

1 **A 3D Shear Velocity Model of the Crust and**  
2 **Uppermost Mantle Beneath the United States from**  
3 **Ambient Seismic Noise**

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**Abstract.**

In an earlier study, *Bensen et al.* [2008] measured surface wave dispersion curves from ambient noise using 203 stations across North America, which resulted in Rayleigh and Love wave dispersion maps from 8 - 70 s period and 8 - 20 s period, respectively. We invert these maps in a two-step procedure to determine a three-dimensional (3D) shear wave velocity model ( $V_s$ ) of the crust and uppermost mantle beneath much of the contiguous US. The two steps are a linearized inversion for a best fitting model beneath each grid node, followed by a Monte-Carlo inversion to estimate model uncertainties. In general, a simple model parameterization is sufficient to achieve acceptable data fit, but a Rayleigh/Love discrepancy at periods from 10 to 20 sec is observed in which a simple isotropic model systematically misfits Rayleigh and Love waves in some regions. Crustal features observed in the model include sedimentary basins such as the Anadarko, Green River, Williston Basins as well as California's Great Valley and the Mississippi Embayment. The east-west velocity dichotomy between the stable eastern US and the tectonically deformed western US is shown to be abrupt in the crust and uppermost mantle, but is not coincident in these regions; crustal high velocity material tends to lap over the high velocities of the uppermost mantle. The Rayleigh/Love discrepancy between 10 and 20 sec period is crustal in origin and is observed in a number of regions, particularly in extensional provinces such as the Basin and Range. It can be resolved by introducing radial anisotropy in the lower or middle crust with  $V_{sh} > V_{sv}$  by about 1%.

## 1. Introduction

33 Seismic tomography on both global and regional scales has been performed in recent  
34 years covering all or part of the continental United States. The resulting models, however,  
35 have had either limited geographic extent or relatively low resolution. Recent studies have  
36 shown that surface wave ambient noise tomography (ANT) helps to fill the gap between  
37 regional and continental or global scale tomographic models (e.g., *Shapiro et al.* [2005], *Yao*  
38 *et al.* [2006], *Moschetti et al.* [2007], *Lin et al.* [2007], *Yang et al.* [2007a]). Nevertheless,  
39 constraints from ANT on 3D models of the crust and uppermost mantle have been applied  
40 mainly at regional scales (e.g., *Yang et al.* [2007a], *Yao et al.* [2006]). We show that  
41 ANT can be applied to produce 3D structural information at the continental scale and  
42 that ANT helps to diminish the typical resolution/coverage trade-off that characterizes  
43 earthquake based studies on this scale. Seismic data now emerging from Earthscope's  
44 USArray provide the potential for further improvement in resolution for which our model  
45 may serve as a useful reference.

46 This study is an extension of work presented by *Bensen et al.* [2007] and *Bensen et al.*  
47 [2008]. *Bensen et al.* [2007] presented a technique for computing reliable empirical Green's  
48 functions (EGF) from long sequences of ambient noise. They also presented an automated  
49 procedure to measure the dispersion of EGFs as well as selection criteria to ensure that  
50 only high-quality signals are retained. Using these methods, *Bensen et al.* [2008] estimated  
51 maps of Rayleigh and Love wave group and phase speed across the US. Using 203 stations  
52 across North America (labeled as black triangles in Figure 1) for up to two years of  
53 ambient noise data, they developed surface wave dispersion maps on a  $0.5^\circ \times 0.5^\circ$  grid.

54 They constructed dispersion maps from 8 - 70 s period for Rayleigh waves and 8 - 20  
55 s period for Love waves. These dispersion maps form the basis for the current study.  
56 Aspects of the work by *Bensen et al.* [2007] and *Bensen et al.* [2008] are summarized here  
57 as appropriate.

58 Regional investigations of surface wave propagation and dispersion in the United States  
59 date back over 30 years (e.g., *Lee and Solomon* [1978]). Tomographic studies using in-  
60 creasing volumes of data in the US (e.g., *Alsina et al.* [1996], *van der Lee and Nolet* [1997],  
61 *Godey et al.* [2003], *Li et al.* [2003], *Marone et al.* [2007], *Nettles and Dziewonski* [2008])  
62 have presented dispersion maps and models that have been improving resolution over sim-  
63 ilar studies at global scales (e.g., *Trampert and Woodhouse* [1996], *Ekström et al.* [1997],  
64 *Ritzwoller et al.* [2002]). A large number of regional studies also have been performed  
65 to investigate the seismic structure of North America. Among these are tomographic  
66 studies in regions such as the Rio Grande Rift (e.g., *Gao et al.* [2004]), Cascadia (e.g.,  
67 *Ramachandran et al.* [2005]), California (e.g., *Thurber et al.* [2006]), the Rocky Moun-  
68 tains (e.g., *Yuan and Dueker* [2005]) and the eastern US (e.g., *van der Lee* [2002]), to  
69 name a few recent studies among many others. Many refraction studies have provided  
70 profiles across North America, including CD-ROM (e.g., *Karlstrom et al.* [2002]), Deep  
71 Probe (e.g., *Snelson et al.* [1998]) and others. Receiver functions have provided valuable  
72 constraints on crustal thickness and structure in parts of the continent (e.g., *Crotwell and*  
73 *Owens* [2005]).

74 ANT complements these methods and possesses several features that commend its use.  
75 First, within the context of a seismic array, high path density can be achieved with paths  
76 contained entirely within the study region, minimizing bias from structures outside the

77 region of interest. Second, station locations are known precisely, unlike earthquake lo-  
78 cations. Third, phase velocity measurements from ambient noise are free from an initial  
79 source phase (*Lin et al.* [2007]), which reduces uncertainty compared with earthquake  
80 derived measurements. Fourth, ambient noise dispersion measurements are repeatable,  
81 which allows measurement uncertainties to be estimated (*Bensen et al.* [2008]). Fifth, the  
82 bandwidth of ambient noise dispersion measurements (i.e., 6 - 100 s period) constrains  
83 the structure both of the crust and uppermost mantle. In contrast, it is difficult across  
84 much of the US to obtain earthquake based surface wave dispersion measurements below  
85  $\sim 15$  s period. Previous surface wave studies, therefore, obtained high-quality dispersion  
86 measurements predominantly at longer periods and, therefore, reported velocity structure  
87 predominantly in the mantle (e.g., *Shapiro and Ritzwoller* [2002], *van der Lee and Fred-*  
88 *eriksen* [2005], *Nettles and Dziewonski* [2008]). Body wave studies of similar geographic  
89 extent also provide only weak constraints on crustal structure (e.g., *Grand* [1994], *Grand*  
90 [2002]).

91 The 3D model derived here will be useful to improve earthquake locations in some  
92 regions, aid receiver function studies, and provide a starting model for other investiga-  
93 tions across the US. This may be especially important in the context of the advancing  
94 USArray/Transportable Array experiment.

## 2. Data

95 The data used in this study are the Rayleigh and Love wave group and phase speed  
96 dispersion maps from *Bensen et al.* [2008]. These maps are based on Rayleigh and Love  
97 wave group and phase speed dispersion measurements obtained from EGFs computed  
98 between the stations shown in Figure 1. Dispersion measurements are made on EGFs

99 created by cross-correlating long ambient noise time series using the data processing and  
100 measurement techniques described in detail by *Bensen et al.* [2007] and *Lin et al.* [2007].  
101 Nearly 20,000 paths are used for this experiment and up to 13 unique measurements  
102 from different temporal subsets of the two-year time series along each path are computed  
103 for each wave type. Measurement uncertainties are estimated from the repeatability of  
104 the measurements across the temporal subsets. An automated Frequency Time Analysis  
105 (FTAN) is used to measure the dispersion of these Rayleigh and Love wave signals (*Levshin*  
106 *et al.* [1972], *Bensen et al.* [2007]). *Bensen et al.* [2008] developed acceptance criteria to  
107 ensure that only EGFs of high quality are retained. Starting with nearly 20,000 paths  
108 across the United States and Canada, a maximum of 8,932 paths remained after selection.  
109 The result was group and phase speed tomography maps for Rayleigh waves between 8  
110 and 70 s period and between 8 and 20 s for Love waves. Low signal quality for Love waves  
111 at longer periods causes the narrower bandwidth and apparently results from higher local  
112 noise on horizontal components, particularly in the eastern US. The resulting bandwidth  
113 presents sensitivity to shear velocity from the surface into the upper mantle to a depth  
114 of about 150 km, as seen in Figure 3. Although uncertainty estimates were presented on  
115 the raw dispersion measurements, local uncertainty estimates were not produced on the  
116 resulting dispersion maps.

117 Starting with the set of Rayleigh and Love wave group and phase speed dispersion  
118 maps at different periods presented by *Bensen et al.* [2008], we construct local dispersion  
119 curves at each point on a  $0.5^\circ \times 0.5^\circ$  grid across the US. This process is similar to many  
120 previous studies (e.g., *Ritzwoller and Levshin* [1998], *Villaseñor et al.* [2001], *Shapiro and*  
121 *Ritzwoller* [2002], *Weeraratne et al.* [2003], and others).

122 For the 3D inversion, at each grid point we need an uncertainty value for each period  
123 and measurement type. *Bensen et al.* [2008] did not provide this information. *Shapiro*  
124 *and Ritzwoller* [2002] assigned uncertainty based on the overall RMS tomography misfit  
125 weighted by resolution. Their uncertainties were geographically invariant except in regions  
126 of very low resolution. In our study, there is much more variability in data coverage and  
127 quality and we require geographically variable uncertainties. In the interior of the US,  
128 much of the uncertainty in the dispersion maps derives from the subjectivity of the choices  
129 made in regularization and damping. Near the periphery, however, uncertainty grows due  
130 to relatively poorer data coverage and quality.

131 To address these factors, we create a set of dispersion maps at each period and wave  
132 type by varying regularization and smoothing parameters systematically in the inversion  
133 (*Barmin et al.* [2001]). The minimum and maximum velocity at each point for each  
134 period then define an uncertainty window for that wave type. The uncertainties in the  
135 interior of the US, therefore, reflect the confidence in our ability to localize the dispersion  
136 information, in contrast with raw measurement errors that reflect the repeatability of the  
137 measurements. Within the maps, the regions of greatest uncertainty occur near significant  
138 velocity anomalies. The Love wave group speed dispersion curves display much greater  
139 variability upon varying regularization and smoothing, and we discard them because of  
140 our much lower confidence in their robustness. Finally, we increase uncertainties near the  
141 edges of the study region based on estimated resolution which degrades near the edges  
142 of the maps. For reference, the 500 km resolution contour for the 16 s Rayleigh wave  
143 phase speed map is shown in Figure 2. The mean uncertainty over all periods for the  
144 measurements used in this study is shown in Figure 4. The uncertainty values we assign

145 are smaller than RMS tomography misfit values from *Bensen et al.* [2008] at all periods  
146 for all wave types, but remain quite conservative.

147 In performing the Monte-Carlo sampling, we did not vary the  $V_p/V_s$  or  $V_p/\rho$  ratios.  
148 Doing so, mainly affects the model in the upper crust, affecting the mean model minimally  
149 but increasing model uncertainty.

