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# High-Velocity Surface Layer Effects on Rayleigh Waves: Recommendations for Improved Shear-Wave Velocity Modeling

Gabriel Gribler Boise State University

Lee M. Liberty Boise State University

T. Dylan Mikesell Boise State University

1	High velocity surface layer effects on Rayleigh waves:
2	Recommendations for an improved shear wave velocity model
3	Gabriel Gribler <sup>1</sup> , Lee M. Liberty <sup>1</sup> , and T. Dylan Mikesell <sup>1</sup>
4	<sup>1</sup> Department of Geosciences, Boise State University, Boise, Idaho, USA
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#### Abstract

Soil stiffness estimates are critical to geologic hazard and risk assessment in urban centers. Multichannel analysis of surface wave (MASW) data collection along city streets is now a standard, costeffective, and non-invasive soil stiffness approximation tool. With this approach, shear wave velocities q (Vs) are derived from Rayleigh wave signals. While the current MASW practice is to neglect the effect 10 of a high-velocity road layer on soil Vs estimates, our models show measurable impacts on Rayleigh wave 11 amplitudes and phase velocities when seismic data are acquired on a road surface. Here, we compare 12 synthetic models to field MASW and downhole Vs measurements. Our modeling indicates that a road 13 layer attenuates Rayleigh wave signals across all frequencies, introduces coherent higher mode signals, 14 and leads to overestimated Vs and Vs30 values. We show that Vs30 can be overestimated by more 15 than 7% when soft soils underlie a rigid road surface. Inaccurate Vs estimates can lead to improper 16 soil classification and bias earthquake site response estimates. For road-based MASW data analysis, we 17 recommend incorporating a surface road layer in the Rayleigh wave inversion to improve Vs estimate 18 accuracy with depth. 19

### 20 Introduction

6

Site-based measurement of soil or rock properties is critical for the design of both new construction and for building retrofits. Shear wave velocities (Vs) directly relate to soil stiffness, and can be used to estimate local weak and strong earthquake ground motions (e.g., Aki, 1993; Kramer, 1996) and soil deformation potential (e.g., Andrus and Stokoe II, 2000; Kayen et al., 2013). Thus, a standard characterization practice is to estimate Vs to a depth of 10 to 30 m, either through direct downhole measurements (e.g., Robertson

et al., 1986) or through surface-based measurements. The downhole seismic measurement approach can be 26 expensive and provides only one-dimensional Vs estimates to borehole depths. The multi-channel analysis 27 of surface wave (MASW) active-source surface seismic approach (McMechan and Yedlin, 1981; Song et al., 28 1989; Park et al., 1999) has been shown to approximate Vs with depth in one dimension (Park et al., 1999; Stephenson et al., 2005) and is useful for non-invasive and rapid measurements. Traditionally, active source 30 surveys rely on geophone coupling to firm soil. While this approach is feasible in undeveloped areas, it is often 31 not practical in urban areas where native materials lie beneath a paved surface. To estimate soil stiffness 32 in developed landscapes, the seismic land streamer MASW approach, where the source and receiver(s) are 33 directly coupled to a road surface, is now a common acquisition method (e.g., Van Der Veen et al., 1999; 34 Van der Veen et al., 2001; Pugin et al., 2004; Liberty and Gribler, 2014). Here, we explore the impact of Vs 35 estimates in the presence of a paved surface layer, specifically to estimate high frequency site response and soil structure beneath urban centers. 37

