estimate the source time functions for P waves, which has been successfully used for teleseismic full waveform inversion (Y. Wang et al., 2016; K. Wang et al., 2021). However, the estimation of source time functions for S waves should be carefully considered, because of relatively larger incident angles compared to P waves. This results in the leakage of S waves into vertical and transverse components. Therefore, rotation of vertical, radial and transverse components into P, SV and SH polarization directions could be helpful for source time function estimations for S waves (Bostock, 1998; Rondenay, 2009).

495

4.3 Migration velocity models

One advantage of our methodology is that we are able to implement migration based 496 on 3-D velocity models with strong heterogeneities. The accuracy of migration velocity 497 models is essential for mapping seismograms to correct locations and avoiding stacking 498 artifacts (Mora, 1989; He & Liu, 2020). To illustrate the advantages of using more ac-499 curate migration velocity models for imaging, we conduct another migration for slab imag-500 ing, which is shown in Figure 13. The migration velocity model is obtained by smooth-501 ing the true velocity model using a Gaussian function with a radius of 5.0 km both hor-502 izontally and vertically. We speculate that, with a more reasonable migration velocity 503 model, the vertical boundaries (see true model in Figure 6a) on the left and right sides 504 of the slab are better constrained compared to those shown in Figure 6g. However, for 505 our suture zone imaging, a smoothed regional 3-D model is not accurate enough because 506 the grid spacing (~ 25 km) of model US2016 (Shen & Ritzwoller, 2016) is too large to 507 capture velocity heterogeneities in such a small study region. Therefore, teleseismic body 508 wave traveltime tomography (D. Zhao et al., 1992; Tong, 2021) or ambient noise tomog-509 raphy (C. Zhang et al., 2018) could be used to construct a more reasonable migration 510 velocity model for our imaging in the future. 511

512 5 Conclusion

In this study, we propose to solve weak-form solutions to decompose elastic wave-513 fields into vector P and S waves for teleseismic reverse time migration based on the spec-514 tral element method. Both synthetic tests and Cheyenne Belt suture zone imaging demon-515 strate the capability of our method to image complex structures with strong velocity het-516 erogeneity. For linear array migration, our synthetic tests show that teleseismic events 517 with back azimuthal angles parallel to the linear array direction contribute more to sub-518 surface migration images than those away from the linear array direction. However, the 519 latter could still contribute to the image beneath the array thanks to larger Fresnel zone 520 contributions at greater depths. In addition, we reveal several south-dipping structures 521 in the Laramie basin, which are interpreted as the Cheyenne Belt suture zone, and are 522 consistent with geological interpretations from previous studies. For better performance 523 of the migration-based imaging method, 2-D/3-D data interpolation is required to avoid 524 spatial aliasing during the construction of wavefields in the subsurface. 525

526 Open Research

The Laramie array seismic data used in this study can be obtained from the IRIS Data Management Center (https://ds.iris.edu/ds) under the network codes XF. We use SPECFEM3-D Cartesian 4.0.0 (Komatitsch et al., 1999; Komatitsch & Tromp, 2002b, 2002a) published under the GPL 3 license for synthetic and real data simulations. PyGMT (Wessel et al., 2019) download from (https://www.pygmt.org/latest/) is used to plot figures.

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539 Author Contributions

The authors confirm their contribution to the paper as follows: study conception and design: Hejun Zhu; Weak-form solutions were first proposed by Yu Chen, then theoretically derivated and tested by Bin He. Data collection: Bin He; Analysis and interpretation of results: Bin He, Hejun Zhu, David Lumley, Qinya Liu, Hitoshi Kawakatsu and Nozomu Takeuch; Draft manuscript preparation: Bin He, Hejun Zhu. All authors review the results and approve the final version of the manuscript.

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Event	Original Time	$\operatorname{Lon}(^{\circ})$	$\operatorname{Lat}(^{\circ})$	Depth(km)
P1	2000_10_04_14_37_44	-62.5590	11.1240	110.3
P2	2000_10_05_13_39_11	-40.9580	31.7320	10
P3	2000_11_08_06_59_58	-77.8290	7.0420	17
P4	2000_11_29_10_25_13	-70.8860	-24.8690	58.2
P5	2000_12_12_05_26_45	-82.6790	6.0150	10
P6	2001_01_10_16_02_42	-153.2810	56.7744	36.4
$\mathbf{P7}$	2001_01_13_17_33_32	-88.6600	13.0490	60
$\mathbf{P8}$	2001_02_13_14_22_05	-88.9380	13.6710	10
P9	2001_03_24_06_27_53	132.5260	34.0830	50
P10	2001_04_09_09_00_57	-73.1090	-32.6680	11
P11	2001_04_14_23_27_26	141.7680	30.0920	10
S1	2000_10_04_16_58_44	166.9100	-15.4210	23
S2	2000_10_27_04_21_51	140.4600	26.2660	388
S3	2000_10_29_08_37_08	153.9450	-4.7660	50
S4	2000_12_21_01_01_27	151.1220	-5.7060	33
S5	2001_01_09_16_49_28	167.1700	-14.9280	103
S6	2001_04_09_09_00_57	-73.1090	-32.6680	11
S7	2001_04_14_23_27_26	141.7680	30.0920	10
S8	2001_04_28_04_49_53	-176.9370	-18.0640	351.8

Table 1: Teleseismic event information for passive source RTM using the Laramie array.



Figure 1: Strong-form solutions for wavefield decomposition based on Equation 6. Panel (a) shows the X component of the back-propagating elastic wavefield. Panels (b) and (d) represent the separated P and S wavefields, respectively. Panel (c) shows the difference between the reference (a) and the summation of strong-form decomposed P (b) and S (d) waves. The Z component has a similar phenomenon, which is not shown here.



Figure 2: Same as Figure 1 but shows the decomposed wavefields based on Equations 7 and 8.



Figure 3: Synthetic teleseismic data with topographic changes in the model shown in Figure 4b. Panels (a) and (b) are Z and X component seismograms, respectively. The plane wave incident depth is 120 km, with a back azimuthal angle of 90°, and an incident angle of 27°. Therefore, X components could be considered as the radial components of teleseismic waves. The black dashed rectangle highlights the effect of topographic changes on recorded data. The magenta short lines denote the time window used to isolate waveforms for RTM.



Figure 4: passive source RTM using 12 teleseismic events with incident-angles ranging from 12° to 27°. The back-azimuthal angle for the first six events is 90° and is 270° for the others. The black line on top of the image represents the station locations for injecting adjoint sources during migration. (a) Without considering the topographic variation, i.e., the stations are assumed to be at 0 km depth. (b) By considering the topographic variations, the stations are modeled at the correct elevations.



Figure 5: Synthetic Z (a) and X (b) components for the first teleseismic event with an incident angle of 27° and a back azimuthal angle of 90°. The magenta short lines denote the time window used to isolate waveforms for RTM. In panel (a), P denotes teleseismic P waves, MP1 and MP2 represent different-order P wave multiples. MSlab together with the dashed blue line, represent slab-related multiples. B is used to denote reflection artifacts due to absorbing boundary conditions. In panel (b), S denotes Ps-converted waves, MS1 and MS2 are corresponding P-multiple converted waves. Slab1 and Slab2 are slab-related Ps-converted waves.


Figure 6: 3-D subducting slab imaging using 12 teleseismic events. The incident angle of the incident wavefield ranges from 12° to 27°. The back-azimuthal angle for events 1, 3, 7, 9 and 11 is 90° and 270° for the others. Panels (a) and (b) show the true velocity profile along the X (west-east) and Y (noth-source) directions for generating synthetic datasets. Panels (c) and (d) show the corresponding migration velocity along the same profiles, which is smoothed from the two-layer background model after removing the slab. Panels (e) and (f) show the stacked images from the first four teleseismic events, while panels (g) and (h) show the final image stacked over all teleseismic events. Panels (e-h) share the same color bars. M1, M2 and M3 in panels (e) and (g) represent multiple artifacts. The artifacts pointed by the green arrows are caused by strong scattering at the sharp edges of the slab.



Figure 7: Similar to panel (g) in Figure 6, but the back azimuthal angles used for these two examples are 0° for events 1, 3, 7, 9 and 11, and 180° for the others (a), which are 45° and 225° in panel (b). The green arrows are used to compare imaged Moho interfaces with Figure 6.



Figure 8: Geological settings of the study region after Hansen and Dueker (2009). The background shows the topography. Several shear zones: LPSZ, Laramie Peak shear zone; FLSZ, Farwell Mt. Lester Mt. suture zone; SFSZ, Soda Creek-Fish Creek shear zone; SGSZ, Skin Gulch shear zone, are denoted by orange lines to illustrate the complexity of the subsurface structures. Other geographic features: SM, Sierra Madres; MB, Medicine Bow Mountains and LM, Laramie Mountains are also labeled. The Cheyenne Belt suture (CB) is denoted by the white line, dashed where it is inferred. WY and CO denotes Wyoming and Colorado, respectively. The black rectangle in the upper-left inset map indicates the location of our study region in North America.



Figure 9: Caption next page.

Figure 9: Telseismic events and data used for migration. Red dots in Panels (a) and (d) represent the distributions of teleseismic P and S events. The light blue dashed line indicates the direction of the Laramie array. Panels (b) and (c) show the vertical (Z) and radial (R) components of the 5th (the blue dot in panel a) teleseismic P events recorded by the Laramie array. The background black solid lines denote the selected high-quality data (some traces are removed due to low signal-to-noise ratio), while the red solid lines represent the 2-D cubic-spline interpolated data at each station. Panels (e) and (f) represent the vertical and radial components of the 6th teleseismic S events. The magenta lines overlaying the seismograms denote either the onsets of P or S waves.



Figure 10: The passive source RTM image beneath the Laramie array. On the top of the figure, we show the elevation of raw (back triangles) and interpolated (red triangles) stations. CB represents the inferred CBSZ location on the surface. Panels (a) and (b) show the images obtained using teleseismic P and S events, respectively. Panel (c) shows the image obtained using interpolated teleseismic P events. The orange belt indicates the interpreted CBSZ. The magenta and dark green dots indicate the interpreted Archean Moho (ArM) and Proterozoic Moho (PtM).



Figure 11: Regularization of the recorded teleseismic data to a fine grid. Panels (a) and (b) compare the raw and interpolation station locations. The interpolated station locations are obtained with GMT projection between the first (L01) and last station (L31) positions of the Laramie array along the great circle for every 1.0 km, which is not on a straight line after the UTM projection. Panels (c) and (d) show comparisons of the raw data (black) and interpolated seismograms (red) for vertical (Z) and radial (R) components, respectively.



Figure 12: Imaging with sparse station spacing. The imaging parameters are the same as panel (g) in Figure 6 except that the station spacing is 2 km and 4 km for panels (a) and (b), respectively. The aliasing artifacts are illustrated by black arrows.



Figure 13: Imaging with a more accurate background migration velocity model. Panels (a) and (b) show the migration velocity profiles along the X (west-east) and Y (noth-source) directions. Panels (c) and (d) show the corresponding RTM images. The green arrows are used to highlight the improvement in imaging of the slab boundaries compared to panel (g) in Figure 6.

Passive source reverse time migration based on the spectral element method

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¹⁰ Key Points:

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2

11	• A 3-D passive source reverse time migration based on the spectral element method
12	is proposed to image complex structures.
13	• Amplitude-preserved vector P and S wavefields are accurately decomposed by solv-

- ¹⁴ ing corresponding weak-form solutions.
- Several south-dipping structures can be identified from P and S migration results
 beneath the Laramie basin, which are interpreted as the Cheyenne Belt suture zone.

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17 Abstract

Increasing deployment of dense arrays has facilitated detailed structure imaging 18 for tectonic investigation, hazard assessment and resource exploration. Strong velocity 19 heterogeneity and topographic changes have to be considered during passive source imag-20 ing. However, it is quite challenging for rav-based methods, such as Kirchhoff migration 21 or the widely used teleseismic receiver function, to handle these problems. In this study, 22 we propose a 3-D passive source reverse time migration strategy based on the spectral 23 element method. It is realized by decomposing the time reversal full elastic wavefield into 24 amplitude-preserved vector P and S wavefields by solving the corresponding weak-form 25 solutions, followed by a dot-product imaging condition to get images for the subsurface 26 structures. It enables us to use regional 3-D migration velocity models and take topo-27 graphic variations into account, helping us to locate reflectors at more accurate positions 28 than traditional 1-D model-based methods, like teleseismic receiver functions. Two syn-29 thetic tests are used to demonstrate the advantages of the proposed method to handle 30 topographic variations and complex velocity heterogeneities. Furthermore, applications 31 to the Laramie array data using both teleseismic P and S waves enable us to identify sev-32 eral south-dipping structures beneath the Laramie basin in southeast Wyoming, which 33 are interpreted as the Cheyenne Belt suture zone and agree with, and improve upon pre-34 vious geological interpretations. 35

36

Plain Language Summary

Increasing deployment of dense arrays has allowed detailed structure imaging for 37 tectonic investigation, hazard assessment and resource exploration. However, traditional 38 ray-based migration methods or 1-D velocity model-based receiver function methods may 39 greatly degrade the imaging quality without considering the full wavefield propagation 40 effect and velocity heterogeneities. Therefore, in this study, we develop a 3-D passive source 41 migration based on the spectral element method, which is capable of handling topographic 42 variations as well as complex velocity heterogeneities for real data applications. Several 43 synthetic tests and applications to the Laramie array data using both teleseismic P and 44 S waves are used to demonstrate the capability of our method to image complex struc-45 tures, such as subduction and suture zones. 46