150 In summary, the uncertainties assigned to the dispersion maps are subjective, but, on  
151 average, represent our confidence in the maps quantitatively. The uncertainties in the  
152 resulting 3D model should be understood in these terms. More rigorous uncertainties  
153 will require a different method of surface wave tomography. Fortunately, advances in this  
154 direction are on the horizon (e.g., *Lin et al.* [2009]).

### 3. Methods

155 Two commonly used methods exist for estimating shear wave velocity structure from  
156 surface wave dispersion measurements. The first is linearized waveform fitting as de-  
157 scribed by *Snieder* [1988], *Nolet* [1990] and others. This technique has been used in many  
158 geographical settings with earthquake surface wave signals, including the US (*van der Lee*  
159 *and Nolet* [1997]). The second method, which we adopt, is a two-stage procedure in which  
160 period specific 2D tomographic maps created from the dispersion measurements are used  
161 first to produce dispersion curves at each geographic grid point. The dispersion curves  
162 are then inverted for 1D  $V_s$  structure at all grid points and the 1D models are compiled  
163 to obtain a 3D volume. This procedure has been described by *Shapiro and Ritzwoller*  
164 [2002], *Yang et al.* [2007b], and elsewhere.

165 Our specific approach to the second stage of inversion divides into two further steps.  
166 The first step is a linearized inversion of the dispersion curves for the 1D velocity structure



167 at each grid point similar to the method of *Yang and Forsyth* [2006]. However, the best  
168 fitting model does not account for the non-uniqueness of the inverse problem; a variety of  
169 acceptable models may be created that fit the data within the estimated uncertainties. In  
170 the second step, for this reason, we perform a Monte-Carlo search of a corridor of model  
171 space defined by the results of the linearized inversion. From this we define an ensemble  
172 of velocity models that fit the data acceptably. These two steps are outlined further  
173 below. The linearized inversion procedure only uses Rayleigh and Love wave phase speed  
174 measurements while Rayleigh wave group speed measurements are also included in the  
175 Monte-Carlo procedure.

### 3.1. Starting Models, Parameterization, and Allowed Variations

176 Both the linearized inversion and the Monte-Carlo resampling of model space require a  
177 starting model. In the linearized inversion, we observe faster and more stable convergence  
178 by using unique starting models at each geographic point. For this purpose, we extract  $V_s$   
179 values from the 3D model of *Shapiro and Ritzwoller* [2002]. The procedure also requires  
180 values of P-wave speed ( $V_p$ ) and density ( $\rho$ ). We use the average continental  $V_p/V_s$   
181 ratios of 1.735 in the crust and 1.756 in the mantle from *Chulick and Mooney* [2002] who  
182 found little deviation from these values across the US. Furthermore, surface waves are less  
183 sensitive to  $V_p$  than  $V_s$  except in the uppermost crust. Density ( $\rho$ ) is assigned similarly  
184 using a  $\rho/V_s$  ratio of 0.81 as described by *Christensen and Mooney* [1995]. Following  
185 previous work (e.g., *Weeraratne et al.* [2003]; *Yang and Forsyth* [2006]), we parameterize  
186 each model with 18 layers. Three crustal layers are used where the top layer thickness is  
187 set at the greater of 2 km or the sediment thickness from the model of *Laske and Masters*  
188 [1997]. The depth to the Moho was extracted from *Bassin et al.* [2000]. These two inputs

189 define a thin upper crustal layer and a thick middle to lower crustal layer. The lower  
190 crustal layer was separated into two layers of equal thickness defining the middle and  
191 lower crust. The 15 layers in the mantle are between 20 and 50 km thick and extend  
192 to 410 km depth, but are relatively unconstrained by our data beneath 150 km. An  
193 illustration of the parameterization is shown in Figure 5a. In the linearized inversion, the  
194 velocities of all layers are allowed to change although regularization is applied to ensure  
195 smoothness, as discussed in Section 3.2 below.  $V_p/V_s$  and  $\rho/V_s$  are maintained at the  
196 values stated above. Finally, only the thicknesses of the lower crust and uppermost mantle  
197 are permitted to change. However, if poor data fit is observed, we perturb the upper and  
198 middle crustal layer thicknesses (while maintaining the initial crustal thickness) and the  
199 inversion is rerun.

200 For Monte-Carlo sampling we use the result of the linearized inversion as a starting  
201 model. However, we also impose an explicit requirement of monotonically increasing  
202 crustal velocity with depth. Within our study area, *Wilson et al.* [2003] and *Ozalaybey*  
203 *et al.* [1997] found evidence for a low-velocity zone (LVZ) in the crust from localized magma  
204 bodies and regional partial melt, respectively. Using receiver functions and surface wave  
205 dispersion to constrain the crust, *Ozalaybey et al.* [1997] allowed  $\sim 20$  crustal layers. At a  
206 variety of locations, their crustal LVZ was often 5 km or less in thickness. These crustal  
207 LVZs are of insufficient vertical extent for us to image reliably. Furthermore, a model  
208 parameterization using monotonically increasing isotropic crustal velocities still produces  
209 fairly good data fit in most cases. In the mantle, Monte-Carlo sampling of 15 mantle layers  
210 would be prohibitively expensive and would potentially create unrealistic models or require  
211 the additional complexity of a smoothing regularization. For speed and smoothness, we

212 parameterize the mantle with five B-splines. An illustration of this parameterization of  
213 the model is shown in Figure 5b. *Ozalaybey et al.* [1997] found evidence for an upper  
214 mantle LVZ in northwestern Nevada, which is permitted in our mantle parameterization.

215 From the linearized inversion described above, we obtain smooth, simple 1D velocity  
216 profiles at all grid points which typically fit the data reasonably well. For the Monte-Carlo  
217 inversion, we define the allowed range of models based on this best fitting model. First,  
218 we impose a constraint on the permitted excursions from the initial velocity values. The  
219 velocity must be within  $\pm 20\%$  of the initial model in the upper crust and  $\pm 10\%$  in  
220 the lower crust and mantle. We choose this range rather than a specific velocity window  
221 (e.g.,  $\pm 0.5$  km/s) because of the potential for unrealistically low values in the crust.  
222 By comparison, our allowed corridor is wider than that of *Shapiro and Ritzwoller* [2002].  
223 Again, we maintain the  $V_p/V_s$  and  $V_s/\rho$  values stated above. However, the thicknesses  
224 of the crustal layers can now vary while the sum of crustal layers must be within  $\pm 5$  km  
225 of the Crust 2.0 model of *Bassin et al.* [2000]. The  $Q$  model from PREM (*Dziewonski*  
226 *and Anderson* [1981]) is used for the physical dispersion correction, and all models are  
227 reduced to 1 sec period.

228 Complexities probably exist within the crust and upper mantle that may not be well  
229 represented by our simple parameterization. However, if data fit is within uncertainties  
230 in the dispersion maps, we cannot empirically justify a more complicated model without  
231 inclusion of independent information (e.g., receiver functions), which is beyond the scope  
232 of this study.

### 3.2. Linearized Inversion

233 The linearized inversion process uses the starting model described in section 3.1 to create  
 234 predicted dispersion curves. Perturbing the input model provides misfit information and  
 235 iterating converges upon the best-fitting model. The linearized inversion process follows  
 236 the work of *Li et al.* [2003], *Weeraratne et al.* [2003], *Forsyth and Li* [2005], *Yang and*  
 237 *Forsyth* [2006] and others. In this case, the forward code used to compute dispersion  
 238 curves from an input model is based on *Saito* [1988]. Only Rayleigh and Love wave phase  
 239 speed curves are used in the inversion. Rayleigh wave group speed curves are introduced  
 240 in the Monte-Carlo inversion, however.

241 The technique to find the best fitting velocity model is outlined by *Weeraratne et al.*  
 242 [2003] and is based on the iterative least-squares approach of *Tarantola and Valette* [1982].  
 243 *Li et al.* [2003] concisely summarize the approach, which we excerpt here. The solution is  
 244 described by the equation:

$$245 \quad \Delta \mathbf{m} = (\mathbf{G}^T \mathbf{C}_{nn}^{-1} \mathbf{G} + \mathbf{C}_{mm}^{-1})^{-1} (\mathbf{G}^T \mathbf{C}_{nn}^{-1} \Delta \mathbf{d} - \mathbf{C}_{mm}^{-1} [\mathbf{m} - \mathbf{m}_0]) \quad (1)$$

246 where  $\mathbf{m}$  is the current model,  $\mathbf{m}_0$  is the starting model at the outset of each iteration,  
 247 and  $\Delta \mathbf{m}$  is the change to the model.  $\Delta \mathbf{d}$  is the difference between the observed and  
 248 predicted data.  $\mathbf{G}$  is a sensitivity matrix relating changes in  $\mathbf{d}$  to changes in  $\mathbf{m}$ .  $\mathbf{C}_{mm}$   
 249 is the model covariance matrix where non-zero values (we use 0.1) are introduced into  
 250 the off-diagonal terms in order to provide a degree of correlation between velocity values  
 251 obtained for adjacent layers and ensure a reasonable model (i.e., a model without large  
 252 velocity jumps or oscillations).  $\mathbf{C}_{nn}$  is the diagonal data covariance matrix where the  
 253 diagonal elements are calculated from the local uncertainties in the dispersion maps.

254 As a measure of data fit, we use reduced  $\chi^2$  (henceforth  $\chi^2$ ). Unique  $\chi^2$  values are  
 255 computed for Rayleigh wave and Love wave phase speed;  $\chi^2$  is also computed for Rayleigh  
 256 wave group speed in the Monte-Carlo resampling described below.  $\chi^2$  is defined as

$$257 \quad \chi^2 = \frac{1}{n} \sum_{i=1}^n \frac{(\tilde{d}_i - d_i)^2}{\sigma_i^2} \quad (2)$$

258 where  $i$  is the index of the period of the measurement through all wave types used.  
 259 Periods used are on a 2 second grid from 8 - 20 s period and every 5 seconds for 25  
 260 - 70 s period. Therefore,  $n$  is 7 for Love waves and 17 for Rayleigh waves. Thus, in  
 261 the linearized inversion, 24 measurements are used but in the Monte-Carlo inversion, 41  
 262 measurements are applied because Rayleigh wave group speeds are utilized.  $\tilde{d}$  and  $d_i$  are  
 263 the model predicted and measured wave speeds, respectively, and  $\sigma_i$  is the uncertainty  
 264 of the measured velocity unique to each period, wave type, and location, as described in  
 265 Section 2 above. A  $\chi^2$  value of 2 or less represents fairly good data fit, although misfit  
 266 systematics may still exist for  $\chi^2$  ranging from 1.5 to 2. Higher values indicate inferior  
 267 fit, inadequate model parameterization, or underestimated data uncertainties.

268 An example of input data and model output from the linearized inversion is shown in  
 269 Figure 6 for a point in Illinois. For reference, the location of this point is plotted as a grey  
 270 circle in Figure 1. Dispersion observations and associated errors are plotted as error bars  
 271 in Figure 6a. The resulting best fitting model and related dispersion curves produced by  
 272 linearized inversion are shown as thin black lines. For comparison, the starting model and  
 273 the related dispersion curves are shown in Figure 6 as dotted grey lines.