Inazaki (2006) compared active source seismic data acquired on a road surface to data acquired on 38 native materials adjacent to the road. While they noted a relative decrease in surface wave amplitude while 39 acquiring data on the road, Inazaki (2006) did not quantify these amplitude effects. Foti et al. (2018) noted 40 greater coherence of higher mode surface waves at higher frequencies compared to fundamental mode signals 41 in the presence of a rigid surface layer, but they did not explore the phase velocity effects on the resulting soil 42 stiffness or Vs profile estimates. Through modeling and field data analysis, we explore the effect of a road 43 surface on Rayleigh wave amplitude, dispersion, and inversion-derived Vs estimates. We compare MASW 44 data collected through downtown Salt Lake City via a seismic weight drop/land streamer system (Liberty 45 et al., 2018) to Vs estimates derived from downhole seismic cone penetration test (SCPT) data (McDonald 46 and Ashland, 2008). These SCPT data are assumed to be ground truth, as they offer accurate Vs estimates, 47 with errors ranging from 1-3% in unconsolidated sediments (Moss, 2008). We generate synthetic Rayleigh 48 wave dispersion data from the SCPT data and then modify the upper 10 to 25 cm Vs to simulate different 49 road conditions. We forward model dispersion data to quantify phase velocity and amplitude changes as 50 a function of frequency for the different Vs models. We then make recommendations to best estimate Vs 51 with depth from MASW data acquired on a road surface. Here, we use the National Earthquake Hazard 52 Reduction Program (NEHRP) site classifications for both field and modeled data (BSSC, 2009). 53

This paper is organized in the following manner. We first introduce the two types of (common) road compositions that we consider in this study. We then introduce the modeled and field acquisition geometry, as well as soil properties. Using numerical modeling, we investigate the effects due to different road surfaces on the amplitude and phase velocity of Rayleigh waves and the eventual Vs models estimated by phase velocity inversion. Finally, we discuss the potential for incorrect site classification due to neglecting the road

<sup>59</sup> surface and we quantify the potential for phase velocity bias with frequency based on the natural material
<sup>60</sup> beneath the road surface. We specifically highlight the road effects on soft soils, where Vs overestimation is
<sup>61</sup> greatest.

### 62 Road properties

While a wide range of road designs and materials are used in construction, we focus our models on two road 63 designs that commonly comprise city streets (Yoder and Witczak, 1975). We define a rigid road as a layer 64 that contains a binder (commonly cement) within a crushed aggregate base. This design is topped with an 65 asphalt surface. A rigid road is commonly used for high-traffic city streets, as it offers many of the strength 66 and load carrying capabilities of concrete, but is simple and cost effective to construct. A *flexible* road 67 provides less support when compared to a rigid road, as it has no binder in the crushed aggregate base layer. 68 This layer is also topped with an asphalt surface, and is commonly used for low-traffic neighborhood roads 69 and parking lots. Physical property estimates, including Vs, of the two modeled road types are outlined in 70 Nazarian et al. (1988). Here we model 1) a rigid 25 cm thick road layer with Vs=1300 m/s and 2) a flexible 71 10 cm thick road layer with Vs=800 m/s. For both road types, we use a Poisson ratio of 0.40 and a density 72 of  $2100 \text{ kg/m}^3$ . While we model both road surface conditions, our field data were collected on a road surface 73 that best matches a rigid high-traffic city road (Liberty et al., 2018). 74

### 75 Seismic acquisition and model parameters

Our models simulate seismic data acquired upon soft and stiff soils of the Salt Lake basin (e.g. McDonald 76 and Ashland, 2008). Our modeled and field acquisition geometry consist of 48 vertically oriented geophones, 77 spaced 1.25 m apart, with an impulsive seismic source located 5 m from the nearest geophone. We base our 78 model and field comparisons on co-located SCPT and surface-seismic MASW measurements. The shallow 79 stratigraphy at the field site consists of alternating fine- and course-grained lacustrine and alluvium (Per-80 sonius and Scott, 1992), and the depth to groundwater is within the upper few meters. For comparison to 81 surface seismic measurements, we use borehole 146 (McDonald and Ashland, 2008), where SCPT derived 82 Vs ranges from 120 to 220 m/s in the upper 30 m, with relatively stiff NEHRP Class D soils (Vs>200 m/s) 83 between 2 to 5 m depth and 21 to 25 m depth (Figure 1). SCPT results suggest NEHRP Class E soils 84 (Vs < 180 m/s) at the surface, between 5 to 15 m depth, and 21 to 28 m depth. This site is classified as NEHRP site class E with an average Vs for the upper 30 meters (Vs30) of 170 m/s. 86