-2-

47 **1** Introduction

In past decades, teleseismic receiver functions have been widely used for imaging 48 discontinuities within the Earth's crust and upper mantle (Burdick & Langston, 1977; 49 Langston & Corvallis, 1977; Vinnik, 1977; Ai et al., 2007; Kawakatsu & Watada, 2007; 50 Kind et al., 2012; H. Zhang et al., 2016; Long et al., 2019). Based on the assumption of 51 a 1-D Earth model where discontinuities are horizontal, common convert point stack-52 ing (Dueker & Sheehan, 1997; L. Zhu, 2000; Gilbert et al., 2003) has generally been used 53 to improve the image quality by mapping the receiver functions to depths. In the case 54 of a single station, like the SEIS on Mars (Banerdt et al., 2020; Knapmeyer-Endrun et 55 al., 2020; Lognonné et al., 2020) or sparse station distributions such as the USArray (Meltzer 56 et al., 1999), such simplifications are natural and useful. However, it indeed degrades the 57 image quality of geologically complex structures, such as subsurface environments with 58 steep faults and laterally discontinuous interfaces, without considering the effects of wave-59 field diffraction and scattering caused by lateral variations of impedance contracts and 60 velocities (L. Chen et al., 2005; Shang et al., 2012). In addition, increasing deployment 61 of dense and nodal arrays has facilitated detailed structure imaging for tectonic inves-62 tigation, hazard assessment and resource exploration (Okada et al., 2004; Zheng et al., 63 2008; Nábělek et al., 2009; H. Zhang et al., 2016; X. Yang et al., 2017; X. Chen et al., 64 2018; Li et al., 2018; Clayton et al., 2019; X. Wang et al., 2021; Onyango et al., 2022; 65 Wu et al., 2023). Therefore, it calls for advanced imaging methods, such as seismic mi-66 gration/inversion, which exploit the full complexity of recorded wavefields and rely less 67 on a priori information about the Earth's structures, to get accurate images of complex 68 subsurface discontinuities (Sheehan et al., 2000; L. Chen et al., 2005; Shang et al., 2012; 69 Cheng et al., 2016; Li et al., 2018; Millet et al., 2019). 70

Among many seismic migration approaches, Kirchhoff migration/inversion (Schneider, 71 1978; Gray & May, 1994) has been widely used in the oil and gas industry to image shal-72 low sedimentary structures owing to its high efficiency and simplicity. It was later in-73 troduced to global seismology for imaging large-scale structures using multi-component 74 teleseismic data recorded by local dense arrays (Burridge et al., 1998; Bostock, 1998; Bo-75 stock & Rondenay, 1999; Rondenay et al., 2000; Bostock et al., 2001; Bostock, 2002; Levan-76 der et al., 2005; Rondenay, 2009; Liu & Levander, 2013; Cheng et al., 2016; Millet et al., 77 2019). Teleseismic Kirchhoff migration/inversion is implemented by weighting and stack-78 ing data along diffraction hyperbola for every possible scattering point in a regular grid 79

(Schneider, 1978; Gray & May, 1994; Bostock et al., 2001; Cheng et al., 2016; Millet et 80 al., 2019). To date, most Kirchhoff migration/inversion approaches are based on the in-81 finite high-frequency assumption by neglecting the finite frequency property of seismic 82 wave propagation, which enables us to efficiently get images by using ray tracing (Richards 83 & Aki, 1980) or solving the Eikonal equation (Gray & May, 1994; Cheng et al., 2016). 84 However, It is difficult for ray-tracing to handle multiple arrivals, shadow zones, and even 85 chaotic rays in complex subsurface environments (Audebert et al., 1997; L. Chen et al., 86 2005). Although the Eikonal equation-based Kirchhoff migration is capable of dealing 87 with complex structures, it is not easy to handle the multipathing challenge (H. Zhao, 88 2005; Waheed et al., 2015; Tong, 2021). Furthermore, Kirchhoff migration requires more 89 efforts to correct image amplitudes, and therefore, the wave-equation Kirchhoff migra-90 tion has been developed to balance accuracy and efficiency (Andrade et al., n.d.; Jin & 91 Etgen, n.d.). To mitigate the finite-frequency, multi-arrivals and inaccurate amplitude 92 problems, Gaussian beam migration (Hill, 1990, 2001) has been proposed to apply lo-93 cal slant stacks by using complex-valued traveltimes and amplitudes. These complex quan-94 tities come from the approximation of seismic wave propagation with a sum of Gaussian 95 beams, which are finite-frequency, ray-theoretical approximations to the wave equation 96 (Notfors et al., n.d.). It is further extended to the wave-equation-based two-way beam 97 wave method, which has comparable accuracy and efficiency in comparison to the wave-98 equation-based methods (J. Yang et al., 2022). 99

On the other hand, wave equation-based migration methods can essentially avoid 100 most difficulties from the ray-based method by using the full wavefields (Claerbout, 1985). 101 Among many different types of wave equation-based migration methods, synthetic and 102 field experiments have shown that reverse-time migration (RTM) (Baysal et al., 1983; 103 Whitmore, 1983; McMechan, 1983) has better performance for imaging complex struc-104 tures. Active source RTM includes a forward wavefield modeling from the source and 105 a backward wavefield modeling from the receivers, followed by applying imaging condi-106 tions to these two wavefields. In contrast, passive source RTM only requires one back-107 ward wavefield modeling for each earthquake. Therefore, inaccuracies due to source lo-108 cations and velocity variations between earthquake locations and local study regions could 109 be neglected. During the wavefield back-propagation, P and S wavefields are usually sep-110 arated by polarization decomposition (Shang et al., 2012; H. Zhu, 2017; J. Yang et al., 111 2018). Then, the imaging condition is applied to cross-correlate separated P and S wave-112

-4-

fields, followed by the summation of individual events. As pointed out by Shang et al. 113 (2012), passive source RTM of converted waves differs fundamentally from single station 114 receiver function analysis, and also in several important ways from more traditional re-115 ceiver function migration (L. Chen et al., 2005). However, both current 2-D and 3-D pi-116 oneer works are based on the finite difference method, which requires projecting station 117 locations on the model grids, and it is not easy to handle complex topography in real 118 applications (Rajasekaran & McMechan, 1995; Bevc, 1997; Shragge, 2014; Shang et al., 119 2017; Yi et al., 2019). In addition, there are few real data applications for 3-D passive 120 source RTM to image crustal and uppermost mantle structures. 121

In this study, we combine passive source RTM with the spectral-element method 122 (SEM) (Komatitsch & Vilotte, 1998; Fichtner et al., 2009). Owning to its advantages 123 in handling complex topographic variations, wavefield coupling between different media, 124 flexibility to handle station distributions and so on, the SEM has been widely applied 125 to construct crustal and mantle velocity models at regional and global scales (Tromp et 126 al., 2005; Fichtner et al., 2009; Tape et al., 2009; Fichtner, 2010; Liu & Levander, 2013; 127 H. Zhu et al., 2012; Y. Wang et al., 2016; K. Wang et al., 2020, 2021; Tromp, 2020; H. Zhu 128 et al., 2020; Maguire et al., 2022). Here, we use it for passive source RTM. Because the 129 wave-equation is solved in a weak-form fashion in SEM, directly taking strong-form spa-130 tial derivatives for wavefield decomposition results in significant discontinuities between 131 element boundaries. To solve this problem, we propose to solve the corresponding weak-132 form solutions for P and S wavefield decomposition. This paper is organized into the fol-133 lowing parts, we first briefly review passive source RTM based on the wavefield decom-134 position, followed by the theory of weak-form solutions for wavefield decomposition, and 135 then several synthetic tests are used to demonstrate the advantages of the proposed method. 136 Finally, we apply it to the Laramie array for suture zone imaging. 137

- 138 2 Methodology
- 130

2.1 Passive source RTM based on vector wavefield decomposition

Based on the principle that all back-propagating direct P and S receiver wavefields $(\mathbf{u}^P \text{ and } \mathbf{u}^S)$ should coincide at "source" locations, wave-equation-based passive source imaging seeks strong energy in space as the source locations (McMechan, 1982; Nakata & Beroza, 2016; J. Yang et al., 2020; Duan et al., 2021). Similarly, the vector \mathbf{u}^P and u^S wavefields constructed by back-propagating multichannel direct P and Ps converted
 (or direct S and Sp converted) seismograms, should coincide at the scattering locations
 in depths, such as the Moho (Shang et al., 2012; Li et al., 2018). The imaging condition
 can be given as

$$I(\mathbf{x}) = \sum_{i_e=1}^{Ne} \int_{t=0}^{T} \boldsymbol{u}(\mathbf{x})_{i_e}^{P}(t) \cdot \boldsymbol{u}(\mathbf{x})_{i_e}^{S}(t) dt \quad ,$$
(1)

where $I(\mathbf{x})$ is the image, $\boldsymbol{u}(\mathbf{x})_{i_e}^P$ and $\boldsymbol{u}(\mathbf{x})_{i_e}^S$ are the constructed vector P and S wavefields for i_e th event at the spatial coordinate \mathbf{x} . N_e is the total number of events for imaging. t represents the current time and T denotes the total time for wavefield back-propagation. " \cdot " denotes the dot product between two vector fields. The final image is obtained by summing over all the time steps during wavefield back-propagation and all seismic events.

Because the vector \mathbf{u}^P and \mathbf{u}^S wavefields are coupled during elastic wavefield simulations, one may consider back-propagate multichannel P and Ps (or S and Sp) seismograms independently by solving two acoustic wave-equations (Sun et al., 2004; Duan et al., 2021) or decouple them during the wavefield back-propagation. H. Zhu (2017) proposed an amplitude-preserved wavefield decomposition method to decouple the vector \mathbf{u}^P and \mathbf{u}^S from elastic wavefield \mathbf{u} by introducing a vector \mathbf{w} in the form of

$$\nabla^2 \mathbf{w} = \mathbf{u} = \mathbf{u}^P + \mathbf{u}^S \quad , \tag{2}$$

¹⁵⁹ where vector P and S wavefields can be obtained according to

$$\mathbf{u}^P = \nabla (\nabla \cdot \mathbf{w}), \quad \mathbf{u}^S = -\nabla \times \nabla \times \mathbf{w} \quad . \tag{3}$$

Here ∇ , ∇ and ∇ are the gradient, divergence and curl operators, respectively. How-

ever, solving Equation 3 requires explicitly solving the Poisson's equation in Equation

¹⁶² 2. To reduce the computational costs, J. Yang et al. (2018) further extended this method

to avoid solving Poisson's equation under the isotropic media assumption with

$$\mathbf{u}^P = \nabla (\nabla \cdot \alpha^2 \hat{\mathbf{u}}), \quad \mathbf{u}^S = -\nabla \times (\nabla \times \beta^2 \hat{\mathbf{u}}) \quad , \tag{4}$$

where $\hat{\mathbf{u}}$ denotes the extrapolated wavefield excited by a double integral of the original seismograms, α and β are the local P and S velocities, respectively.

166

2.2 Strong- and weak-form solutions for the wavefield decomposition

¹⁶⁷ In this section, we introduce H. Zhu (2017) and J. Yang et al. (2018)'s method to ¹⁶⁸ the SEM, so that it is more convenient to handle irregular surface topography and sta¹⁶⁹tion locations. Different from the finite difference (FD) method, SEM seeks a weak so-

- 170 lution $\bar{\boldsymbol{u}}(\mathbf{x},t)$ to the equation of motion in the Galerkin sense (Komatitsch & Vilotte,
- ¹⁷¹ 1998; Fichtner, 2010). Lagrange polynomials collocated at the Gauss–Lobatto–Legendre(GLL)
- points are used to interpolate the wavefield at any spatial locations in the form of

$$\boldsymbol{u}(\mathbf{x},t) \approx \bar{\boldsymbol{u}}(\mathbf{x},t) = \sum_{i,j,k=1}^{N+1} \boldsymbol{u}^{ijk}(t) L_{ijk}(\mathbf{x}) \quad , \tag{5}$$

where $\boldsymbol{u}(\mathbf{x},t)$ and $\bar{\boldsymbol{u}}(\mathbf{x},t)$ represent the true and weak-form solutions of the wave-equation, respectively. \mathbf{x} represents the spatial coordinate and t represents time. N denotes the degree of the Lagrange polynomials, and therefore we have N + 1 GLL points for one element along each direction. $L_{ijk} = l_i(x)l_j(y)l_k(z)$ represents the basis function chosen as the product of three Lagrange polynomials l(x), l(y) and l(z) along the x, y and z directions collocated at the corresponding GLL points.