274 Variability in data fit is present in the study area. Figure 7 shows two more examples  
 275 like Figure 6 but with higher resulting  $\chi^2$  values. Considering that the location of data  
 276 used in Figure 7c,d is in an area of particularly good resolution (southern California),

277 the misfit most likely derives from improper model parameterization. In this case, the  
278 short period under-prediction of Love wave speeds and over-prediction of Rayleigh wave  
279 speeds may indicate the need for radial anisotropy in the crust. More discussion of alter-  
280 native parameterizations follows in Section 6.3. Examination of the sensitivity curves in  
281 Figure 3 suggests that higher misfit (e.g., Figure 7a,c) could be due to improper model  
282 parameterization at depths from 0 - 30 km.

### 3.3. Monte-Carlo Resampling and Model Uncertainty Estimation

283 To estimate uncertainties in geophysical inverse problems, model space sampling meth-  
284 ods such as Monte-Carlo methods have been in use for over 40 years (*Keilis-Borok and*  
285 *Yanovskaya* [1967]) and can provide useful uncertainty estimates even when the *a priori*  
286 probability density of solutions is unknown (see *Mosegaard and Tarantola* [1995]). Varia-  
287 tions among Monte-Carlo methods are summarized by *Sambridge and Mosegaard* [2002].  
288 One particular concern in our inverse problem is the tradeoff between velocity values in  
289 the lower crust and uppermost mantle with crustal thickness. This is considered a sig-  
290 nificant problem by *Marone and Romanowicz* [2007] and elsewhere and provides part of  
291 our motivation to estimate model uncertainty. We quantify the variation of acceptable  
292 models and use this variation as an indication of the robustness of the resulting velocity  
293 model. Nevertheless, as discussed in section 2, because the estimates of the uncertainty  
294 of the dispersion maps are subjective, the estimates of model uncertainty are also.

295 The Monte-Carlo procedure is a two-step process that first creates models through  
296 uniformly distributed random perturbations within the permitted corridor around the  
297 model produced by linearized inversion. Second, a random walk is used to refine the  
298 search for acceptable models. Rayleigh wave group and phase and Love wave phase speed

299 dispersion curves are generated for each model using the faster forward code of *Herrmann*  
300 [1987] which we verified agrees well with the code of *Saito* [1988] used in the linearized  
301 inversion. If the predicted dispersion curves match the measured results at an acceptable  
302 level, the model is retained. An acceptable model is defined as one having a  $\chi^2$  value  
303 within 3 times the  $\chi^2$  value obtained from the linearized inversion. Fairly conservative  
304 error estimates result from these choices. In order to accelerate the process of obtaining  
305 a sufficient number of acceptable models, the random walk procedure generates small  
306 perturbations to search adjacent model space for additional acceptable models. After the  
307 random walk identifies an acceptable model, the search re-initializes in the neighborhood of  
308 that model until we construct 100 acceptable models. This number of models is arbitrary,  
309 but appears to be large enough to quantify model uncertainty to form the basis for our  
310 inferences and is computationally tractable. An example of the observed dispersion curves  
311 and the Monte-Carlo results are shown in Figure 8 for points labeled as grey squares in  
312 Figure 1.

313 We select a “favored model” from the set of resulting velocity models. The best-fitting  
314 model is very similar to that determined through linearized inversion and may not rep-  
315 resent the ensemble of models very well. We favor the model closest to the mean of the  
316 distribution, where greater depths are given lesser weight. This captures the essence of the  
317 ensemble but diminishes the occasional problems of lateral roughness found when only the  
318 best fitting velocity models are considered. For illustration, the models identified as most  
319 near the mean of the distribution are plotted in red in Figure 8a,c,e and are, henceforth,  
320 referred to as the “favored models”. Further discussion of model variability across the  
321 study area is reserved for Section 5 below.

#### 4. Crustal Rayleigh/Love Wave Speed Discrepancy

322 The observation of relatively poor data fit in regions of good resolution deserves further  
323 comment. A three-layer crust and multi-layer mantle can usually fit either Rayleigh or  
324 Love wave measurements satisfactorily. However, fitting data to both simultaneously  
325 is more difficult. Figure 9 shows the difference in misfit to Rayleigh and Love waves  
326 phase velocities across the US where, unlike  $\chi^2$ , the sign of the misfit is retained. The  
327 predicted curves are from the “favored model” derived by Monte-Carlo inversion from  
328 which we subtract the observed dispersion at each geographical point and divide this by  
329 the estimated data error. These values are averaged only from 8 - 20 s period. Green  
330 and orange colors signify that the model is faster than an observation at a point. Blue  
331 colors indicate that the model is too slow to fit the observations. The widespread result of  
332 Rayleigh and Love wave speeds being over- and under-predicted, respectively, is apparent.  
333 The period band (8 - 20 s) indicates that the source of this discrepancy lies in the crust.  
334 We, therefore, refer to this as the crustal Rayleigh/Love discrepancy to distinguish it  
335 from the well known mantle Rayleigh/Love discrepancy caused by radial anisotropy due  
336 to olivine alignment in the mantle (e.g., *Dziewonski and Anderson* [1981]). Section 6.3  
337 below discusses possible causes of this observation and the preferred explanation.

#### 5. Results

338 We construct a “favored model” from an ensemble of models that fit the data accept-  
339 ably developed through Monte-Carlo inversion at each grid point. Combining these 1D  
340 isotropic models, we obtain a 3D shear wave velocity model for the continental US with  
341 lateral coverage bounded approximately by the black contour in Figure 2 and depth range



342 from the surface to 150 km. Here, we characterize the model by highlighting examples of  
343 the types of features it contains. The names of features listed in Figure 2.

344 Because the model is over-parameterized, we smooth the model features and soften  
345 abrupt contrasts between layers by vertically averaging in 4 km increments in the crust  
346 and 10 km in the mantle. Thus, a depth section at 10 km is the average from 8 - 12 km  
347 depth. No smoothing is applied across the Moho. In addition, we average model values  
348 from four horizontally adjacent grid nodes (across 1 degree) so that map views represent  
349 a 1 degree average of the original model values. Tests indicate that such smoothing does  
350 not degrade data fit substantially. However, the lateral smoothing does reduce vertical  
351 striping on plots of vertical cross-sections.

### 5.1. Characteristics of the 3D Model

352 Horizontal slices of isotropic shear wave speed at a selection of depths are shown in  
353 Figure 10 including 4 km above (Figure 10c) and 4 km below (Figure 10d) the estimated  
354 Moho depth. The most striking features at 4 km depth (Figure 10a) are several large  
355 sedimentary basins. The Mississippi Embayment and the Green River Basin appear most  
356 strongly. Additionally, the Williston Basin and Anadarko Basin in Montana and Okla-  
357 homa, respectively, clearly appear as slow velocity anomalies. Low velocities associated  
358 with the sediments of the Great Valley in California abut slow crustal velocities of the  
359 Cenozoic Pacific Northwest volcanic province farther north. A trend of generally faster  
360 velocities in the eastern US compared with the western US is also seen. This is observed  
361 at all depths and we refer to it as the east-west crustal “dichotomy” in the US. At 10 km  
362 (Figure 10b), the most pronounced feature is again the deep sediments of the Mississippi  
363 Embayment, which may be partially extended to this depth by the vertical averaging.

364 The crustal velocity dichotomy at this depth is located along the boundary between the  
365 Great Plains and Central Lowlands as will be discussed further in Section 6.2 below.

366 In the lower crust at 4 km above the Moho, Figure 10c shows that the crustal velocity  
367 dichotomy in the central US shifts west to coincide with the transition from the Great  
368 Plains to the Rocky Mountain Front. The slow anomaly in the Basin and Range may  
369 be attributed to high crustal temperatures in this extensional province, as evidenced by  
370 high surface heat flow in the area (see e.g., *Blackwell et al.* [1990]). The fast anomaly  
371 in Michigan may result from regionally thicker crust; a slice at 4 km above the Moho is  
372 at a greater depth than the surrounding region. However, the slower speeds beneath the  
373 Appalachian Highlands to the east is within similarly thick crust, implying that composi-  
374 tional differences between the Appalachian Highlands and the continental shield are the  
375 more likely cause of this velocity anomaly. For reference, the estimated crustal thickness  
376 is shown in Figure 11.

377 In the upper mantle 4 km below the Moho (Figure 10d), the east-west velocity dichotomy  
378 is in a similar but not identical location to the lower crust. This will be discussed further  
379 in Section 6.2 below. East of this transition, more laterally homogeneous mantle velocities  
380 appear. To the west, the prominent slow anomaly below the eastern Basin and Range  
381 corroborates the suggested removal of mantle lithosphere from 10 Ma to present (e.g.,  
382 *Jones et al.* [1994]) and replacement with warmer, low velocity asthenospheric material.  
383 The slow anomaly in the Pacific Northwest can be attributed to the volatilized mantle  
384 wedge residing above the subducting slab. At 80 km depth (Figure 10e), however, the slow  
385 anomaly associated with the mantle wedge is no longer visible, suggesting that this depth  
386 is below or within the subducting slab. Also, a slow mantle velocity anomaly extends in

387 the northwest to southeast direction, roughly following the outline of the entire Basin and  
388 Range province. A similar feature was also observed in the tomographic models of *Alsina*  
389 *et al.* [1996] and others and has been attributed to inflow of warm mantle material during  
390 Cenozoic extension (e.g., *Wernicke et al.* [1988]). At 120 km depth in Figure 10f, features  
391 are similar to 80 km depth, but anomalies are of lower amplitude.

392 The estimated crustal thickness is similar to the starting model of Crust 2.0 (*Bassin*  
393 *et al.* [2000]) and is shown in Figure 11. On average, the crust is 1.6 km thinner than  
394 Crust 2.0 and the RMS difference from Crust 2.0 across the study region is 1.5 km. The  
395 relation of crustal thickness with topography and implications for topographic support or  
396 compensation are discussed in section 6.1.

397 Figure 12 presents a series of vertical cross-sections with locations indicated on the map  
398 in Figure 12a. A smoothed elevation profile is plotted above each cross-section and a  
399 profile of the recovered crustal thickness is overplotted. We use different color scales for  
400 crustal and mantle shear wave speeds. The vertical exaggeration of the cross-sections is  
401 roughly 25:1 and the same horizontal scale is used for N-S and E-W cross-sections.

402 As with the horizontal depth-sections presented in Figure 10, the most pronounced shal-  
403 low crustal velocity anomalies are from sedimentary basins, although vertical smoothing  
404 extends these features to greater depths. Profiles C-C' and F-F', for example, show that  
405 the sediments of the Mississippi Embayment extend inland from the coast for hundreds  
406 of kilometers. The most pronounced velocity contrasts result from the location of the  
407 east-west velocity dichotomy in the crust and upper mantle, as discussed further in Sec-  
408 tion 6.2. Slow mantle velocities extend from the Rocky Mountains to the west and are  
409 particularly low in the Basin and Range.

## 5.2. Model Uncertainties

410 As discussed in section 3.3, a model is considered to be a member of the ensemble of  
411 acceptable models if its  $\chi^2$  misfit is within three times that of the best fitting model from  
412 the linearized inversion. The standard deviation ( $\sigma$ , not to be confused with measurement  
413 uncertainty) of this ensemble at each grid point then defines the confidence in the velocity  
414 values through depth and across the study region. Average values for  $\sigma$  versus depth are  
415 shown in Figure 13a. Except near the surface, the average value of uncertainty is about  
416 1.5%, with this value increasing slightly with depth. The RMS of velocities as a function  
417 of depth taken over the entire region of study is also shown in Figure 13 to be about 3%,  
418 except near the surface. Thus, lateral velocity anomalies are, on average, about twice the  
419 size of the uncertainties.