### <sup>87</sup> Numerical modeling

To explore road-surface effects on MASW-derived Vs estimates, we model Rayleigh wave propagation in a 88 seven-layer 1-D velocity model simplified from the SCPT measurements in borehole 146 (dashed-red line in 89 Figure 1). We use the generalized reflection and transmission (R/T) coefficient forward modeling method ٩n (Kennett, 1974; Kennett and Kerry, 1979; Pei et al., 2008, 2009; Kennett, 2009) to calculate theoretical 91 dispersion curves for fundamental and higher mode Rayleigh waves. We then model the waveform time 92 series, which contain only Rayleigh wave energy, using 1-D modeling code from Michaels and Smith (1997) 93 and Michaels (2018). We model waveforms for three cases: 1) native materials; 2) the same as (1) with an additional 25 cm thick rigid road surface; 3) the same as (1) with an additional 10 cm flexible road surface. 95 We first compare the seismic power spectra from the synthetic waveforms in each model (Figure 2). We then 96 compare dispersion derived from the synthetic model shot gathers to dispersion derived from a field record 97 located adjacent to borehole 146 (Figure 3). 98

#### <sup>99</sup> Spectral power analysis

To investigate the road-surface effect on seismic power spectra, we calculate the total signal power by applying 100 the Fast Fourier Transform to each seismic waveform and summing over all receivers. We show the 3–40 Hz 101 power spectra for the synthetic models, with and without the road layer, in Figure 2. Compared to the 102 native material model, we observe an approximate peak seismic power reduction of 2.5 orders of magnitude 103 when we include a rigid 25 cm thick road layer (Figure 2). In the model with a 10 cm thick flexible road 104 surface, we observe a similar power spectra reduction between 25–40 Hz as compared to the rigid road model. 105 However, at lower frequencies the power is only reduced by about one order of magnitude when compared 106 to the native material model. Although we do not show the amplitude effects from other velocity models, 107 our modeling indicates an increasing amplitude reduction with increasing velocity contrast between the road 108 material and the underlying strata. We attribute this reduction in seismic power to an increase in Young's 109 modulus of the surface layer, reducing the resulting stress exerted on the underlying native material from 110 the fixed force source. 111

In summary, both the rigid and flexible road surface models reduce power spectra amplitudes compared to the native material model, with the greatest power decrease observed in the rigid road surface model at frequencies below 25 Hz. This result is consistent with field observations (e.g. Inazaki, 2006), and we conclude that the Rayleigh wave amplitude is strongly related to shallow subsurface properties and the velocity contrast with the underlying strata. To compensate for amplitude attenuation, a larger seismic source may be needed on rigid or flexible road surfaces when compared to data collection on native materials.

However, the amplitude of other coherent signals may also change with different shallow velocity models, so noise amplitudes may also vary. Field testing should confirm whether adequate coherent surface wave signals are consistently generated while acquiring data on a road surface.

#### <sup>121</sup> Phase velocity analysis

In addition to the influence of the road surface on amplitude, we investigate the road surface influence 122 on phase velocity dispersion of the Rayleigh wave. Using the seven-layer SCPT-derived native material 123 model (Figure 1), we generate 3–40 Hz theoretical dispersion images and dispersion curves (Figure 3a). We 124 then generate dispersion images and dispersion curves for two eight-layer models; the seven-layer model 125 SCPT-derived model plus the additional rigid or flexible road layer (Figure 3b and c, respectively). While 126 we focus on fundamental mode energy in our analysis, we recognize that higher mode signals impact the 127 fundamental mode coherence and dispersion curve picks. Therefore, for our waveform modeling, we include 128 the first and second higher modes with the MASW-derived dispersion images and we plot these theoretical 129 dispersion curves. We explore both phase velocities with frequency, and changing patterns of coherence in 130 the dispersion images. We scale each image by the peak amplitude in each image. 131

Our first observation is that the theoretical fundamental mode dispersion curves differ across each model. 132 while the higher mode curves remain mostly unchanged (Figure 3). This suggests that the thin road surface 133 minimally impacts higher mode phase velocities for our models. However, although the higher mode phase 134 velocities are consistent between models, we observe that varying higher mode coherence influences the 135 fundamental mode coherence through interference. This interference is best observed on the rigid road 136 model, where the peak coherence for frequencies greater than 20 Hz is associated with higher mode signals 137 (Figure 3b). For the native material and flexible road surface models, we observe little interference between 138 the fundamental and higher modes, and the majority of the energy coherence in both images tracks the 139 fundamental model dispersion curves (Figure 3a and c, respectively). We relate this difference to lower 140 coherence higher modes compared to the fundamental mode amplitudes for the native and flexible road 141 models and conclude that the high-frequency fundamental mode coherence is compromised in the rigid road 142 layer due to higher mode interference. 143