Owing to the continuous prosperities of the Lagrange polynomials, a straightforward solution of Equation 4 based on SEM, noted as the strong-form solution, can be realized by

$$\mathbf{u}^{P} = \nabla [\nabla \cdot \alpha^{2} \sum_{i,j,k=1}^{N+1} \boldsymbol{u}^{ijk}(t) L_{ijk}(\mathbf{x})], \quad \mathbf{u}^{S} = -\nabla \times [(\nabla \times \beta^{2} \sum_{i,j,k=1}^{N+1} \boldsymbol{u}^{ijk}(t) L_{ijk}(\mathbf{x})] \quad , \quad (6)$$

which is fairly easy to implement because the spatial derivatives from the gradient, di-182 vergence and curl operators are taken from the Lagrange polynomials. However, we ex-183 pect quite strong artifacts in the separated vector wavefields, which could contaminate 184 the final imaging result. As illustrated in Figure 1, strong S mode artifacts appear in the 185 P mode. This is because the solution of SEM is a weak-form approximation, and although 186 the wavefield is continuous cross elements, the continuity of spatial derivatives is not guar-187 anteed. To mitigate this leakage during wavefield decomposition, we use the weak so-188 lutions for Equation 4 as follows: 189

$$\int_{G_e} \mathbf{\Phi} \cdot \mathbf{u}^P d^3 \mathbf{x} = \int_{G_e} \mathbf{\Phi} \cdot \nabla (\nabla \cdot \alpha^2 \mathbf{u}) d^3 \mathbf{x} = -\int_{G_e} [\nabla \cdot \mathbf{\Phi}] [\nabla \cdot (\alpha^2 \mathbf{u})] \quad , \tag{7}$$

$$\int_{G_e} \mathbf{\Phi} \cdot \mathbf{u}^S d^3 \mathbf{x} = -\int_{G_e} \mathbf{\Phi} \cdot \nabla \times (\nabla \times \beta^2 \mathbf{u}) d^3 \mathbf{x} = \int_{G_e} [\nabla \times \mathbf{\Phi}] \cdot [\nabla \times \beta^2 \mathbf{u}] d^3 \mathbf{x} \quad , \quad (8)$$

where Φ denotes any arbitrary, differentiable, time-independent test function, which will be chosen as the basis functions similar to the way we get SEM solutions. As shown in Figure 2, both P and S modes are clearly separated without significant leakages. This indicates that the weak-form solutions can significantly improve the quality of wavefield decomposition, which is the basis for the following imaging tests. The deviations of Equations 7 and 8 can be found in the supplementary material. 197

2.3 The workflow for passive source RTM

The implementation of passive source RTM can be summarised in the following four steps:

200	1.	Teleseimic data preprocessing. During this step, we estimate source time functions
201		for each teleseismic event based on the principle component analysis for the aligned
202		incident waves (Bostock & Rondenay, 1999; Shang et al., 2017), such as P, S or
203		SKS waves. For this purpose, the incident waves are aligned according to the on-
204		set time of the incident phases estimated using the TauP package (Crotwell et al.,
205		1999). Then, an iterative time deconvolution (Kikuchi & Kanamori, 1982) is ap-
206		plied to each component of original (unaligned) data to remove the source time func
207		tion, followed by bandpass filtering for later finite frequency simulations.
208	2.	Injecting the preprocessed teleseismic data for the SEM solver as adjoint sources
209		to the acceleration vector. Given the linear relation between the source time func-
210		tion and wavefield, this step further avoids obtaining the double integral of the
211		teleseismic data, which should be injected to construct the second time integral
212		of the displacement wavefield indicated by Equation 4.
213	3.	During the reverse time wavefield propagation, solving the weak-form solutions
214		shown in Equations 7 and 8 to get vector wavefields $\mathbf{u}_{i_e}^P$ and $\mathbf{u}_{i_e}^S$, followed by ap-
215		plying the zero-lag cross-correlation imaging condition (Equation 1), noted as $\mathbf{I}_{ie}^{PS}.$
216		To save computational costs, one may consider performing wavefield separation
217		for every several time steps according to the Nyquist rule. We also apply the imag-
218		ing condition to $\mathbf{u}_{i_e}^P$ with $\mathbf{u}_{i_e}^P$ to get the PP image as $\mathbf{I}_{i_e}^{PP}$.

- 4. Sum all corresponding event images to obtain \mathbf{I}_{PS} and \mathbf{I}_{PP} , and then get the image with $\mathbf{I}_{PS} = \frac{\mathbf{I}_{PS}}{\mathbf{I}_{PP}+\epsilon}$ to compensate for energies at great depths. Here, ϵ is a small damping value used to avoid dividing by zeros.
- 222 **3** Applications

In this section, we use two synthetic tests and one real data example to validate our method based on linear array recordings although we are using 3-D modeling and migration. It is well-known that station spacing together with the data frequency are important factors that may result in spatial aliasing during passive source RTM (Gray, 2013; Shang et al., 2012). Therefore, most teleseismic migration methods used to image

lithospheric structures rely on dense arrays (station spacing is ~ 5 km) (Bostock, 1998; 228 Rondenay et al., 2000; Rondenay, 2009; Shang et al., 2017). Therefore, all synthetic sim-229 ulations in our tests are implemented using a 3-D plane-wave injection method, FK-SEM 230 (Tong, Komatitsch, et al., 2014; Tong, Chen, et al., 2014) and recorded by a dense lin-231 ear array. Then, passive source RTM is implemented based on 3-D migration. We note 232 here, there are mainly three advantages to implement 3-D rather than 2-D migration even 233 using recordings from a linear array: 1) no 3-D/2-D transformation is needed ("Line-source 234 simulation for shallow-seismic data. Part 1: Theoretical background, author=Forbriger, 235 Thomas and Groos, Lisa and Schäfer, Martin", 2014; Schäfer et al., 2014; C. Zhang et 236 al., 2018), which is important to correct phase and amplitude differences between 3-D 237 and 2-D simulations; 2) more teleseismic events can be chosen for real data applications; 238 3) no needs to project stations onto a line, which is required for 2-D migration even with 239 linear arrays because their locations are not critically located on 2-D linear mesh grid 240 (Shang et al., 2017). 241

242

3.1 Synthetic test 1: effects of topographic variations

The first synthetic test is used to highlight the effects of topographic variations on 243 passive source migration when using high-frequency teleseismic data. The input model 244 is a two-layer model with an interface at 30 km depth. The density, Vp and Vs for the 245 first layer between 0 and 30 km depth are 2.72 g/cm^3 , 5.8 km/s and 3.46 km/s, which 246 are 3.423 g/cm^3 , 8.06 km/s and 4.53 km/s between 30 and 60 km depths. These values 247 are taken from the AK135 model (Kennett et al., 1995). The simulation domain spans 248 -90 to 90 km along the X (longitude) direction, -60 to 60 km along the Y (latitude) di-249 rection and 0 to 60 km along the depth direction. The element size is 1.0 km, which al-250 lows for the wavefield simulation with the highest frequency around 3.0 Hz. The Stacey 251 absorption boundary condition is used for each face of the model except for the top, which 252 is a free-surface boundary condition for our simulations (Mahrer, 1990). A 1.5-D topo-253 graphic change (no change along the Y direction) following the solid line shown in Fig-254 ure 4b is designed to validate its effects on passive source RTM. We note here, the model 255 will be stretched in depth direction within the SPECFEM package to handle the topo-256 graphic changes. 161 stations with a horizontal station spacing of 1.0 km are distributed 257 on the surface of the model. We use a Ricker source wavelet with a peak frequency of 258 1.0 Hz to initialize the incident plane waves. 12 P-wave plane waves, with incident an-259

gles ranging from $12^{\circ}-27^{\circ}$, are injected into the local simulation domain to simulate the 260 teleseismic waves. The back-azimuthal angles for events 1, 3, 7, 9 and 11 are 90° and 270° 261 for the rest events. This makes a symmetric coverage of plane waves propagating through 262 the model. The corresponding teleseismic data are recorded by the station for each event. 263 Seismograms for the first teleseismic event with an incident angle of 27° and the back-264 azimuthal angle of 90° are shown in Figure 3. By using the short-time average over long-265 time average (STA/LTA) algorithm (Withers et al., 1998), it is feasible to detect the on-266 set of the incident wavefield. The magenta lines in Figure 3 denote the time windows used 267 in our imaging, which are 5 seconds before and 6 seconds after the onset of the incident 268 P waves. It helps us to isolate multiple reflections that may distort our imaging results. 269 We speculate that the onset of incident P waves in both Z and X components shows the 270 imprint of the topographic changes. Without considering the topographic changes, i.e., 271 injecting teleseismic data at 0 km elevation of each station, a significant imprint of to-272 pography can be observed in the imaging result in Figure 4a, which looks like a mirror 273 of the topography. On the contrary, the reflector can be accurately imaged by injecting 274 the teleseismic data at the right locations by considering the topographic changes (Fig-275 ure 4b). 276

277

3.2 Synthetic test 2: imaging subduction slabs

In this synthetic test, we validate the capability of our method to image complex 278 structures in subduction zones. The background model is a two-layer crust and mantle 279 structure, with density, Vp and Vs, the same as the previous example. A 3-D subduct-280 ing slab (Figure 6a-b) is designed between -60 to 60 km along the X (Longitude) direc-281 tion, -30 to 30 km along the Y (Latitude) direction and 40 to 140 km in depth with a 282 thickness of 50 km. The Moho depth is 35 km. The model parameters (density, Vp and 283 Vs) of the subducting slab are 12% greater than the background model. Our simulation 284 domain spans from -120 to 120 km along the X direction, -48 to 48 km along the Y di-285 rection and 0 to 160 km along the depth direction. We use 201 stations with 1.0 km sta-286 tion spacing evenly distributed from -100.0 km to 100.0 km. 12 teleseismic P wave events 287 with the same back-azimuth and incident angles as the previous example are injected 288 into the local study region, starting from an initial depth at 240 km. The rest param-289 eters are the same as the previous synthetic test. Synthetic Z and X component seismo-290 grams for the first teleseismic event are illustrated in Figures 5a and b. Because the back-291

azimuth angle of this event is 90°, the X components are equal to the radial components,
which could be used to investigate Ps-converted waves. Due to the complexity of the velocity model, we speculate several kinds of multiple reflections (Moho and slab-related)
for the P waves in Z components. The corresponding Ps-converted waves can be investigated from X components in Figure 5b. Interestingly, despite being weaker than Mohorelated converted waves (S), slab-related as well as multiple reflection-related S-converted
waves can also be seen.

Our RTM is implemented using a smoothed two-layer background model (Figures 299 6 c-d) to avoid artifacts arising from sharp velocity interfaces during the wavefield sep-300 aration (J. Yang et al., 2018). It is smoothed vertically using a Gaussian function with 301 a radius of 5.0 km. This means we do not include the slab in our migration velocity model, 302 which is close to the real cases. After stacking and illuminating, the interfaces of the slab 303 are well imaged as displayed in Figure 6g. It suggests that our method is capable of imag-304 ing complex structures, such as velocity anomalies with high dipping angles. Because it 305 is quite difficult to isolate multiple reflections in this case, we also see different-order mul-306 tiple imaging artifacts in our stacked result. We also show the imaging result from the 307 first four teleseismic events, which also successfully captures the main features of the slab 308 and the Moho, but with stronger artifacts due to not enough stacking. In addition, the 309 images along the Y (Latitude) direction (Figures 6f and h), which is perpendicular to 310 our stations' distribution, indicate that with only several effective stations, the migra-311 tion will map the converted waveforms along the isochrone interfaces in depths (Schneider, 312 1978). Because we only use one linear array across the slab, it is necessary to consider 313 the contributions of teleseismic events with back-azimuthal angles away from the array 314 direction. We test another three back-azimuthal angle pairs, which are $0^{\circ}/180^{\circ}$, $45^{\circ}/225^{\circ}$ 315 and $135^{\circ}/315^{\circ}$ with the same incident angle as the previous test. Figure 7 shows the mi-316 gration results from the back-azimuthal angle pairs of $0^{\circ}/180^{\circ}$ and $45^{\circ}/225^{\circ}$. The mi-317 gration result from $135^{\circ}/315^{\circ}$ azimuthal-angle pair is similar to the one from $45^{\circ}/225^{\circ}$. 318 As expected, when the azimuthal angles are away from the linear array direction, the 319 interfaces of the imaged slab and the Moho become weaker and more incoherent arti-320 facts become stronger. Interestingly, the Moho interface disappears when using teleseis-321 mic events with the $0^{\circ}/180^{\circ}$ azimuthal pair. This could be explained by the concept of 322 the Fresnel zone. Bostock (1998) pointed out that the Fresnel zone of the scattering points 323 depends on the dominant frequencies and depths of the interface. Assuming that con-324

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structive interference arises for phase differences less than a quarter period, the diam-325 eter of the Fresnel zone around the interface varies between ~ 25 km for $P_m S$ (Moho con-326 verted wave) at 0.5 Hz to ~160 km for $P_{660}S$ (generated at the 660 km discontinuity) 327 at 0.25 Hz. This means the converted waves recorded at stations that arise from deeper 328 interfaces (e.g., X=0, Y=0 and Z=660 km) could be the response to a larger range of 329 interfaces, e.g., X=[-80, 80], Y=[-80, 80], Z=[580, 740] km. We note here, the estima-330 tions of X and Y ranges are fairly rough. For teleseismic events with back-azimuthal an-331 gles away from the linear array, their Fresnel zone is also away from the structures be-332 neath the linear array. As a consequence of small Fresnel zones, shallow interfaces, such 333 as the Moho, gradually disappear beneath the linear array due to incoherence stacking 334 from different events. Therefore, reasonable good station, incident angle and azimuthal 335 angle coverages are preferred for our method if we intend to image the interfaces beneath 336 a linear array (Shang et al., 2012; Li et al., 2018). 337

338 339

3.3 Imaging the Cheyenne Belt suture zone (CBSZ) using the Laramie array

340

3.3.1 Geological setting

The CBSZ is a tectonic suture zone between the Archean Wyoming craton to the 341 north and the Paleoproterozoic Yavapai province to the south (Karlstrom & Houston, 342 1984; Sims & Stein, 2003; Hansen & Dueker, 2009; Jones et al., 2010). It contains a set 343 of steeply south-dipping shear zones formed during the 1.78–1.75 Ga Medicine Bow orogeny 344 when the Proterozoic Green Mountain arc collided with the passive margin of the Wyoming 345 craton via south-facing subduction (Tyson et al., 2002; Hansen & Dueker, 2009). Steep 346 stretching lineations and shear-sense features indicate south-side-up motion (Tyson et 347 al., 2002; Hansen & Dueker, 2009). To better constrain the structure of the CBSZ, the 348 Laramie array was deployed within the Laramie basin across the inferred trace of the 349 Cheyenne belt (Figure 8) (Hansen & Dueker, 2009). It is a dense 80 km long linear ar-350 ray with broadband seismometers, which consists of 30 sensors spaced 2.2 km apart and 351 was deployed for a period of eight months in 2000–2001. Based on P and S receiver func-352 tions together with teleseismic P wave traveltime tomography, Hansen and Dueker (2009) 353 found an imbricated Moho north of the Cheyenne belt. It is basically consistent with the 354 interpretation of seismic results from the CDROM (Continental Dynamics Rocky Moun-355 tain) project (Tyson et al., 2002). However, either due to the limited aperture of the Laramie 356

array, or the methodologies, the CBSZ hasn't been clearly imaged as shown in Figure 357

7 of Hansen and Dueker (2009). Ruigrok et al. (2010) used seismic interferometry to ex-358

tract reflection responses from the coda of the transmitted energy from distant earth-359

- quakes, where they found discontinuities in their migration images, which were interpreted 360
- as the CBSZ. Here, we use our passive source RTM method to further investigate the 361
- detailed shape of the CBSZ with converted teleseismic P and S waves. 362
- 3.3.2 Data 363