420 Figure 14 shows the amplitude and distribution of  $\sigma$  across the study region at the  
421 depths presented in Figure 10. At 4 km depth,  $\sigma$  is greatest near the edges of the study  
422 area, in part due to higher expected data errors caused by lower resolution. Low  $\sigma$   
423 values at 10 km depth (Figure 14b) through much of the study region are due to the lack  
424 of boundaries above and below this layer with which to trade-off. A parameterization  
425 that allows topography on more crustal layers would generate greater middle crustal  
426 uncertainty. In the lower crust (Figure 14c),  $\sigma$  is greater than in the mid-crust due to  
427 the tradeoff between wave speed and crustal thickness. Similar values are observed in the  
428 upper mantle (Figure 14d) due to the same tradeoff. At 80 km (Figure 14e),  $\sigma$  is lower  
429 than at shallower depths and is more uniform. The uniformity extends to about 120 km  
430 depth (Figure 14f), although the amplitude of  $\sigma$  increases slightly at this depth due to

431 poorer sensitivity at greater depths as indicated in Figure 3. Below 150 km depth, the  
432 model is very poorly constrained.

433 Figure 13b shows the average standard deviation in the dispersion curves produced by  
434 the ensemble of acceptable models. Greater variability in model velocity values in the  
435 uppermost crustal layer results in the higher standard deviation values at short periods  
436 (i.e.,  $< 15$  s period). Rayleigh and Love wave phase speed variability is nearly constant  
437 at 0.5% while the Rayleigh wave group speed variability is higher due to the higher.

## 6. Discussion

438 A detailed interpretation of the estimated 3D model is beyond the scope of this paper.  
439 We discuss three specific questions and emphasize using the model uncertainties to address  
440 them. First, we consider the relation between crustal thickness and surface topography  
441 across the US. Second, we constrain the location of the east/west velocity dichotomy in the  
442 lower crust and uppermost mantle. Finally, we present alternative model parameterizations  
443 in the attempt to illuminate the cause of the crustal Rayleigh/Love velocity discrepancy  
444 discussed in Section 4 above.

### 6.1. Topographic Compensation

445 The relation between surface topography, crustal thickness, and crust and mantle veloc-  
446 ities allows qualitative conclusions to be drawn regarding the support for high topography  
447 in the US. In general, surface topography within the US is not well correlated with crustal  
448 thickness. For example, the north-south profiles in Figure 10 display very little relation  
449 between the surface and Moho topography. Profile E-E', in particular, reveals crustal  
450 thickness to be anti-correlated with topography and substantial Moho topography exists

451 under regions with almost no surface topography in Profiles F-F' and G-G'. In addition,  
452 the Basin and Range province is characterized by high elevations, but the crust is relatively  
453 thin. In all of these areas, however, high elevations with relatively thin crust are under-  
454 lain by a slower and presumably less dense crust and mantle, indicative of a Pratt-type  
455 of compensation or dynamical support for the topography. There are exceptions, how-  
456 ever. Running from west to east along Profile B-B', the highest elevations coincide with  
457 a mantle that is relatively slow and the crust is thick. Farther east in the Great Plains,  
458 the thinning crust and decreasing elevation are coincident, suggesting an Airy-type of  
459 compensation.

## 6.2. East-West Shear Velocity Dichotomy

460 The difference in crustal and uppermost mantle shear wave speeds between the faster  
461 tectonically stable eastern US and the slower tectonically active western US is visible in the  
462 horizontal and vertical cross-sections presented in Figures 10 and 12. This is also a feature  
463 of older tomographic models. Here, we use the ensemble of models from the Monte-Carlo  
464 inversion to estimate the location of and uncertainty in this velocity dichotomy.

465 First, Figure 15 presents histograms of velocity values along 40°N within the eastern  
466 and western US for the lower crust and in the mantle at 80 km depth. The values  
467 are taken from the favored model produce by the Monte Carlo inversion. The eastern  
468 and western US are separated approximately by a shear velocity of about 3.75 km/sec  
469 in the lower crust and 4.55 km/sec in the uppermost mantle, but the exact choice of  
470 these values affects our conclusions only slightly. Note first that the two distributions are  
471 nearly disjoint, indicating a strong compositional and/or thermal difference between the  
472 tectonically active western US and the stable eastern US. Secondly, the distribution in

473 the eastern US is more peaked, particularly in the lower crust, demonstrating that the  
474 eastern US is somewhat more homogeneous than the west.

475 To estimate the location of the boundary of the east-west dichotomy, shear velocity  
476 values for the lower crust and at 80 km depth are sorted and ranked by  $V_s$  value for the  
477 ensemble of 100 acceptable models produced by the Monte Carlo inversion at each grid  
478 point. In Figure 16, contours are plotted through the 20th and 80th maps (which can  
479 be thought of as the 20th and 80th percentile values within the ensemble of accepted  
480 models at each point) for values of 3.75 km/s in the lower crust and 4.55 km/s at 80  
481 km depth as grey and black lines, respectively. The separation between the tectonically  
482 active western US and the stable eastern US lies approximately between these contours.  
483 In the lower crust (Figure 16a), the western velocity contrast roughly follows the Rocky  
484 Mountain Front from Wyoming to the south, but veers to the west north of central  
485 Wyoming, crossing the Rocky Mountain front. This east-west contrast occurs abruptly.  
486 In fact, examining the lower crustal velocity values across a variety of latitudes, a velocity  
487 change of roughly 300 m/s typically occurs over less than 100 km laterally. Both the  
488 20th and 80th percentile values are seen in the western US. In the eastern US, the 20th  
489 percentile contour outlines the southeastern edge between the North American craton and  
490 the Appalachian Highlands farther east. This velocity contour does not precisely follow  
491 the western edge of the Appalachian highlands as plotted in Figure 2, which may be due  
492 to the lower resolution in the eastern US. The Mid-Continental Rift (MCR), oriented in a  
493 NNE-SSW direction in the central US, is also apparent. This feature is subtle in velocity  
494 depth- and cross-sections but clearly appears here, with a location that agrees with the  
495 configuration apparent in gravity maps.

496 At 80 km depth in the mantle, a similar set of contours outlines the eastern edge of the  
497 slower western US. However, the location of these contours now aligns better with the  
498 Rocky Mountain Front in the northern part of the study area and lies farther east in the  
499 southern portions. The eastern contour provides an outline of the cratonic lithosphere.

500 In summary, the range of locations is sufficiently narrow to constrain the boundary  
501 of the dichotomy in the lower crust and uppermost mantle and to observe that these  
502 locations are similar but not identical. First, the fact that slower and presumably less  
503 dense mantle material often extends well east of the Rocky Mountain Front suggests that  
504 mantle compensation plays a role in the high topography of that region. Second, the di-  
505 chotomy boundary in the lower crust lies west of the mantle boundary in the western US.  
506 Assuming that this boundary marks the approximate edge of the craton, this means that  
507 the cratonic crust extends out farther from the interior of the craton than the cratonic  
508 mantle. This apparent overhanging of the cratonic crust may be caused by mantle litho-  
509 spheric erosion due to small-scale convection. Third, the lower crustal boundary crosses  
510 the Rocky Mountain front, probably reflective of crustal deformation beneath and west  
511 of the northern Rocky Mountains.

### 6.3. Resolving the Crustal Rayleigh/Love Wave Speed Discrepancy

512 Section 4 documents the systematic misfit of Rayleigh and Love wave phase velocities  
513 below about 20 sec period by a simple isotropic parameterization of the crust with mono-  
514 tonically increasing velocities with depth. Figure 9 presents a summary that shows that,  
515 on average, Rayleigh wave speeds are overpredicted and Love wave speeds are underpre-  
516 dicted by the isotropic model that aims to fit both simultaneously. Figure 17a shows  
517 an example inversion for a point in northwest Utah (located with a grey star in Figure



518 1) illuminating how the estimated isotropic model (red line) predicts Love wave speeds  
519 that are too slow and Rayleigh wave speeds that are too fast, particularly below about  
520 15 sec period. Apparently, the model parameterization is inadequate to fit both types of  
521 data simultaneously. The most likely cause of the problem is either the constraint that  
522 imposes vertical monotonicity within the crust or the fact that only isotropic models are  
523 constructed within the crust. We test both alternatives.

524 To determine whether crustal radial anisotropy can resolve the short period Rayleigh-  
525 Love discrepancy, we allow only the middle crust to be radially anisotropic. The rest  
526 of the model is fixed on the favored model from the isotropic profiles determined from  
527 the Monte-Carlo inversion. We perform a grid search over small perturbations in  $V_s$  in  
528 the middle crust ( $\pm 500$  m/s) which attempts to fit the Rayleigh and Love wave phase  
529 velocity measurements below 25 sec separately. In the inversion with the Rayleigh wave  
530 data alone we recover a set of allowed  $V_{sv}$  values in the middle crust and with the Love  
531 wave data we get a set of allowed  $V_{sh}$  values. The model is isotropic outside the middle  
532 crust. The result for the best fitting radially anisotropic model for the point in northwest  
533 Utah is shown in Figure 17a (blue line). The model itself with bifurcated  $V_{sh}$  and  $V_{sv}$   
534 values is shown in Figure 17b where blues denote  $V_{sv}$  and reds denote  $V_{sh}$  in the middle  
535 crust and the model outside the middle crust is isotropic ( $V_{sh} = V_{sv} = V_s$ ). In general,  
536 allowing radial anisotropy in the middle crust can resolve the Rayleigh - Love discrepancy.  
537 We have also performed the experiment allowing lower crustal radial anisotropy, but on  
538 average it does not fit the data as well as middle crustal anisotropy alone. A combination  
539 of middle and lower crustal radial anisotropy cannot be ruled out, however.

540 Although Love waves are predominantly sensitive to  $Vsh$  and Rayleigh waves to  $Vsv$ ,  
541 there is weak sensitivity of each wave type to the alternate shear wave speed. Thus,  
542 separately inverting Love and Rayleigh waves for  $Vsh$  and  $Vsv$ , respectively, is not fully  
543 accurate. To test the approximation, we performed tests using the anisotropic “MINEOS”  
544 code of *Masters et al.* [2007]. We created synthetic dispersion curves from models possess-  
545 ing radial anisotropy in the crust and then inverted them to estimate the anisotropy using  
546 the procedure outlined above. The approximation we apply recovers the initial model to  
547 within about 5 m/s ( $\sim 0.1\%$ ), which is an order of magnitude smaller than the amplitude  
548 of the dispersion signals that are attempting to explain. The approximation that we use,  
549 therefore, is sufficiently accurate for the inferences drawn here.