To compare the three modeled dispersion curves to each other and to our MASW field data, we plot the calculated theoretical dispersion curves for the fundamental mode on a dispersion image obtained from a shot gather that lies adjacent to SCPT borehole 146 (Figure 3d). (Recall that the theoretical curves come from the smoothed SCPT model at borehole 146.) The MASW data were acquired on a high traffic road surface and should best match the results from rigid road surface. The three curves show that the road surface

models, with 25 cm rigid and 10 cm flexible layers, exhibit an increase in phase velocity across all frequencies 149 when compared the native material model. At frequencies between 5 and 10 Hz, we observe a  $\sim 10\%$  phase 150 velocity increase due to the presence of either road surface. This difference decreases slightly between 10 151 and 20 Hz. At frequencies above  $\sim 20$  Hz, the fundamental mode phase velocity diverges for each model, and 152 above 30 Hz, the phase velocity differences among the models level off, with the rigid and flexible models 153 showing velocity increases of  $\sim 30\%$  and  $\sim 12\%$ , respectively, compared to the native material model. Note 154 that the theoretical dispersion from the rigid road surface model displays very little velocity variability at 155 frequencies above 20 Hz. This suggests that if a user inverted dispersion picks above  $\sim 25$  Hz in the presence 156 of a road surface layer, the likely result would be a smooth and overestimated Vs depth profile at shallow 157 depths, compared to data acquired on native material. We note that coherence of the fundamental mode 158 signal on the field record diminishes above  $\sim 22$  Hz (Figure 3d). This coherence pattern is most similar to the 159 25 cm rigid road surface model where higher modes dominate the dispersion image (Figure 3b). Therefore, 160 above this frequency, mode mis-identification is possible and the shallow velocity structure is masked by road 161 layer effects. Below  $\sim 22$  Hz we observe the highest coherence trend tracking more closely to the models with 162 the road surface included (Figure 3d). Therefore, we conclude that Vs differences in the top 10 or 25 cm 163 impact phase velocity picks at all frequencies, resulting in an overestimation of the Vs model for depths well 164 below the road layer. 165

#### <sup>166</sup> Estimation of Vs with depth

The purpose of this section is to determine the role that neglecting the road surface plays in the final Vs 167 depth profile. In this section, we invert the theoretical fundamental mode dispersion curves for the native 168 material and rigid road surface models using the genetic algorithm (GA). We use the GA because it offers 169 a robust approach to estimate Vs with depth (Yamanaka and Ishida, 1996; Dal Moro et al., 2007). For the 170 GA inversion, we set upper and lower bounds on all parameters being estimated and five generations of 171 500 individuals are run (total of 2500 trial models). We compare all of the resulting inverse models to the 172 seven-layer SCPT-derived model from borehole 146, which has Vs30=170 m/s and is classified as a class E 173 soft soil site (BSSC, 2009). The forward model that we use in the genetic algorithm is the same R/T method 174 used to generate the theoretical dispersion curves (Figure 3). For the dispersion inversion, we allow both 175 depth and Vs to vary for each layer. For computational savings and inversion consistency, we set the bounds 176 on velocity to vary by plus or minus 20% and layer thicknesses to vary by plus or minus 10%, compared to 177 the true model. The data used in the inversion are the theoretical dispersion curves outlined in Figure 3 from 178 4-35 Hz. This limited pass band reduces the potential for shallow guided wave effects producing instabilities 179

in our forward model. We note that while we are using noise-free theoretical dispersion curves, the inversion problem itself is highly non-unique. Therefore, we do not expect to perfectly recover the Vs depth model. We do, however, expect to observe similar general trends in the profiles (i.e. gross structure like Vs10 or Vs30). To quantify the error associated with the inversion, we run 12 iterations for each GA inversion, and calculate the mean and one standard deviation of the resulting velocity profiles (Figure 3a-c).