364

The teleseismic P wave dataset is constructed from 11 events at $30^{\circ}-90^{\circ}$ distance with body-wave magnitudes greater than 5.5. The S wave dataset is selected from 1 S-365 wave and 7 SKS events from $55^{\circ}-85^{\circ}$ and $85^{\circ}-120^{\circ}$ epicentral distances, respectively (Wilson 366 & Aster, 2005; Yuan et al., 2006). Detailed information about these events can be found 367 in Table 1 and displayed in Figure 9. The data is selected, downloaded and preprocessed 368 with the standing order for data package (SOD) (Owens et al., 2004). For each P wave 369 event, three-component waveforms within the time window of two minutes before and 370 three minutes after direct P arrivals predicted by the AK135 model (Kennett et al., 1995) 371 are collected. The north-south and west-east component seismograms are rotated to ra-372 dial and transverse components after removing instrument response, linear trend, and 373 mean values, followed by a bandpass filter of 1-20 s. Then, the preprocessed three-component 374 waveforms of each event are visually inspected, and only those with a signal-to-noise ra-375 tio (SNR) larger than 3.0 and 2.0 for vertical and radial components are kept. Afterward, 376 we use the open-source software AIMBAT (Lou et al., 2013) to remove bad seismograms 377 with spurious amplitudes and cross-correlation coefficients lower than 0.80 for vertical 378 and radial components. To avoid spatial aliasing, only events with more than 15 seis-379 mograms are remained. The data selection process for S and SKS events is similar ex-380 cept: (1) the time window is defined as two minutes before and three minutes after in-381 cident S arrivals; (2) seismograms with SNR larger than 2.0 and 3.0 for vertical and ra-382 dial components are kept. This is because S waves are mainly in radial components. In 383

the end, most events have more than 25 high-quality seismograms for each component. 384

Gray (2013) and Li et al. (2018) suggested that station spacing is an important fac-385 tor for spatial aliasing during passive source RTM. For example, 5 km station spacing 386 will result in slightly aliased P-wave migration with 1 Hz data, given the incident an-387 gles range from 12 ° to 27°. Therefore, we interpolate those deleted bad seismograms 388

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using 2-D cubic-spline interpolation (J. Zhang & Zheng, 2015). The 2-D cubic-spline in-389 terpolation was originally used to refine the receiver functions from sparse station dis-390 tributions. Here, we use it to interpolate the aligned vertical and radial component seis-391 mograms. It mainly includes three steps: (1) align the seismograms according to the on-392 set of incident waves (P or S), predicted by the reference model, such as AK135 (Kennett 393 et al., 1995). The reference time could be further adjusted by applying a multichannel 394 cross-correlation algorithm (VanDecar & Crosson, 1990); (2) perform 2-D interpolation 395 for each time step using the cubic-spline method for each component; (3) shift the in-396 terpolated data back according to the reference time. We note here, the absolute am-397 plitudes are kept during interpolation for each channel and component, which is differ-398 ent from J. Zhang and Zheng (2015). This is important for dealing with multi-component 399 data. As compared in Figure 9, the interpolated seismograms follow the same trend of 400 nearby stations in great details for both P and S waves. We use 10 s before and 30 s af-401 ter the onset of P and 30 s before and 30 s after the onset of S waves for the following 402 migration. 403

404

3.3.3 Migration

Our computational domain for wavefield propagation spans from -106.25° to -105.30° 405 along longitude, 40.85° to 41.75° along latitude, and -10 to 110 km in the depth direc-406 tion. We use 48, 60 and 80 elements along these three directions, yielding an average el-407 ement size of 1.5 km. Given the minimal Vp of 5.05 km/s from our migration model and 408 about two elements for each P wavelength, it allows us to use periods greater than 0.65409 s for accurate wave simulation. A 7.5 km (about 3 elements) perfect-matched layer (PML) 410 absorbing boundary condition is applied to each surface of the simulation domain to avoid 411 artificial reflections from the boundaries (Komatitsch & Tromp, 2003). As outlined in 412 section 2.3, source time functions are estimated from the aligned vertical component of 413 each P wave event (radial for S wave), followed by an iterative time deconvolution (Kikuchi 414 & Kanamori, 1982) to remove the source time function effect (Bostock, 1998; Rondenay 415 et al., 2000; Bostock et al., 2001; Bostock, 2002; Shang et al., 2017). Given an average 416 station spacing about 2-3 km, the deconvolved event data are bandpass filtered at 1-20 417 s as the adjoint sources for finite frequency wavefield modeling. The migration velocity 418 model is extracted from a 3-D regional model US2016 (Shen & Ritzwoller, 2016). We 419 smooth it using a Gaussian function with a radius of 5.0 km both vertically and hori-420

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421 422 zontally. Then, the migration is implemented for each event and summed up to get the final stacked image. In this study, we perform the migration for P and S waves separately.

423

3.3.4 Results

Due to the finite frequency property, we identify velocity increase as positive mi-424 gration phases flanked by negative ones as indicated by synthetic tests (Figures 4 and 425 6). The basement-sediment contact beneath the Laramie basin is identified in the P-wave 426 image as the positive high amplitude phases beneath the negative phases around 2-7 km 427 (Figure 10a), which becomes slightly deeper (dip to the northwest) between 30-40 km 428 in the horizon, right beneath the Laramie basin. Whereas this feature is not clear from 429 the S wave image shown in Figure 10b, possibly due to the back azimuthal angles of most 430 S events being basically perpendicular to the array direction (Figure 9d). The most promi-431 nent features in both P and S wave images are the bunches of south-dipping positive/negative 432 phases between -40 and 10 km in lateral direction and 0 and 70 km in depths. They are 433 interpreted as the CBSZ, which situated juxtaposition of accreted Proterozoic terranes 434 with the Archean Wyoming craton (Hansen & Dueker, 2009; Tyson et al., 2002). The 435 northward crustal thickening seems to be indicated by the Proterozoic Moho, highlighted 436 by dark green dots in the P wave image (Figure 10a). However, due to spatial aliasing 437 in both P and S wave images between 20 and 40 km horizontally, the Archean Moho is 438 not well constrained (Hansen & Dueker, 2009). Therefore, we try to suppress the spa-439 tial aliasing effect by refining the stations for every 1.0 km as illustrated in Figure 11. 440 The comparison of the interpolated seismograms shows a good match both in trends and 441 amplitudes of different events. We conduct P wave migration again using the refined data. 442 As shown in Figure 11c, most spatial aliasing artifacts are suppressed and both the Archean 443 Moho (~ 45 km) and the Proterozoic Moho (~ 60 km) can be clearly identified. They are 444 slightly dipping northward to the north of the CBSZ. These are consistent with previ-445 ous studies (Allmendinger et al., 1982; Prodehl et al., 1989; Snelson et al., 1998; Moro-446 zova et al., 2002). However, our Proterozoic Moho seems to be distorted with the CBSZ 447 around 15 km in the horizontal direction (south of the CBSZ), whereas Hansen and Dueker 448 (2009) found it to be continuous beneath the entire array. Nevertheless, we interpreted 449 the CBSZ as the orange shadow zones in Figure 11c. Further investigation for a detailed 450 migration velocity model will be helpful for better imaging and interpretation. 451

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452 4 Discussions

We combine full wavefield passive source RTM with the spectral element method so that it is convenient to handle topographic changes and velocity heterogeneities. Both synthetic and suture zone imaging examples demonstrate the performance of our method.

456

4.1 2-D data interpolation

Migration antialiasing due to spatial frequency components (station spacing) is a 457 longstanding problem in seismic imaging, for example, 5-D data interpolation (Trad, 2009) 458 has been developed to improve the imaging quality. Therefore, it is important to con-459 sider the effects of station spacing for passive RTM (L. Chen et al., 2005; Li et al., 2018). 460 The Nyquist-Shannon sampling theorem suggests that the station spacing is required to 461 be smaller than the apparent half-wavelength to fully construct the wavefield from recorded 462 seismograms at the surface (Gray, 2013; Li et al., 2018). Therefore, given constant crustal 463 P and S wave velocities of 5.8 and 3.2 km/s, respectively, and incident angles of 12° to 464 27°, full construction of 1.0 Hz P waves at near-surface requires a station spacing rang-465 ing from 6.4 to 14.5 km, but 3.5 to 8.0 km for 1.0 Hz S waves. This gives us an upper 466 bound limit for the station spacing. Smaller station spacing is expected because the in-467 cident angle for converted Ps waves could be larger, especially when the subsurface struc-468 tures are complex, like subduction and suture zones. For example, given a high-frequency 469 cut about 2.5 Hz, a 2.0 km station spacing would result in slight spatial aliasing as shown 470 in Figure 12a compared to the result obtained with a 1.0 km station spacing shown in 471 Figure 6g. As expected, a 4.0 km station spacing would result in even stronger spatial 472 aliasing as shown in Figure 12b. Depending on the specific imaging target, 2-D data in-473 terpolation might be necessary for migration. However, most data interpolation (regu-474 larization) strategies developed for seismic exploration, like FX domain trace interpo-475 lation (Spitz, 1991), antileakage Fourier transformation (Xu et al., 2005) or curvelet trans-476 formation (Herrmann & Hennenfent, 2008; Shang et al., 2017) require linearity of sta-477 tion distributions, which is not straightforward to handle 2-D irregularly-spaced (3-D if 478 we consider the station elevations) data. Therefore, we prefer to use the 2-D cubic-spline 479 interpolation (J. Zhang & Zheng, 2015) or the radial function-based method (Shepard, 480 n.d.), which can naturally handle irregularly-spaced data for our migration. We note here 481 that, unlike its first application for interpolating 2-D receiver functions (J. Zhang & Zheng, 482 2015), we need to align the multi-channel data prior to interpolation for each component, 483

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and then shift the interpolated data back according to the onsets of incident waves formigration.

486

4.2 Source time function estimations

It is straightforward to estimate the source time functions for P waves, which has been successfully used for teleseismic full waveform inversion (Y. Wang et al., 2016; K. Wang et al., 2021). However, the estimation of source time functions for S waves should be carefully considered, because of relatively larger incident angles compared to P waves. This results in the leakage of S waves into vertical and transverse components. Therefore, rotation of vertical, radial and transverse components into P, SV and SH polarization directions could be helpful for source time function estimations for S waves (Bostock, 1998; Rondenay, 2009).

495

4.3 Migration velocity models

One advantage of our methodology is that we are able to implement migration based 496 on 3-D velocity models with strong heterogeneities. The accuracy of migration velocity 497 models is essential for mapping seismograms to correct locations and avoiding stacking 498 artifacts (Mora, 1989; He & Liu, 2020). To illustrate the advantages of using more ac-499 curate migration velocity models for imaging, we conduct another migration for slab imag-500 ing, which is shown in Figure 13. The migration velocity model is obtained by smooth-501 ing the true velocity model using a Gaussian function with a radius of 5.0 km both hor-502 izontally and vertically. We speculate that, with a more reasonable migration velocity 503 model, the vertical boundaries (see true model in Figure 6a) on the left and right sides 504 of the slab are better constrained compared to those shown in Figure 6g. However, for 505 our suture zone imaging, a smoothed regional 3-D model is not accurate enough because 506 the grid spacing (~ 25 km) of model US2016 (Shen & Ritzwoller, 2016) is too large to 507 capture velocity heterogeneities in such a small study region. Therefore, teleseismic body 508 wave traveltime tomography (D. Zhao et al., 1992; Tong, 2021) or ambient noise tomog-509 raphy (C. Zhang et al., 2018) could be used to construct a more reasonable migration 510 velocity model for our imaging in the future. 511

512 5 Conclusion

In this study, we propose to solve weak-form solutions to decompose elastic wave-513 fields into vector P and S waves for teleseismic reverse time migration based on the spec-514 tral element method. Both synthetic tests and Cheyenne Belt suture zone imaging demon-515 strate the capability of our method to image complex structures with strong velocity het-516 erogeneity. For linear array migration, our synthetic tests show that teleseismic events 517 with back azimuthal angles parallel to the linear array direction contribute more to sub-518 surface migration images than those away from the linear array direction. However, the 519 latter could still contribute to the image beneath the array thanks to larger Fresnel zone 520 contributions at greater depths. In addition, we reveal several south-dipping structures 521 in the Laramie basin, which are interpreted as the Cheyenne Belt suture zone, and are 522 consistent with geological interpretations from previous studies. For better performance 523 of the migration-based imaging method, 2-D/3-D data interpolation is required to avoid 524 spatial aliasing during the construction of wavefields in the subsurface. 525

526 Open Research

The Laramie array seismic data used in this study can be obtained from the IRIS Data Management Center (https://ds.iris.edu/ds) under the network codes XF. We use SPECFEM3-D Cartesian 4.0.0 (Komatitsch et al., 1999; Komatitsch & Tromp, 2002b, 2002a) published under the GPL 3 license for synthetic and real data simulations. PyGMT (Wessel et al., 2019) download from (https://www.pygmt.org/latest/) is used to plot figures.

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539 Author Contributions

The authors confirm their contribution to the paper as follows: study conception and design: Hejun Zhu; Weak-form solutions were first proposed by Yu Chen, then theoretically derivated and tested by Bin He. Data collection: Bin He; Analysis and interpretation of results: Bin He, Hejun Zhu, David Lumley, Qinya Liu, Hitoshi Kawakatsu and Nozomu Takeuch; Draft manuscript preparation: Bin He, Hejun Zhu. All authors review the results and approve the final version of the manuscript.