550 We have also investigated whether breaking the monotonicity constraint can resolve  
551 the Rayleigh - Love discrepancy. An example inversion in which a fourth crustal layer  
552 has been introduced and the monotonicity constraint has been broken is shown with the  
553 green lines in Figure 17. In this case a low velocity zone (LVZ) is introduced in the  
554 lower crust. Breaking the monotonicity constraint and introducing another crustal layer  
555 improves the fit to the data, but does not resolve the discrepancy as well as allowing  
556 a single middle crustal anisotropic layer. We extended this test across all of Nevada  
557 where radial anisotropy improves data fit and where crustal low velocity zones have been  
558 previously documented. *Ozalaybey et al.* [1997] found thin crustal LVZs ( $\sim 5$  km thick) at  
559 points in this area using a joint receiver function/surface wave technique. For the 93 grid  
560 points tested, our procedure was not able to obtain the quality of fit observed using radial  
561 anisotropy, as the misfit results in Table 1 show. The values contained within the table  
562 are averaged over dispersion measurements from 10 to 20 sec period. We find that the  $\chi^2$

563 misfit with the radially anisotropic crust across Nevada is 1.06, yielding  $\sim 42\%$  variance  
564 reduction compared to the isotropic model with monotonically increasing shear wave  
565 speeds. The non-monotonic isotropic model gives only a 15% variance reduction, with a  
566  $\chi^2$  value of 1.54, and misfit systematics continue in evidence. Breaking the monotonicity  
567 constraint and adding a single crustal layer, therefore, does not allow the data to be as fit  
568 well as by allowing radial anisotropy in a single crustal layer. The introduction of more  
569 crustal layers and the development of more complicated models cannot be formally ruled  
570 out as an alternative, but the layerization will have to be extensive and complicated.

571 Thus, the introduction of radial anisotropy to the model parameterization is most ef-  
572 fective at resolving the discrepancy and we believe radial anisotropy is the most likely  
573 physical cause. The mapping of radial anisotropy in the upper mantle using fundamen-  
574 tal mode Rayleigh and Love waves is a well established technique (e.g., *Tanimoto and*  
575 *Anderson* [1984], *Montagner* [1991]). *Shapiro et al.* [2004] used longer period Rayleigh  
576 and Love wave observations to constrain radial anisotropy in the Tibetan crust, which  
577 they attributed to crystal alignment caused by crustal flow. The widespread search for  
578 crustal radial anisotropy has been hindered by a lack of short period dispersion observa-  
579 tions (below 20 sec period) over extended regions, which ambient noise tomography now  
580 provides.

581 Figure 18a presents the middle crustal radial anisotropy for the best fitting radially  
582 anisotropic model, where green and orange colors indicate positive anisotropy ( $V_{sh} >$   
583  $V_{sv}$ ) and blue colors indicate the reverse. In this compilation, most of the US has crustal  
584 radial anisotropy above the level of  $\pm 1\%$  and most areas have positive anisotropy. This  
585 does not mean, however, that the anisotropy is required to fit the data. To determine

586 this we present in Figure 18b the model with the minimal anisotropy that fits the data  
587 acceptably. In this result, the middle crust across much of the US is white (i.e., isotropic)  
588 and the regions with negative anisotropy largely disappear.

589 There remain in Figure 18b several regions in which radial anisotropy in the middle crust  
590 is required to fit the data. These regions tend to be of two main tectonic types: sedimen-  
591 tary basins and extensional regions. The Anadarko (western Oklahoma), Appalachian,  
592 and Green River (western Wyoming) basins are clearly outlined. In these cases, layering  
593 of sediments may cause different  $V_{sh}$  and  $V_{sv}$  values in the uppermost crust and some im-  
594 provement in data fit is observed by allowing radial anisotropy in the middle crust. These  
595 features may be artifacts, however, caused by poor parameterization of the vertical  $V_s$  ve-  
596 locity gradient in the sediments or perhaps by the strong lateral contrast across which the  
597 Love and Rayleigh waves sample differently (e.g., *Levshin and Ratnikova* [1984]). Crustal  
598 radial anisotropy at about 2 - 4 % is observed through much of the Basin and Range,  
599 extending southeast toward the Rio Grande Rift. The observed radial anisotropy may be  
600 due to crystalline reorganization effected during Cenozoic extension. *Shapiro et al.* [2004]  
601 attributed observed radial anisotropy to the alignment of mica crystals in the crust. The  
602 effects of other compositional organizations, such as aligned cracks (e.g., *Crampin and*  
603 *Peacock* [2005]) or layering (e.g., *Crampin* [1970]), have also been shown to cause seis-  
604 mic anisotropy. The multiplicity of sources of radial anisotropy must be considered when  
605 interpreting these results.

606 Presentation of the 3D distribution of  $V_{sh}$  and  $V_{sv}$  and further investigation of alter-  
607 native parameterizations and physical causes await more exhaustive studies based on the  
608 USArray/Transportable Array.

## 7. Conclusions

609 We present a 3D shear velocity model of the crust and uppermost mantle beneath much  
610 of the continental United States. The model is constrained by Rayleigh group and phase  
611 velocity measurements from 8 to 70 s period and Love wave phase velocities from 8 to 20  
612 s, both determined by ambient noise tomography (ANT) presented previously by *Bensen*  
613 *et al.* [2008]. We employ a two-step procedure to obtain shear wave speeds in the crust  
614 and uppermost mantle from the surface to approximately 150 km depth. In the first step,  
615 a linearized inversion is performed to find the best fitting model at each grid point on  
616 a  $0.5^\circ \times 0.5^\circ$  grid across the US. This is followed in the second step by a Monte-Carlo  
617 inversion to estimate the ensemble of models that fit the data acceptably and, hence, to  
618 bound model uncertainties.

619 The 3D model presented here displays higher lateral resolution than earlier models pro-  
620 duced using teleseismic earthquake data on a similar scale. Unexpectedly, the amplitude  
621 of features in the model, however, tend to be muted relative to global models such as that  
622 of *Shapiro and Ritzwoller* [2002]. At the largest scales, the outline of the structural di-  
623 chotomy between the tectonic west and the stable eastern part of the US is clearly defined  
624 in both the crust and uppermost mantle and is observed to be very abrupt. The location  
625 of the transition between the tectonic and stable regions is shown to be similar in the  
626 lower crust and uppermost mantle, but not coincident. In the western US, high velocities  
627 in the crust typically extend further to the west than in the mantle, particularly north of  
628 Colorado. On smaller scales, numerous intriguing features within the model are imaged,  
629 such as sedimentary basins in the shallow crust, the indication of the mid-continental rift  
630 in the lower crust, and the generally variable correlation between surface and Moho to-

631 pography across much of the country. The estimated crustal thickness is similar to model  
632 Crust 2.0 of *Bassin et al.* [2000] across most of the US.

633 The resulting isotropic 3D model systematically misfits Rayleigh and Love wave speeds  
634 between 10 and 20 sec period in some regions, overpredicting Rayleigh wave speeds and un-  
635 derpredicting Love wave speeds. We argue that this Rayleigh/Love discrepancy probably  
636 results from radial anisotropy in the middle and/or lower crust. Crustal radial anisotropy  
637 is required primarily within the Basin and Range and other extensional provinces, with  
638  $V_{sh} > V_{sv}$  by about  $\sim 1\%$  in these regions. A more exhaustive study of the Rayleigh/Love  
639 discrepancy using alternative model parameterizations, higher resolution data (e.g., from  
640 the USArray Transportable Array), and other kinds of data (e.g., receiver functions) is a  
641 natural extension of this work.

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**Table 1.**  $\chi^2$  misfit for Rayleigh and Love waves averaged from 8 to 20 sec period across Nevada. Column 1 lists the method of crustal model parameterization, where “Monotonic Isotropic” denotes 3 crustal layers of monotonically increasing isotropic velocity with depth, “Nonmonotonic Isotropic” is also isotropic but with the monotonicity constraint removed for 4 crustal layers, and “Radial Anisotropy” is where radial anisotropy is allowed in the middle of the 3 crustal layers. Columns 2, 3, and 4 indicate  $\chi^2$  values for Love wave phase speed, Rayleigh wave phase speed, and the average of the two. The final column lists the variance reduction over the monotonic isotropic parameterization.

Param. type	$\chi^2$ -Love	$\chi^2$ -Rayleigh	$\chi^2$ -avg.	Variance Reduction
Monotonic Isotropic	2.21	1.42	1.81	
Nonmonotonic Isotropic	1.45	1.63	1.54	15.2%
Radial Anisotropy	1.05	1.07	1.06	41.6%

**Figure 1.** Map of the study area showing stations used in the experiment as black triangles. Grey circles, squares, and a star are the locations for the examples in Figures 6, 7, 8, and 17.

**Figure 2.** Regions and geographic features. The black contour surrounds the area with lateral resolution better than 500 km for the 16 s Rayleigh wave phase velocity. Tectonic provinces are outlined in red and are labeled (bounded by rectangles) for reference. Features (from east to west) are as follows: Appalachian Highlands (ApH), Ouachita-Ozark Highlands (OH), Central Lowlands (CL), Great Plains (GP), Rocky Mountain Region (RM), Colorado Plateau (CP), Basin and Range (B&R), Columbia Plateau (CP), Sierra Nevada Mountains (SN), and Great Valley (GV). Other features are labeled (bounded by ellipses) as follows: Appalachian Basin (ApB), Michigan Basin (MB), Mississippi Embayment (ME), Mid-continental Rift (MCR), Anadarko Basin (AB), Williston Basin (WB), Rio Grande Rift (RGR), Green River Basin (GRB), Gulf of California (GC), and Pacific Northwest (PNW).

**Figure 3.** Sensitivity kernels for Rayleigh (labeled RC) and Love (labeled LC) wave phase speeds at a selection of periods.

**Figure 4.** Spatially averaged uncertainty across the Rayleigh wave group and phase speed and the Love wave phase speed maps. These are the average values within which we attempt to fit the data.



**Figure 5.** An illustration of the parameterization of the models used to create dispersion curves for (a) the linearized inversion and (b) the Monte-Carlo inversion. Fifteen layers are used in the mantle for the linearized inversion while five B-splines are used in the mantle for the Monte-Carlo inversion.

**Figure 6.** Example of the best fitting model and dispersion curves from the linearized inversion for a point in Illinois. Rayleigh and Love wave phase speed measurements and uncertainties are represented with error bars in (a). The input model in (b) and related dispersion curves in (a) are shown as grey dashed lines. The estimated models and dispersion curves are thin black lines in (b) and (a). The latitude, longitude and approximate location is listed in (b) and labeled as a grey circle in Figure 1. Velocity values at the center of each mantle layer are plotted.

**Figure 7.** Same as Figure 6 but for points in California and Montana, shown as grey circles in Figure 1. The  $\chi^2$  value indicated in (c) is toward the larger end in this study.

**Figure 8.** Examples of the input and output dispersion curves (error bars and grey lines, respectively, in (b), (d), and (f)) and the resulting ensemble of Monte-Carlo models ((a), (c), and (e)). The “favored model” is drawn in red. Locations of the examples presented here are shown as grey squares in Figure 1.

**Figure 9.** Representation of the short-period discrepancy between Rayleigh and Love waves from the isotropic “favored models” that emerge from the Monte Carlo inversion. The difference of the model predicted and measured wave speed is divided by the data error at each point for each period. The results presented here are the average of values from 8 - 20 s period. Greens/oranges indicate that the model is too fast and blues that the model is too slow.

**Figure 10.** A selection of horizontal  $V_s$  depth sections through the isotropic “favored model” from Monte-Carlo inversion. Panels (c) and (d) show the model at 4 km above and below the recovered Moho, respectively.

**Figure 11.** The crustal thickness of the “favored model” from the Monte-Carlo inversion. Crustal thickness is required to be within 5 km of the values of *Bassin et al.* [2000].