We begin by comparing the SCPT-derived Vs model to the inversion result derived from dispersion 185 curve picks obtained using the same model (Figure 3a). The resulting Vs velocity inversion with the GA 186 is very similar to the true model, with the non-unique inversion producing slight velocity differences as 187 depth increases (Figure 4a). Due to limited frequency information, we observe an increasing misfit between 188 observed and calculated Vs models at depths greater than 15 meters. This is specifically related to low 189 frequencies, and we observe this increase in uncertainty in both the field and synthetic data inversions (e.g., 190 Foti et al., 2018; Liberty et al., 2018). To compare bulk Vs values, we plot the average Vs with depth (e.g. 191 the values corresponding to 30 meters depth represent Vs30) (Figure 4e). This plot shows that the average 192 Vs through the entire depth range agrees with the true model. The inverted model yields a Vs30 of 172 m/s, 193 a 1.1% overestimation compared to the smoothed SCPT model. While slightly different, this model still falls 194 within a classification of site class E soil. 195

We next examine the effects introduced by collecting data on a road surface, but neglecting the road 196 surface in the inversion (Figure 4b). When we compare this seven-layer Vs model to the SCPT-derived 197 model, we note an overestimation of Vs, especially in the upper 15 meters. The average Vs poorly matches 198 the SCPT model in the upper 15 meters where a velocity inversion is mapped. Moreover, the average Vs 199 is overestimated at all depths (Figure 4e), and we observe upwards of 20% overestimation of average Vs in 200 the upper two meters, while Vs for depths below 5 m is overestimated by  $\sim 10\%$  (Figure 4f). We observe 201 a Vs30 difference of 7.6% compared to the SCPT model and a Vs30 of 183 m/s, or site class D stiff soils 202 (Figure 4e). Thus, neglecting the high velocity road layer in the inversion has changed the site classification 203 for this model. 204

To accommodate a road layer in our inversion, we fix our first layer to the thickness and velocity of the 205 road material and we retain seven-layers in our soil model (Figure 4c). With this road-surface constraint, we 206 observe a better match to the SCPT-derived Vs model, now very similar to the inversion of native materials 207 (Figure 4d). Average Vs differences in the upper few meters are reduced to  $\sim 3\%$  when compared to the 208 native model (Figure 4f) and the Vs30 difference is now 1.7% (Figure 4e). By including the rigid road layer 209 in the inversion, Vs30 now measures 173 m/s, and the site class E soil type closely matches the SCPT-derived 210 Vs values. This suggests that by including the road surface in our inversion, we better match absolute and 211 depth-averaged Vs values. We note that this requires accurate a priori knowledge of the road thickness 212

and road Vs to obtain an accurate Vs profile with the MASW inversion. Without direct road thickness measurements from engineering logs, destructive coring methods, or through methods like high frequency dispersion analyses (Nazarian et al., 1988), we suggest examining a range of forward models, that include differing road layer properties to recognize the effects of, and compensate for, the road layer effect. We do not treat this model examination further though.

#### <sup>218</sup> Road surface influence on field data

We acquired shot records with a seismic land streamer and accelerated weight drop source along a five km 219 long profile of 700 South road in Salt Lake City, Utah, USA (Liberty et al., 2018). Because Vs generally 220 increases from west to east across the profile (McDonald and Ashland, 2008), we can compare Vs derived 221 from field dispersion images, modeled with and without a road layer. We select shot gathers from a 1.6 km 222 long portion of 700 South, where we observe a transition from site class E soils to site Class D soils (Figure 5). 223 We generated five to seven dispersion images from shots within 10 m of a mid-point location at ten locations 224 spaced about 200 m apart. For our inversion, we first neglect the rigid road surface layer and invert using a 225 seven-layer Vs model to 30 m depth. We then add a rigid road layer between the surface and 25 cm depth 226 with a Vs=1300 m/s, then re-invert the data. We plot error bars representing one standard deviation from 227 the sets of neighboring inversions, showing the variability between adjacent data and the inversion results. 228