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Event	Original Time	$\operatorname{Lon}(^{\circ})$	$\operatorname{Lat}(^{\circ})$	Depth(km)
P1	2000_10_04_14_37_44	-62.5590	11.1240	110.3
P2	2000_10_05_13_39_11	-40.9580	31.7320	10
P3	2000_11_08_06_59_58	-77.8290	7.0420	17
P4	2000_11_29_10_25_13	-70.8860	-24.8690	58.2
P5	2000_12_12_05_26_45	-82.6790	6.0150	10
P6	2001_01_10_16_02_42	-153.2810	56.7744	36.4
$\mathbf{P7}$	2001_01_13_17_33_32	-88.6600	13.0490	60
$\mathbf{P8}$	2001_02_13_14_22_05	-88.9380	13.6710	10
P9	2001_03_24_06_27_53	132.5260	34.0830	50
P10	2001_04_09_09_00_57	-73.1090	-32.6680	11
P11	2001_04_14_23_27_26	141.7680	30.0920	10
S1	2000_10_04_16_58_44	166.9100	-15.4210	23
S2	2000_10_27_04_21_51	140.4600	26.2660	388
S3	2000_10_29_08_37_08	153.9450	-4.7660	50
S4	2000_12_21_01_01_27	151.1220	-5.7060	33
S5	2001_01_09_16_49_28	167.1700	-14.9280	103
S6	2001_04_09_09_00_57	-73.1090	-32.6680	11
$\mathbf{S7}$	2001_04_14_23_27_26	141.7680	30.0920	10
S8	2001_04_28_04_49_53	-176.9370	-18.0640	351.8

Table 1: Teleseismic event information for passive source RTM using the Laramie array.



Figure 1: Strong-form solutions for wavefield decomposition based on Equation 6. Panel (a) shows the X component of the back-propagating elastic wavefield. Panels (b) and (d) represent the separated P and S wavefields, respectively. Panel (c) shows the difference between the reference (a) and the summation of strong-form decomposed P (b) and S (d) waves. The Z component has a similar phenomenon, which is not shown here.



Figure 2: Same as Figure 1 but shows the decomposed wavefields based on Equations 7 and 8.



Figure 3: Synthetic teleseismic data with topographic changes in the model shown in Figure 4b. Panels (a) and (b) are Z and X component seismograms, respectively. The plane wave incident depth is 120 km, with a back azimuthal angle of 90°, and an incident angle of 27°. Therefore, X components could be considered as the radial components of teleseismic waves. The black dashed rectangle highlights the effect of topographic changes on recorded data. The magenta short lines denote the time window used to isolate waveforms for RTM.



Figure 4: passive source RTM using 12 teleseismic events with incident-angles ranging from 12° to 27°. The back-azimuthal angle for the first six events is 90° and is 270° for the others. The black line on top of the image represents the station locations for injecting adjoint sources during migration. (a) Without considering the topographic variation, i.e., the stations are assumed to be at 0 km depth. (b) By considering the topographic variations, the stations are modeled at the correct elevations.



Figure 5: Synthetic Z (a) and X (b) components for the first teleseismic event with an incident angle of 27° and a back azimuthal angle of 90°. The magenta short lines denote the time window used to isolate waveforms for RTM. In panel (a), P denotes teleseismic P waves, MP1 and MP2 represent different-order P wave multiples. MSlab together with the dashed blue line, represent slab-related multiples. B is used to denote reflection artifacts due to absorbing boundary conditions. In panel (b), S denotes Ps-converted waves, MS1 and MS2 are corresponding P-multiple converted waves. Slab1 and Slab2 are slab-related Ps-converted waves.



Figure 6: 3-D subducting slab imaging using 12 teleseismic events. The incident angle of the incident wavefield ranges from 12° to 27°. The back-azimuthal angle for events 1, 3, 7, 9 and 11 is 90° and 270° for the others. Panels (a) and (b) show the true velocity profile along the X (west-east) and Y (noth-source) directions for generating synthetic datasets. Panels (c) and (d) show the corresponding migration velocity along the same profiles, which is smoothed from the two-layer background model after removing the slab. Panels (e) and (f) show the stacked images from the first four teleseismic events, while panels (g) and (h) show the final image stacked over all teleseismic events. Panels (e-h) share the same color bars. M1, M2 and M3 in panels (e) and (g) represent multiple artifacts. The artifacts pointed by the green arrows are caused by strong scattering at the sharp edges of the slab.



Figure 7: Similar to panel (g) in Figure 6, but the back azimuthal angles used for these two examples are 0° for events 1, 3, 7, 9 and 11, and 180° for the others (a), which are 45° and 225° in panel (b). The green arrows are used to compare imaged Moho interfaces with Figure 6.



Figure 8: Geological settings of the study region after Hansen and Dueker (2009). The background shows the topography. Several shear zones: LPSZ, Laramie Peak shear zone; FLSZ, Farwell Mt. Lester Mt. suture zone; SFSZ, Soda Creek-Fish Creek shear zone; SGSZ, Skin Gulch shear zone, are denoted by orange lines to illustrate the complexity of the subsurface structures. Other geographic features: SM, Sierra Madres; MB, Medicine Bow Mountains and LM, Laramie Mountains are also labeled. The Cheyenne Belt suture (CB) is denoted by the white line, dashed where it is inferred. WY and CO denotes Wyoming and Colorado, respectively. The black rectangle in the upper-left inset map indicates the location of our study region in North America.



Figure 9: Caption next page.

Figure 9: Telseismic events and data used for migration. Red dots in Panels (a) and (d) represent the distributions of teleseismic P and S events. The light blue dashed line indicates the direction of the Laramie array. Panels (b) and (c) show the vertical (Z) and radial (R) components of the 5th (the blue dot in panel a) teleseismic P events recorded by the Laramie array. The background black solid lines denote the selected high-quality data (some traces are removed due to low signal-to-noise ratio), while the red solid lines represent the 2-D cubic-spline interpolated data at each station. Panels (e) and (f) represent the vertical and radial components of the 6th teleseismic S events. The magenta lines overlaying the seismograms denote either the onsets of P or S waves.



Figure 10: The passive source RTM image beneath the Laramie array. On the top of the figure, we show the elevation of raw (back triangles) and interpolated (red triangles) stations. CB represents the inferred CBSZ location on the surface. Panels (a) and (b) show the images obtained using teleseismic P and S events, respectively. Panel (c) shows the image obtained using interpolated teleseismic P events. The orange belt indicates the interpreted CBSZ. The magenta and dark green dots indicate the interpreted Archean Moho (ArM) and Proterozoic Moho (PtM).



Figure 11: Regularization of the recorded teleseismic data to a fine grid. Panels (a) and (b) compare the raw and interpolation station locations. The interpolated station locations are obtained with GMT projection between the first (L01) and last station (L31) positions of the Laramie array along the great circle for every 1.0 km, which is not on a straight line after the UTM projection. Panels (c) and (d) show comparisons of the raw data (black) and interpolated seismograms (red) for vertical (Z) and radial (R) components, respectively.



Figure 12: Imaging with sparse station spacing. The imaging parameters are the same as panel (g) in Figure 6 except that the station spacing is 2 km and 4 km for panels (a) and (b), respectively. The aliasing artifacts are illustrated by black arrows.



Figure 13: Imaging with a more accurate background migration velocity model. Panels (a) and (b) show the migration velocity profiles along the X (west-east) and Y (noth-source) directions. Panels (c) and (d) show the corresponding RTM images. The green arrows are used to highlight the improvement in imaging of the slab boundaries compared to panel (g) in Figure 6.

Supporting Information for "Passive source reverse time migration based on the spectral element method"

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1. Text S1

Introduction

In this supporting information, we show how to get the weak-form solutions for wavefield decomposition. If you are familiar with the spectral element method, please go to section 3 directly. For detailed deviations, please refer to Fichtner (2010).

Text S1. Notes for understanding the weak solutions

1. Weak Solutions for the elastic wave-equation

The strong displacement-stress variant of the equations of motion:

$$\rho(\mathbf{x})\ddot{\mathbf{u}}(\mathbf{x},t) - \nabla \cdot \sigma(\mathbf{x},t) = \mathbf{f}(\mathbf{x},t) \quad , \tag{1}$$

$$\sigma(\mathbf{x},t) = \mathbf{C}(\mathbf{x}) : \nabla \mathbf{u}(\mathbf{x},t) \quad , \tag{2}$$

subject to the boundary and initial conditions

$$\mathbf{n} \cdot \sigma |_{\mathbf{x} \in \partial G} = 0$$
 , $\mathbf{u} |_{t=0} = \dot{\mathbf{u}} |_{t=0} = 0$. (3)

For the moment we disregard dissipation, i.e., the time dependence of the elastic tensor **C**. Multiply Eq.(1) by an arbitrary, differentiable, time-independent test function \mathbf{w} and integrating over G gives

$$\int_{G} \rho \mathbf{w} \cdot \ddot{\mathbf{u}} d^{3} \mathbf{x} - \int_{G} \mathbf{w} \cdot (\nabla \cdot \sigma) d^{3} \mathbf{x} = \int_{G} \mathbf{w} \cdot \mathbf{f} d^{3} \mathbf{x} \quad .$$
(4)

Invoking the identity

$$\mathbf{w} \cdot (\nabla \cdot \sigma) = \nabla \cdot (\mathbf{w} \cdot \sigma) - \nabla \mathbf{w} : \sigma \quad . \tag{5}$$

Poof:

$$\nabla \cdot (\mathbf{w} \cdot \sigma) = \partial_i (w_j \sigma_{ij}) = (\partial_i w_j) \sigma_{ij} + w_j (\partial_i \sigma_{ij}) = \nabla \mathbf{w} : \sigma + \mathbf{w} \cdot (\nabla \cdot \sigma) \quad .$$

Together with Gauss's theorem, yields,

$$\int_{G} \rho \mathbf{w} \cdot \ddot{\mathbf{u}} d^{3} \mathbf{x} - \int_{\partial G} \mathbf{w} \cdot \boldsymbol{\sigma} \cdot \mathbf{n} d^{2} \mathbf{x} + \int_{G} \nabla \mathbf{w} : \boldsymbol{\sigma} d^{3} \mathbf{x} = \int_{G} \mathbf{w} \cdot \mathbf{f} d^{3} \mathbf{x} \quad .$$
(6)

Upon inserting the free surface boundary condition, Eq. (6) condenses to

$$\int_{G} \rho \mathbf{w} \cdot \ddot{\mathbf{u}} d^{3} \mathbf{x} + \int_{G} \nabla \mathbf{w} : \sigma d^{3} \mathbf{x} = \int_{G} \mathbf{w} \cdot \mathbf{f} d^{3} \mathbf{x} \quad .$$
(7)

Finding a weak solution to the equations of motion means finding a displacement field \mathbf{u} that satisfies the integral relation Eq. (7) and

$$\int_{G} \mathbf{w} \cdot \sigma d^{3} \mathbf{x} = \int_{G} \mathbf{w} \cdot (\mathbf{C} : \nabla \mathbf{u}) d^{3} \mathbf{x} \quad .$$
(8)

for any test function \mathbf{w} and subject to the initial conditions.

2. Discretisation of the Equations of Motion

By using the Galerkin method, we approximate the *p*-component u_p of the displacement field **u** by a superposition of basis functions

$$\psi_{ijk}(\mathbf{x}) = \psi_{ijk}(x_1, x_2, x_3) \quad , \tag{9}$$

weighted by expansion coefficients u_p^{ijk} :

$$u_p(\mathbf{x},t) \approx \bar{u}_p(\mathbf{x},t) = \sum_{i,j,k=1}^{N+1} u_p^{ijk}(t)\psi_{ijk}(\mathbf{x}) \quad .$$
(10)

The corresponding approximation of the stress tensor components σ_{pq} is

$$\sigma_{pq}(\mathbf{x},t) \approx \bar{\sigma}_{pq}(\mathbf{x},t) = \sum_{i,j,k=1}^{N+1} \sigma_{pq}^{ijk}(t) \psi_{ijk}(\mathbf{x}) \quad .$$
(11)

To find a weak solution in the Galerkin sense, we replace the exact weak formulation forms from Eqs. (7) and (8) by the requirement that approximations $\bar{\mathbf{u}}$ and $\bar{\sigma}$ satisfy

$$\begin{split} \int_{G} \rho \mathbf{w} \cdot \ddot{\mathbf{u}} d^{3} \mathbf{x} + \int_{G} \nabla \mathbf{w} : \bar{\sigma} d^{3} \mathbf{x} &= \int_{G} \mathbf{w} \cdot \mathbf{f} d^{3} \mathbf{x} \quad , \\ \int_{G} \mathbf{w} \cdot \bar{\sigma} d^{3} \mathbf{x} &= \int_{G} \mathbf{w} \cdot (\mathbf{C} : \nabla \bar{\mathbf{u}}) d^{3} \mathbf{x} \quad , \end{split}$$

for any test function, $w_{ijk}^p = \psi_{ijk} \mathbf{e}_p$ in the form of

$$\int_{G_e} \rho \psi_{ijk} \mathbf{e}_p \cdot \ddot{\mathbf{u}} d^3 \mathbf{x} + \int_{G_e} \nabla (\psi_{ijk} \mathbf{e}_p) : \bar{\sigma} d^3 \mathbf{x} = \int_{G_e} \psi_{ijk} \mathbf{e}_p \cdot \mathbf{f} d^3 \mathbf{x}$$
(12)

$$\int_{G_e} \psi_{ijk} \mathbf{e}_p \cdot \bar{\sigma} d^3 \mathbf{x} = \int_{G_e} \psi_{ijk} \mathbf{e}_p \cdot (\mathbf{C} : \nabla \bar{\mathbf{u}}) d^3 \mathbf{x} \quad . \tag{13}$$

Here Eqs. (12) and (13) already assume that u_p and σ_{pq} are considered inside an element $G_e \subset \mathbb{R}^3$, where they can be represented by $(N+1)^3$ basis functions.