**Figure 12.** A selection of  $V_s$  vertical cross sections through the “favored model” from Monte-Carlo inversion. The locations of the cross-sections are indicated in (a) and the horizontal scale of all the cross-sections is the same. The recovered Moho is plotted in all cross-sections as a black line. Different color scales are used in the crust and mantle, as shown at bottom.

**Figure 13.** (a) The average standard deviation of the ensemble of models from the Monte Carlo inversion is plotted versus depth as the solid line. The dashed line is the mean of the absolute value of the velocity anomalies at each depth taken across the entire study region. (b) The standard deviation of the dispersion curves predicted by the ensemble of models averaged across all geographic points is shown.

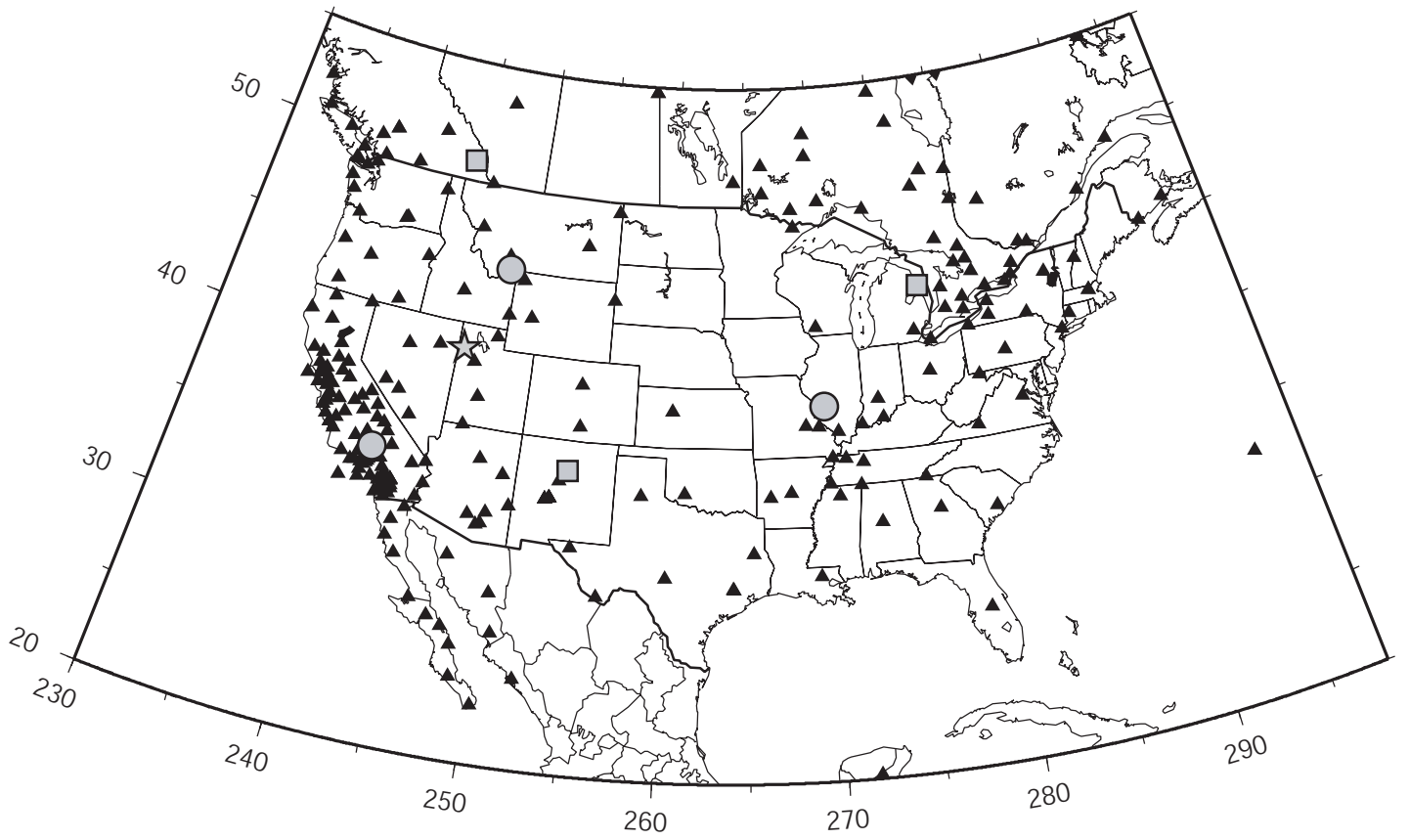
**Figure 14.** Horizontal slices showing the estimated standard deviation of the ensemble of  $V_s$  models derived from the Monte-Carlo inversion at the depths presented in Figure 10. Panels (c) and (d) are results at 4 km above and below the Moho, respectively.

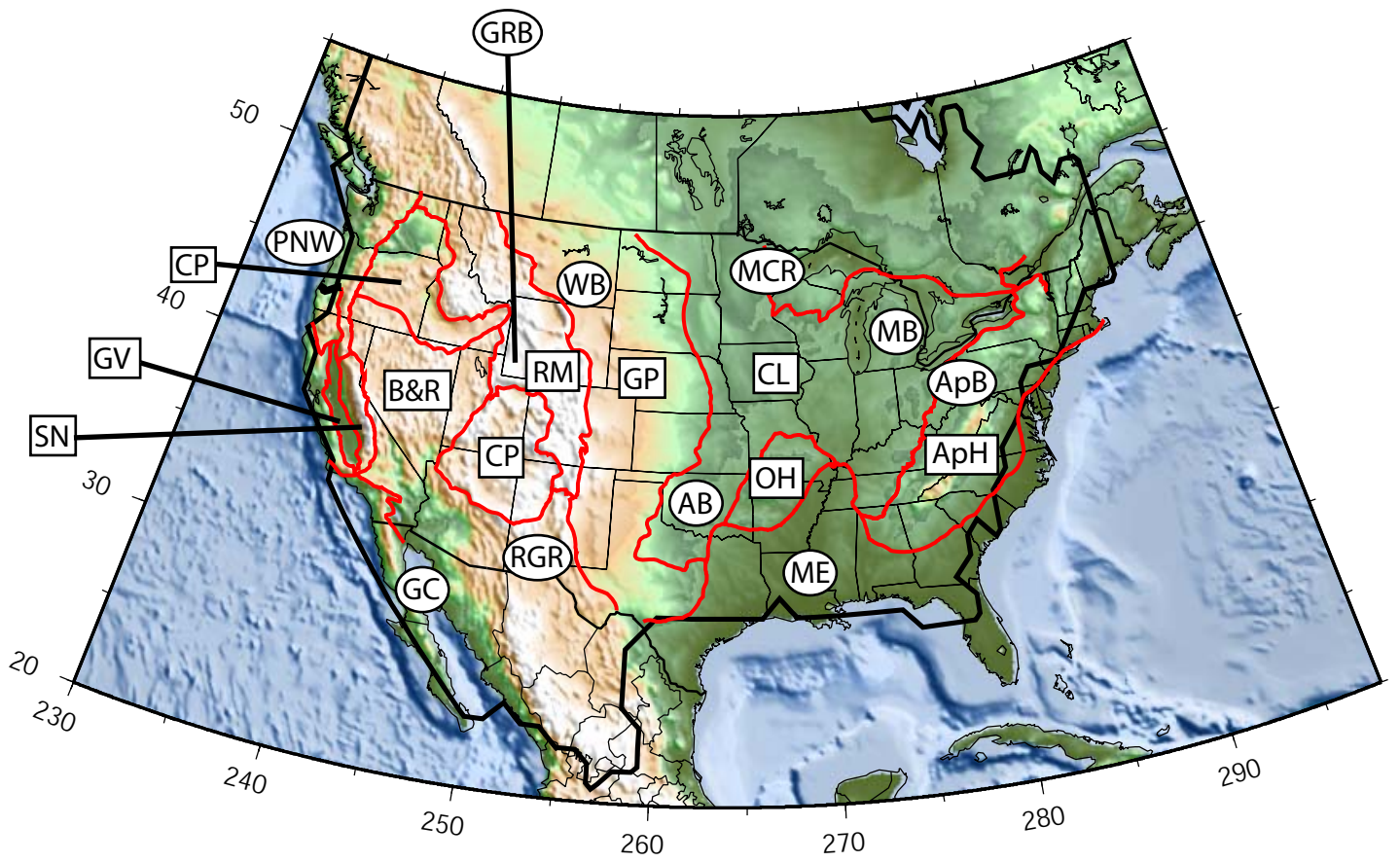
**Figure 15.** Histograms of velocity values taken from the  $0.5^\circ$  grid east and west of the approximate location of the boundary of the crustal dichotomy in the lower crust and at 80 km depth across the profile at  $40^\circ\text{N}$ . The values are from the “favored model” and the boundary is defined at 3.75 and 4.55 km/s in the lower crust and mantle, respectively.

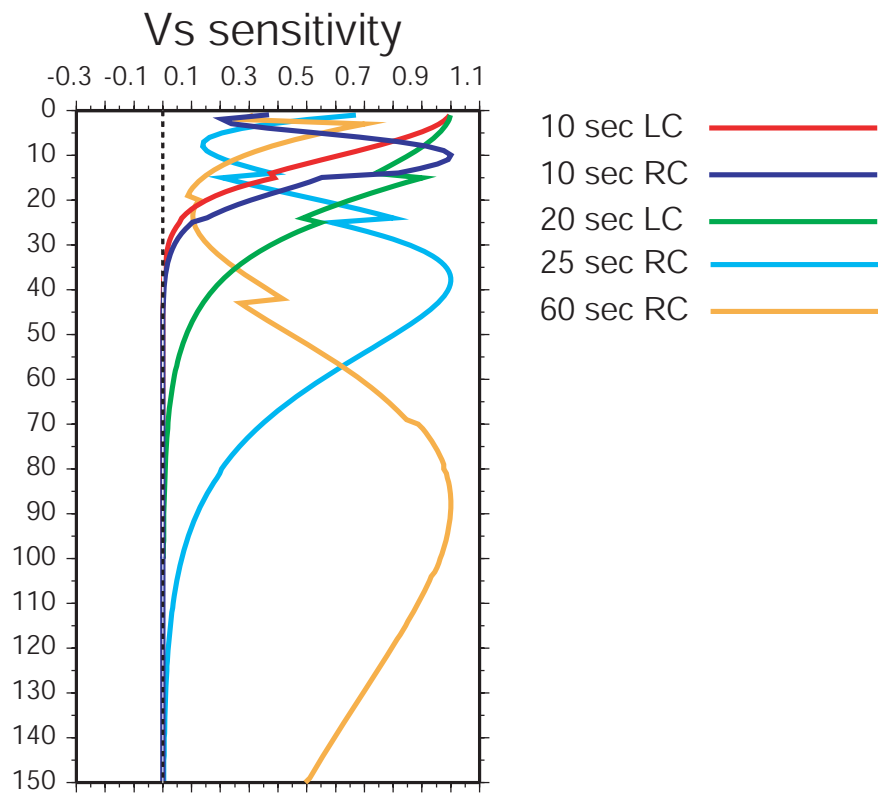
**Figure 16.** The location and uncertainty in the east-west shear velocity dichotomy for the lower crust (a) and the uppermost mantle (b). Contours of velocity are plotted for the 20th (grey) and 80th (black) percentile models at 3.75 km/s for the lower crust and 4.55 at 80 km in the mantle taken from the ensemble of accepted models determined by Monte Carlo inversion. The red contour marks the approximate location of the Rocky Mountain Front.

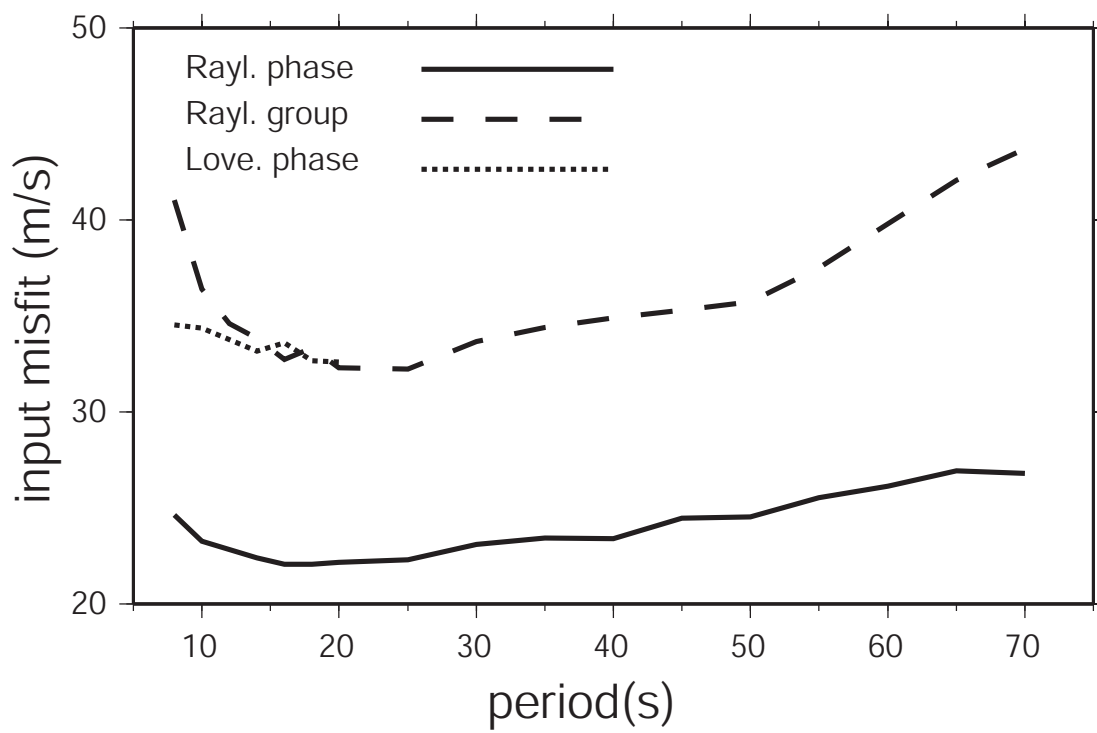
**Figure 17.** An example of the improvement in fit afforded by allowing radial anisotropy or breaking the monotonicity constraint (allowing a low velocity zone, LVZ) in the crust. The dispersion curves for the monotonic isotropic, radial anisotropic, and LVZ model are labeled in (a) and the corresponding models are shown in (b). Radial anisotropy is allowed only in the middle crust.

**Figure 18.** (a) The best fitting middle crustal radial anisotropy model for the US where, for example, a value of 5% signifies  $V_{sh}/V_{sv} = 1.05$ . (b) The minimally anisotropic model from the ensemble of acceptable models that emerge from the Monte-Carlo inversion.

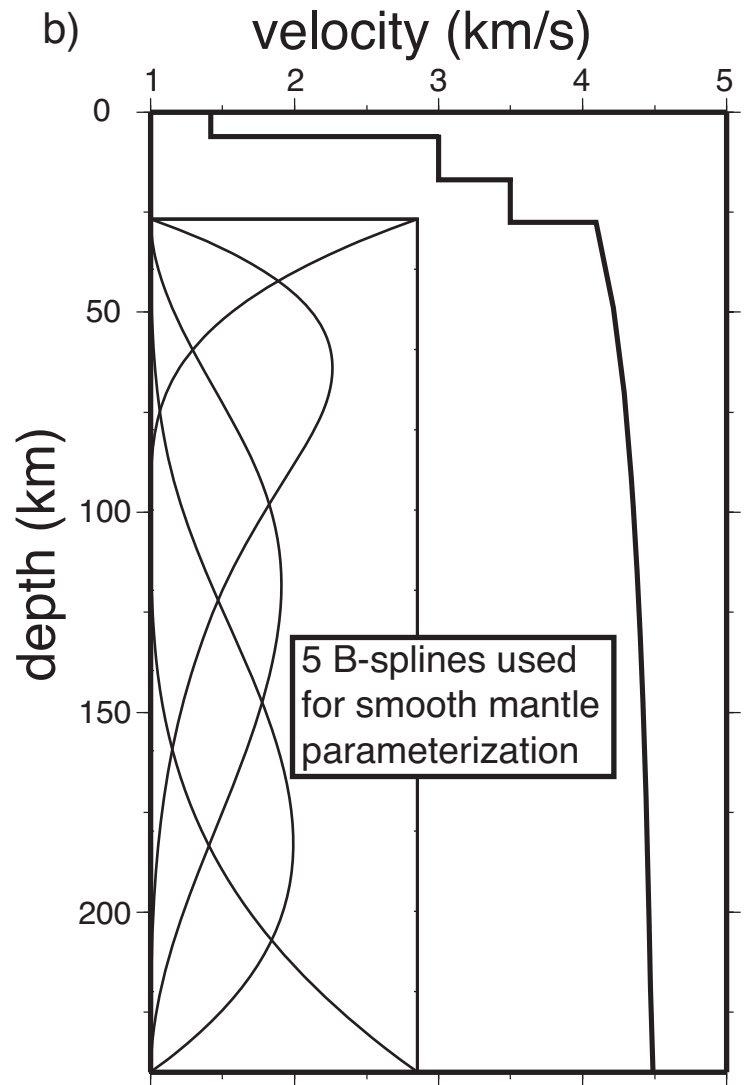
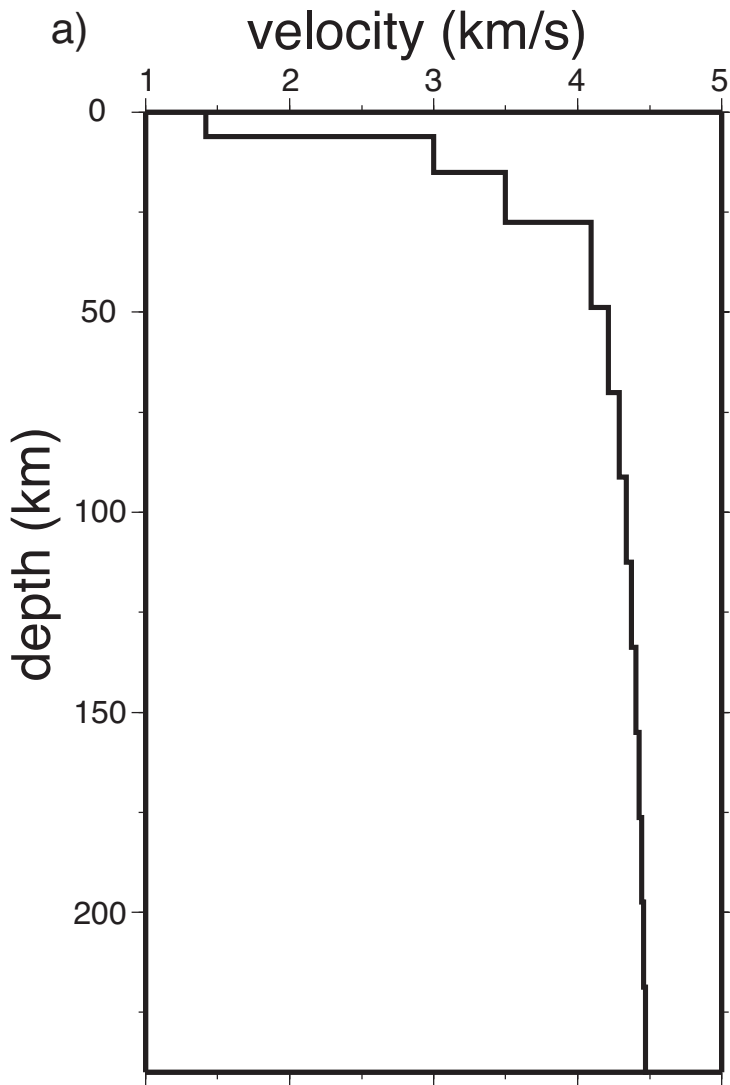


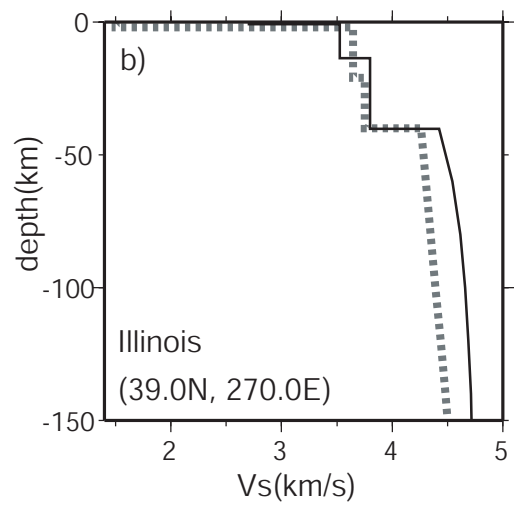
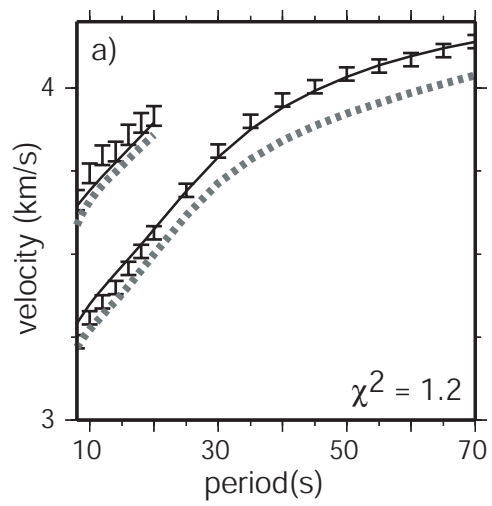