For the first term on the left-hand side of Eq. (12), we find

$$\mathbb{F}_{qrs}(\rho\ddot{u}_{p}) := \int_{G_{e}} \rho\psi_{qrs}\mathbf{e}_{p} \cdot \ddot{\mathbf{u}}d^{3}\mathbf{x} = \sum_{i,j,k=1}^{N+1} \int_{G_{e}} \rho\psi_{qrs}\ddot{u}_{p}^{ijk}\psi_{ijk}d^{3}\mathbf{x}$$

$$= \sum_{i,j,k=1}^{N+1} \int_{\Lambda} \rho[\mathbf{x}(\xi)]\psi_{qrs}[\mathbf{x}(\xi)]\ddot{u}_{p}^{ijk}(t)\psi_{ijk}[\mathbf{x}(\xi)]J(\xi)d^{3}\xi \quad .$$
(14)

For the basis function $\psi_{ijk}[\mathbf{x}(\xi)]$ we chose the product of three Lagrange polynomial collocated at the GLL points:

$$\psi_{ijk}[\mathbf{x}(\xi)] = l_i(\xi_1) l_j(\xi_2) l_k(\xi_3) \quad , \tag{15}$$

then we have

$$\mathbb{F}_{qrs}(\rho\ddot{u}_p) = \sum_{i,j,k=1}^{N+1} \int_{\Lambda} \rho(\xi) l_q(\xi_1) l_r(\xi_2) l_s(\xi_3) \ddot{u}_p^{ijk}(t) l_i(\xi_1) l_j(\xi_2) l_k(\xi_3) J(\xi^{qrs}) d^3\xi \quad .$$
(16)

Apply the GLL quadrature rule to Eq. (16) yields the following simple expression:

$$\mathbb{F}_{qrs}(\rho\ddot{u}_p) = \sum_{i,j,k=1}^{N+1} \sum_{f,g,h=1}^{N+1} w_f(\xi_1) w_g(\xi_2) w_h(\xi_3) \\
\cdot \rho(\xi^{fgh}) \ddot{u}_p^{ijk}(t) J(\xi^{gh}) l_q^f(\xi_1) l_r^g(\xi_2) l_s^h(\xi_3) l_i^f(\xi_1) l_j^g(\xi_2) l_k^h(\xi_3) \\
= \rho(\xi^{qrs}) w_q(\xi_1) w_r(\xi_2) w_s(\xi_3) \ddot{u}_p^{qrs}(t) J(\xi^{qrs}) ,$$
(17)

here, J represents the Jacobin matrix. For the second term on the left-hand side of Eq. (12), we need to know that, according to the proof for Eq. (5), $\nabla \mathbf{w} : \sigma$ actually represents a double inner product for two matrices, which is in the form of $\mathbf{A} : \mathbf{B} = \sum_{i,j} A_{ij} B_{ij}$.

Therefore, $\nabla \mathbf{w} : \sigma = \sum_{i,j=1}^{3} \partial_j w_i \sigma_{i,j}$ then we have

$$\mathbb{F}_{qrs}(\nabla:\sigma)_{p} := \int_{G_{e}} \nabla(\psi_{qrs}\mathbf{e}_{p}) : \bar{\sigma}d^{3}\mathbf{x} \\
= \int_{G_{e}} \sum_{m,n=1}^{3} \left[\frac{\partial_{n}(\psi_{qrs}e_{p})}{dx_{m}}\right] : \bar{\sigma}d^{3}\mathbf{x} \\
= \int_{G_{e}} \sum_{m,n=1}^{3} \left[\frac{\partial(\psi_{qrs}e_{p}^{n})}{dx_{m}}\right] \bar{\sigma}_{mn}d^{3}\mathbf{x} \\
= \int_{G_{e}} \sum_{m,n=1}^{3} \left[\frac{\partial(\psi_{qrs})}{dx_{m}}\delta_{n}^{p}\bar{\sigma}_{mn}d^{3}\mathbf{x}\right] \\
= \int_{G_{e}} \sum_{m=1}^{3} \left[\frac{\partial(\psi_{qrs})}{dx_{m}}\delta_{n}^{p}\bar{\sigma}_{mp}d^{3}\mathbf{x}\right] \\
= \int_{\Lambda} \sum_{m,n=1}^{3} \frac{\partial(\psi_{qrs})}{d\xi_{n}} \frac{d\xi_{n}}{dx_{m}}(\xi)\bar{\sigma}_{mp}(\xi)J(\xi)d^{3}\xi \\
= \int_{\Lambda} \sum_{m}^{3} \frac{\partial[l_{q}(\xi_{1})l_{s}(\xi_{2})]}{d\xi_{n}} \frac{d\xi_{n}}{dx_{m}}\bar{\sigma}_{mp}[x(\xi)]J[x(\xi)]d^{3}\xi \\
= \int_{\Lambda} \sum_{m}^{3} l_{q}(\xi_{1})l_{r}(\xi_{2})l_{s}(\xi_{3}) \frac{d\xi_{1}}{dx_{m}}\bar{\sigma}_{mp}(\xi)J(\xi)d^{3}\xi \\
+ \int_{\Lambda} \sum_{m}^{3} l_{q}(\xi_{1})l_{r}(\xi_{2})\dot{l}_{s}(\xi_{3}) \frac{d\xi_{2}}{dx_{m}}\bar{\sigma}_{mp}(\xi)J(\xi)d^{3}\xi \\
+ \int_{\Lambda} \sum_{m}^{3} l_{q}(\xi_{1})l_{r}(\xi_{2})\dot{l}_{s}(\xi_{3}) \frac{d\xi_{3}}{dx_{m}}\bar{\sigma}_{mp}(\xi)J(\xi)d^{3}\xi \\$$
(18)

Now if we bring the GLL quadrature rule to Eq. (19), we have

$$\mathbb{F}_{qrs}(\nabla:\sigma)_{p} = \sum_{i,j,k=1}^{N+1} \sum_{m=1}^{3} w_{q}w_{r}w_{s}l_{q}^{i}(\xi_{1})l_{r}^{j}(\xi_{2})l_{s}^{k}(\xi_{3})\frac{d\xi_{1}}{dx_{m}}(\xi^{ijk})\bar{\sigma}_{mp}(\xi^{ijk})J(\xi^{ijk})d^{3}\xi \\
+ \sum_{i,j,k=1}^{N+1} \sum_{m=1}^{3} w_{q}w_{r}w_{s}l_{q}^{i}(\xi_{1})l_{r}^{r}(\xi_{2})l_{s}^{k}(\xi_{3})\frac{d\xi_{2}}{dx_{m}}(\xi^{ijk})\bar{\sigma}_{mp}(\xi^{ijk})J(\xi^{ijk})d^{3}\xi \\
+ \sum_{i,j,k=1}^{N+1} \sum_{m=1}^{3} w_{q}w_{r}w_{s}l_{q}(\xi_{1})l_{r}(\xi_{2})\dot{l}_{s}(\xi_{3})\frac{d\xi_{3}}{dx_{m}}(\xi^{ijk})\bar{\sigma}_{mp}(\xi^{ijk})J(\xi^{ijk})d^{3}\xi \\
= \sum_{m=1}^{3} \sum_{i=1}^{N+1} w_{q}w_{r}w_{s}l_{q}^{i}(\xi_{1})\frac{d\xi_{1}}{dx_{m}}(\xi^{irs})\bar{\sigma}_{mp}(\xi^{irs})J(\xi^{irs})d^{3}\xi \\
+ \sum_{m=1}^{3} \sum_{i=1}^{N+1} w_{q}w_{r}w_{s}l_{r}^{i}(\xi_{2})\frac{d\xi_{2}}{dx_{m}}(\xi^{qis})\bar{\sigma}_{mp}(\xi^{qis})J(\xi^{qis})d^{3}\xi \\
+ \sum_{m=1}^{3} \sum_{i=1}^{N+1} w_{q}w_{r}w_{s}l_{s}^{i}(\xi_{3})\frac{d\xi_{3}}{dx_{m}}(\xi^{qri})\bar{\sigma}_{mp}(\xi^{qri})J(\xi^{qri})d^{3}\xi \\$$
(19)

Repeating the above procedure for the source term in Eq. (12) gives

$$\mathbb{F}_{qrs}(\mathbf{f}_p) := w_q w_r w_s f_p(\xi^{qrs}) J(\xi^{qrs})$$
(20)

It remains to consider the approximate weak form of the constitutive relation as specified by Eq. (13). For the left-hand term:

$$\mathbb{F}_{qrs}(\sigma_{mn}) := \int_{G_e} \psi_{qrs}(\mathbf{e}_m \cdot \bar{\sigma})_n d^3 \mathbf{x}$$

$$= \int_{G_e} \psi_{qrs} \sum_{i,j,k=1}^{N+1} \sigma_{mn}^{ijk}(t) \psi_{ijk}(\mathbf{x}) d^3 \mathbf{x}$$

$$= \int_{\Lambda} \psi_{qrs} \sum_{i,j,k=1}^{N+1} \sigma_{mn}^{ijk}(t) \psi_{ijk}(\xi) J(\xi) d^3 \xi$$

$$= \int_{\Lambda} \sum_{i,j,k=1}^{N+1} \sigma_{mn}^{ijk}(t) l_q(\xi_1) l_r(\xi_2) l_s(\xi_3) l_i(\xi_1) l_j(\xi_2) l_k(\xi_3) J(\xi) d^3 \xi$$

$$= w_p(\xi_1) w_r(\xi_2) w_s(\xi_3) \sigma_{mn}^{qrs}(t) J(\xi^{qrs}) \quad ,$$

$$(21)$$

while the right-hand term could be simplified by

$$\begin{split} \mathbb{F}_{qrs}(\mathbf{C}:\nabla\mathbf{u})_{mn} &:= \int_{G_{e}} [\psi_{qrs}\mathbf{e}_{m} \cdot (\mathbf{C}:\nabla\mathbf{u})]_{n} d^{3}\mathbf{x} \\ &= \int_{G_{e}} \psi_{qrs} \sum_{a,b=1}^{3} C_{mrab}(\nabla\mathbf{u})_{ab} d^{3}\mathbf{x} \\ &= \int_{G_{e}} \int_{a,b=1}^{3} \sum_{i,j,k=1}^{N+1} \frac{\partial \psi_{ijk}}{dx_{a}} u_{b}^{ijk} d^{3}\mathbf{x} \\ &= \int_{A} \sum_{a,b=1}^{3} \sum_{i,j,k=1}^{N+1} \psi_{qrs}C_{mab} \frac{\partial (l_{i}(\xi_{1})l_{j}(\xi_{2})l_{k}(\xi_{3}))}{d\xi_{1}} \frac{d\xi_{1}}{dx_{a}} u_{b}^{ijk} d^{3}\xi \\ &+ \int_{A} \sum_{a,b=1}^{3} \sum_{i,j,k=1}^{N+1} \psi_{qrs}C_{mab} \frac{\partial (l_{i}(\xi_{1})l_{j}(\xi_{2})l_{k}(\xi_{3}))}{d\xi_{2}} \frac{d\xi_{2}}{dx_{a}} u_{b}^{ijk} J d^{3}\xi \\ &+ \int_{A} \sum_{a,b=1}^{3} \sum_{i,j,k=1}^{N+1} \psi_{qrs}C_{mab} \frac{\partial (l_{i}(\xi_{1})l_{j}(\xi_{2})l_{k}(\xi_{3}))}{d\xi_{2}} \frac{d\xi_{2}}{dx_{a}} u_{b}^{ijk} J d^{3}\xi \\ &+ \int_{A} \sum_{a,b=1}^{3} \sum_{i,j,k=1}^{N+1} \psi_{qrs}C_{mab} \frac{\partial (l_{i}(\xi_{1})l_{j}(\xi_{2})l_{k}(\xi_{3}))}{d\xi_{3}} \frac{d\xi_{3}}{dx_{a}} u_{b}^{ijk} J d^{3}\xi \\ &= \int_{A} \sum_{a,b=1}^{3} \sum_{i,j,k=1}^{N+1} [l_{q}(\xi_{1})l_{r}(\xi_{2})l_{s}(\xi_{3})]C_{mab}l_{i}(\xi_{1})l_{j}(\xi_{2})l_{k}(\xi_{3}) \frac{d\xi_{1}}{dx_{a}} [\xi_{1}]u_{b}^{ijk} J(\xi) d^{3}\xi \\ &+ \int_{A} \sum_{a,b=1}^{3} \sum_{i,j,k=1}^{N+1} [l_{q}(\xi_{1})l_{r}(\xi_{2})l_{s}(\xi_{3})]C_{mab}l_{i}(\xi_{1})l_{j}(\xi_{2})l_{k}(\xi_{3}) \frac{d\xi_{2}}{dx_{a}} [\xi_{1}]u_{b}^{ijk} J(\xi) d^{3}\xi \\ &+ \int_{A} \sum_{a,b=1}^{3} \sum_{i,j,k=1}^{N+1} [l_{q}(\xi_{1})l_{r}(\xi_{2})l_{s}(\xi_{3})]C_{mab}l_{i}(\xi_{1})l_{j}(\xi_{2})l_{k}(\xi_{3}) \frac{d\xi_{2}}{dx_{a}} [\xi_{1}]u_{b}^{ijk} J(\xi) d^{3}\xi \\ &+ \int_{A} \sum_{a,b=1}^{3} \sum_{i,j,k=1}^{N+1} [l_{q}(\xi_{1})l_{r}(\xi_{2})l_{s}(\xi_{3})]C_{mab}l_{i}(\xi_{1})l_{j}(\xi_{2})l_{b}(\xi_{3}) \frac{d\xi_{1}}{dx_{a}} [\xi_{1}]u_{b}^{ijk} J(\xi) d^{3}\xi \\ &+ \int_{A} \sum_{a,b=1}^{3} \sum_{i,j,k=1}^{N+1} [l_{q}(\xi_{1})l_{r}(\xi_{2})l_{s}(\xi_{3})]C_{mab}l_{i}^{j}(\xi_{1})l_{j}^{j}(\xi_{2})l_{b}^{j}(\xi_{3}) \frac{d\xi_{1}}{dx_{a}} [\xi^{ijk} u_{b}^{ijk} J(\xi) d^{3}\xi \\ &+ \int_{A} \sum_{a,b=1}^{3} \sum_{i,j,k=1}^{N+1} [l_{q}(\xi_{1})l_{r}(\xi_{2})l_{s}^{k}(\xi_{3})]C_{mab}l_{i}^{j}(\xi_{1})l_{j}^{j}(\xi_{2})l_{b}^{j}(\xi_{3}) \frac{d\xi_{1}}{dx_{a}} [\xi^{ijk} u_{b}^{ijk} J(\xi) d^{3}\xi \\ &+ \int_{A} \sum_{a,b=1}^{N+1} w_{f} w_{g} w_{b} \sum_{a,b=1}^{3} \sum_{i,j,k=1}^{N+1} [l_{q}(\xi_{1})l_{r}^{j}(\xi_{2})l_{s}^{k}(\xi_{3})]C_$$