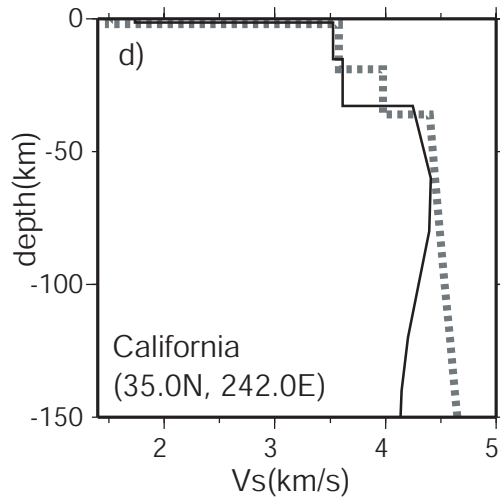
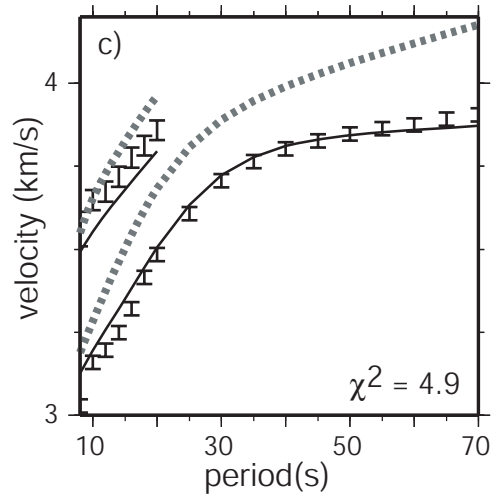
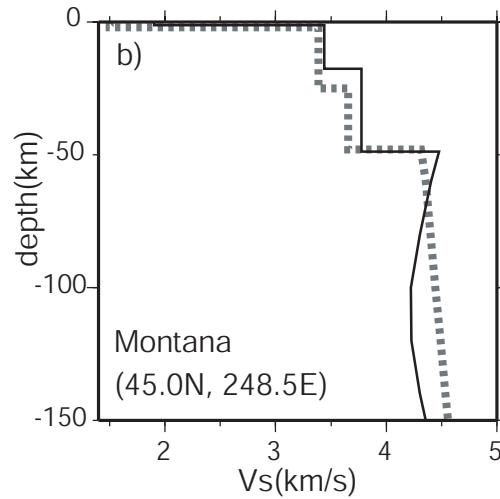
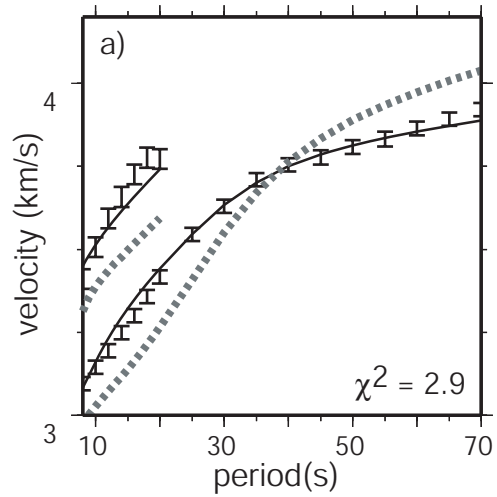


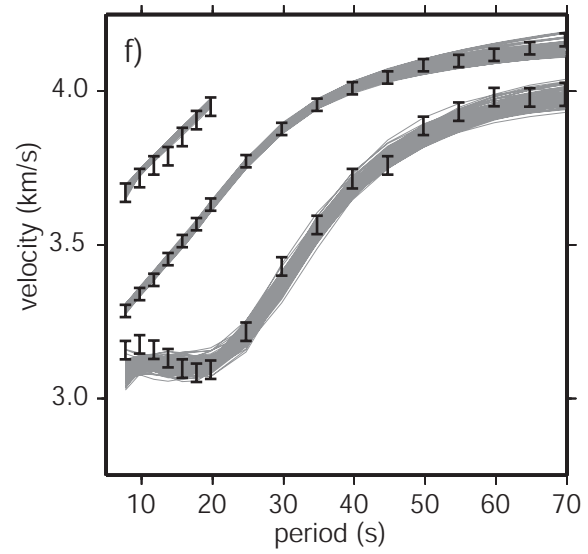
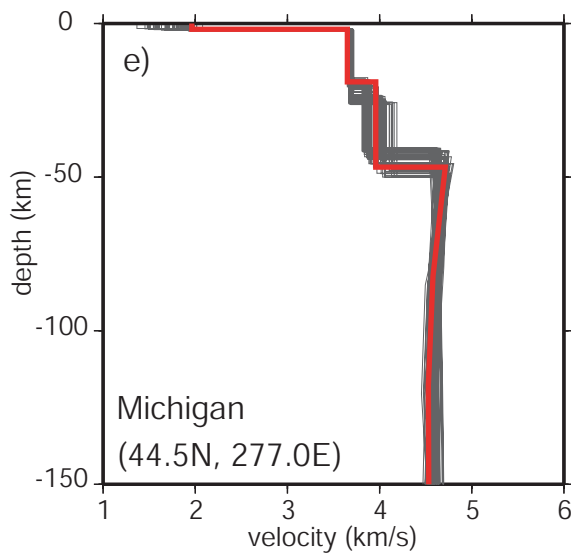
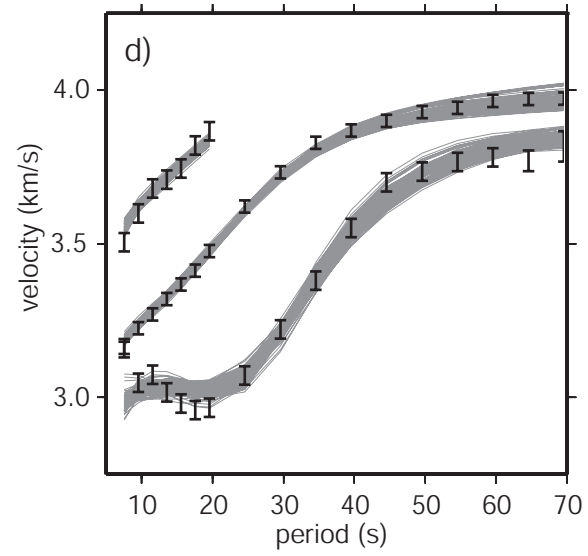
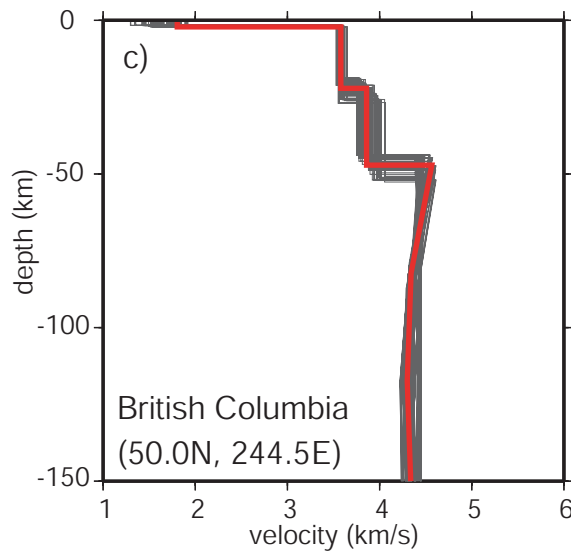
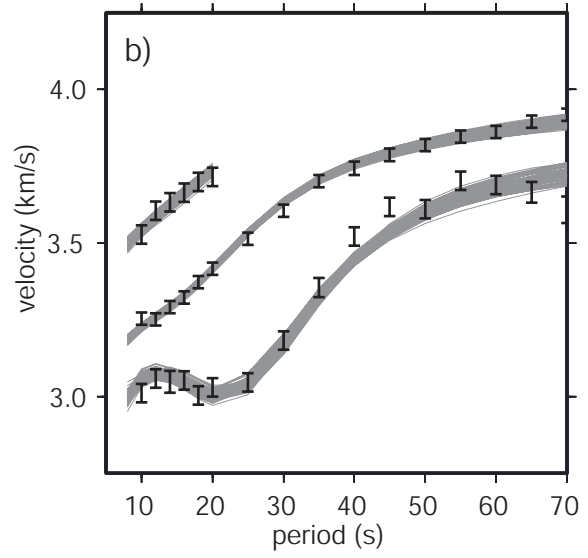
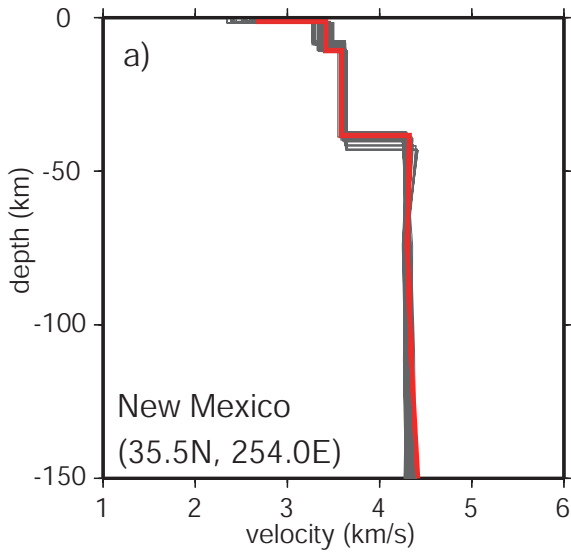




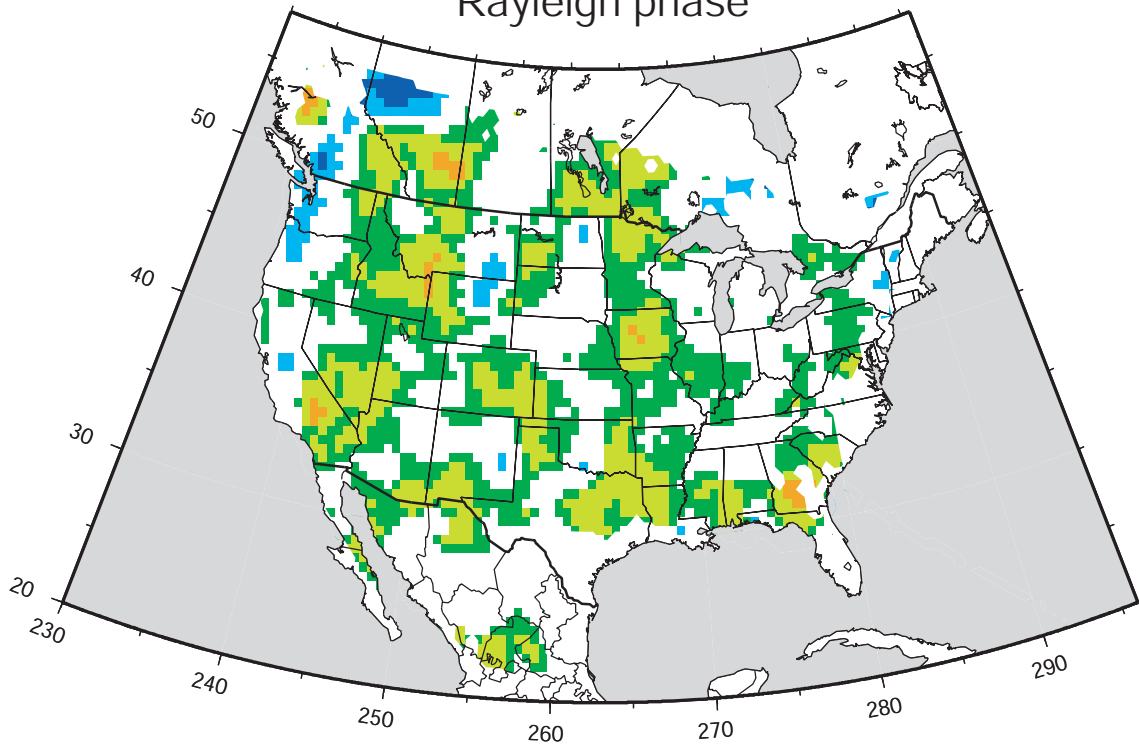








Rayleigh phase



Love phase