Let's see if the above equation is right or not (with the 2D case for simple) with the Voigt

notation for the tensor index as:

$$ij = 11 \ 22 \ 33 \ 23, 32 \ 13, 31 \ 12, 21$$
$$\Downarrow = \Downarrow \ (23)$$
$$\alpha = 1 \ 2 \ 3 \ 4 \ 5 \ 6$$

$$C_{ijkl} \Rightarrow C_{\alpha\beta} = \begin{bmatrix} C_{11} & C_{12} & C_{13} & C_{14} & C_{15} & C_{16} \\ C_{12} & C_{22} & C_{23} & C_{24} & C_{25} & C_{26} \\ C_{13} & C_{23} & C_{33} & C_{34} & C_{35} & C_{36} \\ C_{14} & C_{24} & C_{34} & C_{44} & C_{45} & C_{46} \\ C_{15} & C_{25} & C_{35} & C_{45} & C_{55} & C_{56} \\ C_{16} & C_{26} & C_{36} & C_{46} & C_{56} & C_{66} \end{bmatrix}$$
(24)

For the isotropic case, it only has 2 independent elements:

$$C_{\alpha\beta} = \begin{bmatrix} K + 4\mu/3 & K - 2\mu/3 & K - 2\mu/3 & 0 & 0 & 0 \\ K - 2\mu/3 & K + 4\mu/3 & K - 2\mu/3 & 0 & 0 & 0 \\ K - 2\mu/3 & K - 2\mu/3 & K + 4\mu/3 & 0 & 0 & 0 \\ 0 & 0 & 0 & \mu & 0 & 0 \\ 0 & 0 & 0 & 0 & \mu & 0 \\ 0 & 0 & 0 & 0 & 0 & \mu \end{bmatrix}$$
(25)

$$\sigma_{11}^{qrs} = c_{1111} \sum_{i=1}^{N+1} [\dot{l}_{i}^{i}(\xi_{1}) \frac{d\xi_{1}}{dx_{1}} u_{1}^{is} + l_{1}^{i}(\xi_{2}) \frac{d\xi_{2}}{dx_{1}} u_{1}^{qi}] \\ + c_{1112} \sum_{i=1}^{N+1} [\dot{l}_{i}^{i}(\xi_{1}) \frac{d\xi_{1}}{dx_{1}} u_{2}^{is} + l_{1}^{i}(\xi_{2}) \frac{d\xi_{2}}{dx_{2}} u_{1}^{qi}] \\ + c_{1121} \sum_{i=1}^{N+1} [\dot{l}_{i}^{i}(\xi_{1}) \frac{d\xi_{1}}{dx_{2}} u_{1}^{is} + l_{1}^{i}(\xi_{2}) \frac{d\xi_{2}}{dx_{2}} u_{1}^{qi}] \\ + c_{1122} \sum_{i=1}^{N+1} [\dot{l}_{i}^{i}(\xi_{1}) \frac{d\xi_{1}}{dx_{2}} u_{2}^{is} + l_{1}^{i}(\xi_{2}) \frac{d\xi_{2}}{dx_{2}} u_{2}^{qi}] \\ + c_{1122} \sum_{i=1}^{N+1} [\dot{l}_{i}^{i}(\xi_{1}) \frac{d\xi_{1}}{dx_{1}} u_{2}^{is} + l_{1}^{i}(\xi_{2}) \frac{d\xi_{2}}{dx_{2}} u_{2}^{qi}] \\ + c_{1212} \sum_{i=1}^{N+1} [\dot{l}_{i}^{i}(\xi_{1}) \frac{d\xi_{1}}{dx_{1}} u_{2}^{is} + l_{1}^{i}(\xi_{2}) \frac{d\xi_{2}}{dx_{2}} u_{1}^{qi}] \\ + c_{1212} \sum_{i=1}^{N+1} [\dot{l}_{i}^{i}(\xi_{1}) \frac{d\xi_{1}}{dx_{2}} u_{1}^{is} + l_{1}^{i}(\xi_{2}) \frac{d\xi_{2}}{dx_{2}} u_{1}^{qi}] \\ + c_{1222} \sum_{i=1}^{N+1} [\dot{l}_{i}^{i}(\xi_{1}) \frac{d\xi_{1}}{dx_{2}} u_{2}^{is} + l_{1}^{i}(\xi_{2}) \frac{d\xi_{2}}{dx_{2}} u_{1}^{qi}] \\ + c_{1212} \sum_{i=1}^{N+1} [\dot{l}_{i}^{i}(\xi_{1}) \frac{d\xi_{1}}{dx_{2}} u_{2}^{is} + l_{1}^{i}(\xi_{2}) \frac{d\xi_{2}}{dx_{2}} u_{1}^{qi}] \\ + c_{1212} \sum_{i=1}^{N+1} [\dot{l}_{i}^{i}(\xi_{1}) \frac{d\xi_{1}}{dx_{1}} u_{2}^{is} + l_{1}^{i}(\xi_{2}) \frac{d\xi_{2}}{dx_{2}} u_{1}^{qi}] \\ + c_{2112} \sum_{i=1}^{N+1} [\dot{l}_{i}^{i}(\xi_{1}) \frac{d\xi_{1}}{dx_{2}} u_{2}^{is} + l_{1}^{i}(\xi_{2}) \frac{d\xi_{2}}{dx_{2}} u_{2}^{qi}] \\ + c_{2122} \sum_{i=1}^{N+1} [\dot{l}_{i}^{i}(\xi_{1}) \frac{d\xi_{1}}{dx_{2}} u_{2}^{is} + l_{1}^{i}(\xi_{2}) \frac{d\xi_{2}}{dx_{2}} u_{2}^{qi}] \\ + c_{2122} \sum_{i=1}^{N+1} [\dot{l}_{i}^{i}(\xi_{1}) \frac{d\xi_{1}}{dx_{1}} u_{1}^{is} + l_{1}^{i}(\xi_{2}) \frac{d\xi_{2}}{dx_{2}} u_{2}^{qi}] \\ + c_{2212} \sum_{i=1}^{N+1} [\dot{l}_{i}^{i}(\xi_{1}) \frac{d\xi_{1}}{dx_{1}} u_{2}^{is} + l_{1}^{i}(\xi_{2}) \frac{d\xi_{2}}{dx_{2}} u_{1}^{qi}] \\ + c_{2221} \sum_{i=1}^{N+1} [\dot{l}_{i}^{i}(\xi_{1}) \frac{d\xi_{1}}{dx_{1}} u_{2}^{is} + l_{1}^{i}(\xi_{2}) \frac{d\xi_{2}}{dx_{2}} u_{1}^{qi}] \\ + c_{2221} \sum_{i=1}^{N+1} [\dot{l}_{i}^{i}(\xi_{1}) \frac{d\xi_{1}}{dx_{1}} u_{2}^{is} + l_{1}^{i}(\xi_{2}) \frac{d\xi_{2}}{dx_{2}} u_{1}^{qi}] \\ + c_{2221} \sum_{i$$

$$+c_{2222}\sum_{i=1}^{i=1} [\dot{l}_{i}^{q}(\xi_{1})\frac{d\xi_{1}}{dx_{2}}u_{2}^{is}+\dot{l}_{i}^{1}(\xi_{2})\frac{d\xi_{2}}{dx_{2}}u_{2}^{qi}]$$

Here for the isotropic, $K = \lambda + 2/3\mu$, so that $c_{1111} = \lambda + 2\mu$, $c_{1122} = \lambda$, $c_{1112} = c_{1121} = \mu$, therefore, Eq. (26) could be simplified by

$$\begin{aligned} \sigma_{11}^{qrs} &= (\lambda + 2\mu) \sum_{i=1}^{N+1} [\dot{l}_{i}^{q}(\xi_{1}) \frac{d\xi_{1}}{dx_{1}} u_{1}^{is} + \dot{l}_{i}^{1}(\xi_{2}) \frac{d\xi_{2}}{dx_{1}} u_{1}^{qi}] \\ &+ \lambda \sum_{i=1}^{N+1} [\dot{l}_{i}^{q}(\xi_{1}) \frac{d\xi_{1}}{dx_{2}} u_{2}^{is} + \dot{l}_{i}^{1}(\xi_{2}) \frac{d\xi_{2}}{dx_{2}} u_{2}^{qi}] \\ \sigma_{12}^{qrs} &= \mu \sum_{i=1}^{N+1} [\dot{l}_{i}^{q}(\xi_{1}) \frac{d\xi_{1}}{dx_{1}} u_{2}^{is} + \dot{l}_{i}^{1}(\xi_{2}) \frac{d\xi_{2}}{dx_{2}} u_{1}^{qi}] \\ &+ \mu \sum_{i=1}^{N+1} [\dot{l}_{i}^{q}(\xi_{1}) \frac{d\xi_{1}}{dx_{2}} u_{1}^{is} + \dot{l}_{i}^{1}(\xi_{2}) \frac{d\xi_{2}}{dx_{2}} u_{1}^{qi}] \\ \sigma_{21}^{qrs} &= \mu \sum_{i=1}^{N+1} [\dot{l}_{i}^{q}(\xi_{1}) \frac{d\xi_{1}}{dx_{1}} u_{2}^{is} + \dot{l}_{i}^{1}(\xi_{2}) \frac{d\xi_{2}}{dx_{1}} u_{2}^{qi}] \\ &+ \mu \sum_{i=1}^{N+1} [\dot{l}_{i}^{q}(\xi_{1}) \frac{d\xi_{1}}{dx_{2}} u_{1}^{is} + \dot{l}_{i}^{1}(\xi_{2}) \frac{d\xi_{2}}{dx_{2}} u_{1}^{qi}] \\ \sigma_{22}^{qrs} &= \lambda \sum_{i=1}^{N+1} [\dot{l}_{i}^{q}(\xi_{1}) \frac{d\xi_{1}}{dx_{1}} u_{1}^{is} + \dot{l}_{i}^{1}(\xi_{2}) \frac{d\xi_{2}}{dx_{2}} u_{1}^{qi}] \\ &+ (\lambda + 2\mu) \sum_{i=1}^{N+1} [\dot{l}_{i}^{q}(\xi_{1}) \frac{d\xi_{1}}{dx_{2}} u_{2}^{is} + \dot{l}_{i}^{1}(\xi_{2}) \frac{d\xi_{2}}{dx_{2}} u_{2}^{qi}] \end{aligned}$$

3. Isotropic weak solutions for the PS decoupling

Given the equation for separating amplitude-preserved vector S wave fields as

$$\mathbf{u}^{\mathbf{s}} = -\nabla \times \left(v_s^2 \nabla \times \mathbf{u} \right) \quad . \tag{28}$$

To get a weak solution, we use the test function as

$$\int_{G_e} \mathbf{w} \cdot \mathbf{u}^{\mathbf{s}} d^3 \mathbf{x} = -\int_{G_e} \mathbf{w} \cdot \nabla \times (v_s^2 \nabla \times \mathbf{u}) d^3 \mathbf{x} \quad .$$
(29)

Since $a \cdot (b \times c) = b \cdot (c \times a) = c \cdot (a \times b)$, and note $\Phi := v_s^2 \nabla \times \mathbf{u}$ therefore,

$$\int_{G_e} \mathbf{w} \cdot \mathbf{u}^{\mathbf{s}} d^3 \mathbf{x} = -\int_{G_e} \mathbf{w} \cdot (\nabla \times \mathbf{\Phi}) d^3 \mathbf{x}$$
$$= \int_{G_e} \nabla \times \mathbf{w} \cdot \mathbf{\Phi} d^3 \mathbf{x} \quad .$$
(30)

For any test function, $w_{ijk}^p = \psi_{ijk} \mathbf{e}_p$, the right hand of Eq. (1.30) has the form of

$$\mathbb{F}_{qrs}(\Phi)_{p} := \int_{G_{e}} \nabla \times \psi_{qrs} \mathbf{e}_{p} \cdot \bar{\Phi} d^{3} \mathbf{x} \\
= \int_{G_{e}} \sum_{m,n,p=1}^{3} \epsilon_{mnp} \frac{\partial \psi_{qrs} \mathbf{e}_{p}}{\partial x_{n}} \Phi_{m} d^{3} \mathbf{x} \\
= \int_{G_{e}} \sum_{m,n,p=1}^{3} \epsilon_{mnp} \dot{l}_{q}(\xi_{1}) l_{r}(\xi_{2}) l_{s}(\xi_{3}) \frac{\partial \xi_{1}}{\partial x_{n}}(\xi) \Phi_{m} \mathbf{e}_{p} J(\xi) d^{3} \xi \\
+ \int_{G_{e}} \sum_{m,n,p=1}^{3} \epsilon_{mnp} l_{q}(\xi_{1}) \dot{l}_{r}(\xi_{2}) l_{s}(\xi_{3}) \frac{\partial \xi_{2}}{\partial x_{n}}(\xi) \Phi_{m} \mathbf{e}_{p} J(\xi) d^{3} \xi \\
+ \int_{G_{e}} \sum_{m,n,p=1}^{3} \epsilon_{mnp} l_{q}(\xi_{1}) l_{r}(\xi_{2}) \dot{l}_{s}(\xi_{3}) \frac{\partial \xi_{3}}{\partial x_{n}}(\xi) \Phi_{m} \mathbf{e}_{p} J(\xi) d^{3} \xi \\
= \sum_{m,n,p=1}^{3} \epsilon_{mnp} \sum_{i=1}^{N+1} w_{i} w_{r} w_{s} \dot{l}_{q}^{i}(\xi_{1}) \frac{\partial \xi_{1}}{\partial x_{n}} (\xi^{irs}) \Phi_{m} \mathbf{e}_{p} J(\xi^{irs}) d^{3} \xi \\
+ \sum_{m,n,p=1}^{3} \epsilon_{mnp} \sum_{i=1}^{N+1} w_{q} w_{i} w_{s} \dot{l}_{q}^{i}(\xi_{2}) \frac{\partial \xi_{2}}{\partial x_{n}} (\xi^{qis}) \Phi_{m} \mathbf{e}_{p} J(\xi^{qis}) d^{3} \xi \\
\sum_{m,n,p=1}^{3} \epsilon_{mnp} \sum_{i=1}^{N+1} w_{q} w_{r} w_{i} \dot{l}_{q}^{i}(\xi_{3}) \frac{\partial \xi_{3}}{\partial x_{n}} (\xi^{qri}) \Phi_{m} \mathbf{e}_{p} J(\xi^{qri}) d^{3} \xi \\$$
(31)

Now let's solve $\Phi := v_s^2 \nabla \times \mathbf{u}$ by neglecting the v_s^2 term, similarly, for any test function,

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$$\begin{split} \mathbb{F}_{qrs}[\nabla \times \mathbf{u}]_m &:= \int_{G_e} \mathbf{w} \cdot \nabla \times \mathbf{u} d^3 \mathbf{x} \\ &= \int_{G_e} [\psi_{qrs} \mathbf{e}]_m \cdot [\nabla \times \mathbf{u}]_m d^3 \mathbf{x} \\ &= \int_{G_e} [\psi_{qrs} \mathbf{e}]_m \cdot [\nabla \times \mathbf{u}]_m d^3 \mathbf{x} \\ &= \int_{G_e} [\psi_{qrs} \mathbf{e}]_m \cdot [\sum_{m,n,p=1}^3 \epsilon_{mnp} \frac{\partial u_p}{\partial x_n} e_m] d^3 \mathbf{x} \\ &= \int_{G_e} \sum_{m,n,p=1}^3 \psi_{qrs} \epsilon_{mnp} [\frac{\partial u_p}{\partial \xi_1} \frac{\partial \xi_1}{\partial x_n} + \frac{\partial u_p}{\partial \xi_2} \frac{\partial \xi_2}{\partial x_n} + \frac{\partial u_p}{\partial \xi_3} \frac{\partial u_p}{\partial x_n} e_m d^3 \mathbf{x} \\ &= \int_{G_e} \sum_{m,n,p=1}^3 \psi_{qrs} \epsilon_{mnp} e_m J(\xi) \sum_{i,j,k=1}^{N+1} i_i(\xi_1) l_j(\xi_2) l_k(\xi_3) \frac{\partial \xi_1}{\partial x_n}(\xi) d^3 \xi \\ &+ \int_{G_e} \sum_{m,n,p=1}^3 \psi_{qrs} \epsilon_{mnp} e_m J(\xi) \sum_{i,j,k=1}^{N+1} l_i(\xi_1) l_j(\xi_2) l_k(\xi_3) \frac{\partial \xi_2}{\partial x_n}(\xi) d^3 \xi \\ &+ \int_{G_e} \sum_{m,n,p=1}^3 \psi_{qrs} \epsilon_{mnp} e_m J(\xi) \sum_{i,j,k=1}^{N+1} l_i(\xi_1) l_j(\xi_2) l_k(\xi_3) \frac{\partial \xi_3}{\partial x_n}(\xi) d^3 \xi \\ &+ \int_{G_e} \sum_{m,n,p=1}^3 \psi_{qrs} \epsilon_{mnp} e_m J(\xi) \sum_{i,j,k=1}^{N+1} l_i(\xi_1) l_j(\xi_2) l_k(\xi_3) \frac{\partial \xi_3}{\partial x_n}(\xi) d^3 \xi \\ &+ \int_{G_e} \sum_{m,n,p=1}^3 \psi_{qrs} \epsilon_{mnp} e_m J(\xi) \sum_{i,j,k=1}^{N+1} l_i(\xi_1) l_j(\xi_2) l_k(\xi_3) \frac{\partial \xi_3}{\partial x_n}(\xi) d^3 \xi \\ &+ \int_{G_e} \sum_{m,n,p=1}^3 \psi_{qrs} \epsilon_{mnp} e_m J(\xi) \sum_{i,j,k=1}^{N+1} l_i(\xi_1) l_j(\xi_2) l_k(\xi_3) \frac{\partial \xi_3}{\partial x_n}(\xi) d^3 \xi \\ &+ \int_{G_e} \sum_{m,n,p=1}^3 \epsilon_{mnp} e_m J(\xi) \sum_{i,j,k=1}^{N+1} l_i(\xi_1) l_j(\xi_2) l_k(\xi_3) \frac{\partial \xi_3}{\partial x_n}(\xi) d\xi_2 \\ &+ \sum_{f,g,h=1}^{N+1} w_f w_g w_h l_q^d(\xi_1) l_r^g(\xi_2) l_s^h(\xi_3) \sum_{m,n,p=1}^3 \epsilon_{mnp} e_m J(\xi) d^{fgh} \sum_{i,j,k=1}^{N+1} l_i^f(\xi_1) l_j^g(\xi_2) l_k^h(\xi_3) \frac{\partial \xi_3}{\partial x_n}(\xi) \\ &= w_q w_r w_s \sum_{m,n,p=1}^3 \epsilon_{mnp} e_m J(\xi) d^{grs} \sum_{i=1}^{N+1} l_i^f(\xi_1) l_j^g(\xi_2) l_k^h(\xi_3) \frac{\partial \xi_3}{\partial x_n}(\xi) \\ &+ w_q w_r w_s \sum_{m,n,p=1}^3 \epsilon_{mnp} e_m J(\xi) d^{grs} \sum_{i=1}^{N+1} l_i^f(\xi_1) \frac{\partial \xi_1}{\partial x_n}(\xi_{qrs}) \\ &+ w_q w_r w_s \sum_{m,n,p=1}^3 \epsilon_{mnp} e_m J(\xi) d^{grs} \sum_{i=1}^{N+1} l_i^f(\xi_3) \frac{\partial \xi_3}{\partial x_n}(\xi_{qrs}) \\ &+ w_q w_r w_s \sum_{m,n,p=1}^3 \epsilon_{mnp} e_m J(\xi) d^{grs} \sum_{i=1}^{N+1} l_i^f(\xi_3) \frac{\partial \xi_3}{\partial x_n}(\xi_{qrs}) \\ &+ w_q w_r w_s \sum_{m,n,p=1}^3 \epsilon_{mnp} e_m J(\xi) d^{grs} \sum_{i=1}^{N+1} l_i^f(\xi_3) \frac{\partial \xi_3}{\partial x_n}(\xi_{qrs}) \\ &+ w_q w_r w_s \sum_{m,n,p=1}^3 \epsilon_{mnp} e_m J(\xi) d^{grs} \sum_{i=1}^$$

If given the equation for separating amplitude-preserved vector **P** wave fields as

$$\mathbf{u}^{\mathbf{p}} = \nabla(v_p^2 \nabla \cdot \mathbf{u}) \quad , \tag{33}$$

then the weak solution could be

$$\int_{G_e} \mathbf{w} \cdot \mathbf{u}^{\mathbf{p}} d^3 \mathbf{x} = \int_{G_e} \mathbf{w} \cdot \nabla (v_p^2 \nabla \cdot \mathbf{u}) d^3 \mathbf{x} \quad . \tag{34}$$

According to the Wiki (https : $//en.m.wikipedia.org/wiki/Vector_calculus_identities$), the integral by parts of the vector dot product is in the form of

$$\iiint \alpha \nabla \cdot A d\mathbf{V} = \oint_{\partial V} \alpha A \cdot d\mathbf{S} - \iiint A \cdot \nabla \alpha d\mathbf{V} \quad . \tag{35}$$

According to Eq. (3), we get the integral by paths of Eq. (34) in the form of

$$\int_{G_e} \mathbf{w} \cdot \mathbf{u}^{\mathbf{p}} d^3 \mathbf{x} = \int_{\partial G} \alpha \mathbf{w} \cdot \mathbf{n} d^2 \mathbf{x} - \int_{G_e} \alpha \nabla \cdot \mathbf{w} \quad . \tag{36}$$

If we take the boundary condition into the first term of the last equation, we have

$$\int_{G_e} \mathbf{w} \cdot \mathbf{u}^{\mathbf{p}} d^3 \mathbf{x} = -\int_{G_e} \alpha \nabla \cdot \mathbf{w} \quad , \tag{37}$$

where we note $\alpha = v_p^2 \nabla \cdot u$. Now let's try to get the weak form solution,

$$\begin{split} \mathbb{F}_{qrs} [\nabla \cdot \mathbf{w}]_{p} &:= \int_{G_{e}} [\nabla \cdot (\psi_{qrs} \mathbf{e}_{\mathbf{p}})] \alpha^{qrs} d^{3} \mathbf{x} \\ &= \int_{G_{e}} \frac{\partial \psi^{qrs}}{\partial x_{p}} \alpha d^{3} \mathbf{x} \\ &= \int_{G_{e}} l_{q} l_{r} l_{s} \frac{\partial \xi_{1}}{\partial x_{p}} \alpha(\xi) J(\xi) d^{3} \xi \\ &+ \int_{G_{e}} l_{q} l_{r} l_{s} \frac{\partial \xi_{2}}{\partial x_{p}} \alpha(\xi) J(\xi) d^{3} \xi \\ &+ \int_{G_{e}} l_{q} l_{r} l_{s} \frac{\partial \xi_{3}}{\partial x_{p}} \alpha(\xi) J(\xi) d^{3} \xi \\ &= \sum_{ijk}^{N+1} w_{i} w_{j} w_{k} l_{q}^{i} l_{r}^{j} l_{s}^{k} \frac{\partial \xi_{2}}{\partial x_{p}} (\xi^{ijk}) \alpha(\xi^{ijk}) J(\xi^{ijk}) \\ &+ \sum_{ijk}^{N+1} w_{i} w_{j} w_{k} l_{q}^{i} l_{r}^{j} l_{s}^{k} \frac{\partial \xi_{2}}{\partial x_{p}} (\xi^{ijk}) \alpha(\xi^{ijk}) J(\xi^{ijk}) \\ &= \sum_{i=1}^{N+1} w_{i} w_{r} w_{s} l_{q}^{i} \frac{\partial \xi_{3}}{\partial x_{p}} (\xi^{irs}) \alpha(\xi^{irs}) J(\xi^{irs}) \\ &+ \sum_{i=1}^{N+1} w_{i} w_{r} w_{s} l_{q}^{i} \frac{\partial \xi_{2}}{\partial x_{p}} (\xi^{qis}) \alpha(\xi^{qis}) J(\xi^{irs}) \\ &+ \sum_{i=1}^{N+1} w_{i} w_{r} w_{s} l_{r}^{i} \frac{\partial \xi_{2}}{\partial x_{p}} (\xi^{qis}) \alpha(\xi^{qis}) J(\xi^{irs}) \\ &+ \sum_{i=1}^{N+1} w_{i} w_{r} w_{s} l_{r}^{i} \frac{\partial \xi_{3}}{\partial x_{p}} (\xi^{qri}) \alpha(\xi^{qri}) J(\xi^{irs}) \\ &+ \sum_{i=1}^{N+1} w_{i} w_{r} w_{s} l_{r}^{i} \frac{\partial \xi_{3}}{\partial x_{p}} (\xi^{qri}) \alpha(\xi^{qri}) J(\xi^{irs}) \\ &+ \sum_{i=1}^{N+1} w_{i} w_{r} w_{s} l_{r}^{i} \frac{\partial \xi_{3}}{\partial x_{p}} (\xi^{qri}) \alpha(\xi^{qri}) J(\xi^{irs}) \\ &+ \sum_{i=1}^{N+1} w_{i} w_{r} w_{s} l_{r}^{i} \frac{\partial \xi_{3}}{\partial x_{p}} (\xi^{qri}) \alpha(\xi^{qri}) J(\xi^{irs}) \\ &+ \sum_{i=1}^{N+1} w_{i} w_{r} w_{s} l_{r}^{i} \frac{\partial \xi_{3}}{\partial x_{p}} (\xi^{qri}) \alpha(\xi^{qri}) J(\xi^{irs}) \\ &+ \sum_{i=1}^{N+1} w_{i} w_{r} w_{s} l_{r}^{i} \frac{\partial \xi_{3}}{\partial x_{p}} (\xi^{qri}) \beta(\xi^{qri}) J(\xi^{irs}) \\ &+ \sum_{i=1}^{N+1} w_{i} w_{r} w_{s} l_{r}^{i} \frac{\partial \xi_{3}}{\partial x_{p}} (\xi^{qri}) \beta(\xi^{qri}) J(\xi^{irs}) \\ &+ \sum_{i=1}^{N+1} w_{i} w_{r} w_{s} l_{r}^{i} \frac{\partial \xi_{3}}{\partial x_{p}} (\xi^{qri}) \beta(\xi^{qri}) J(\xi^{irs}) \\ &+ \sum_{i=1}^{N+1} w_{i} w_{r} w_{s} l_{r}^{i} \frac{\partial \xi_{3}}{\partial x_{p}} (\xi^{qri}) \beta(\xi^{qri}) J(\xi^{qri}) \\ &+ \sum_{i=1}^{N+1} w_{i} w_{r} w_{s} l_{r}^{i} \frac{\partial \xi_{3}}{\partial x_{p}} (\xi^{qri}) \beta(\xi^{qri}) J(\xi^{qri}) \\ &+ \sum_{i=1}^{N+1} w_{i} w_{r} w_{s} l_{r}^{i} \frac{\partial \xi_{3}}{\partial x_{p}} (\xi^{qri}) \beta(\xi^{qri}) J(\xi^{qri}) \\ &+ \sum_{i=1}^{N+1} w_{i} w_{r} w_{s} l_{r}^{i} \frac{\partial \xi_{3}}{\partial x_{p}} ($$

It is not necessary to get this weak form of this term $\alpha = v_p^2 \nabla \cdot u$, since it is fairly easy.

References

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