Consistent with our numerical modeling results, we observe that by not including the road layer in the 229 inversion, we overestimate Vs30 values by 5 to 10% (Figure 5). We observe an overestimate of about 10% for 230 the shots where site Class E soils are mapped, and we observe an overestimate of about 5% for the shots where 231 site Class D soils are mapped. More importantly, the Vs30 values that include a road layer match boreholes 232 146 and 147 SCPT-derived Vs30 measurements that were obtained along the streamer profile (McDonald 233 and Ashland, 2008). Our analysis demonstrates that if the road surface is not accounted for, the transition 234 from E class sediments (Vs30 < 180m/s) to D class sediments (180 m/s < Vs30 < 360m/s) shifts to the 235 east by one city blocks (about 200 m distance). Considering that this transition is located within a rapidly 236 expanding urban corridor, the road corrections are critical to include in the MASW inversion analysis. If a 237 road layer is not included in Vs inversions within urban corridors, the uncertainty in site class boundaries 238 should be increased due to Vs30 values being biased to higher values. 230

#### <sup>240</sup> Site Classification Bias

While the inversion approach illustrates the limitations of using MASW to characterize Vs structure with depth, we obtain more accurate depth averaged Vs30 estimates in the numerical models by accounting for a

rigid road surface in the inversion. Now, we explore a range of Vs starting models (from class E to class B) to determine how a rigid surface layer influences Vs30 site classification. We compare theoretical dispersion curves for models with and without a 25 cm thick rigid road layer and calculate the percent difference in phase velocity for frequencies from 3 to 80 Hz (Figure 6). We use eight models, each with a Vs gradient of +4 m/s per meter. Each model contains a different Vs at the surface (V<sub>o</sub>) and extends to 100 m depth. We model to 100 m depth to address the high velocity models at low frequencies, but we mostly focus on the different Vs30 site class or subclass as defined by Wills et al. (2000).

Our models show that the phase velocity is faster across the entire frequency band when the 25-cm thick 250 rigid road layer is included in the model. The greatest phase velocity difference in the rigid road models 251 occurs at the slowest Vs values (Figure 6). For example, we observe a 20% phase velocity overestimation 252 from 20 to 80 Hz when the original Vs30 is 155 m/s (site class E). For starting models with higher average 253 Vs30 (e.g. classes B and C), the road surface influence is less pronounced. Once Vs30 is greater than 760 m/s 254 (site class B), phase velocities differ by less than 3%. These results indicate that for lower velocity materials 255 (i.e. site class E and D), the road layer is significant and should be included in the inversion to accurately 256 estimate Vs30 or other average Vs values. Conversely, for faster velocity materials (NEHRP class C and B), 257 the dispersion curve is less sensitive to a fast surface layer. 258

To compare the influence of the 10 cm thick flexible road surface (Vs=800 m/s), we repeat this analysis, but only for the two end member models of site class E and B (Figure 6). For low Vs (site class E), we again observe a frequency dependent phase velocity difference, with nearly a 20% difference at frequencies above 60 Hz and an 8% difference near 10 Hz. For high Vs material (site class B), we observe no discernible difference in the phase velocities. We conclude that for thin flexible road surfaces, the road surface effect is negligible when the underlying native material has high Vs, but is significant and important to consider when low Vs values underlie any road surface (flexible or rigid).

### 266 Conclusion

In urban environments, we are often limited to collecting active-source seismic data on paved roads. Paved roads act as a thin high Vs surface layer, but this layer is typically neglected when estimating the underlying Vs by inverting Rayleigh wave dispersion curve data. Here we demonstrate that a thin road surface can have a large effect on surface wave amplitude, velocity dispersion, and higher mode coherence. Modeling seismic data on a thin road layer, we observe a two and a half order of magnitude decrease in seismic power. This suggests that in a low signal-to-noise environment, a larger seismic source or additional data stacking (e.g. more hammer hits) may be needed with the MASW approach to attain acceptable data

quality. More importantly, the Rayleigh wave phase velocities are faster across all modeled frequencies when 274 imaging through a road, leading to an overestimation in Vs with depth. To address this velocity difference, 275 we recommend the addition of a high-velocity surface layer in the Rayleigh wave inversion. In our example, 276 we find that the Vs30 is overestimated by approximately 10% because of a 25 cm thick high-velocity road 277 layer. In addition, the road surface can produce higher mode dispersion interference that may result in 278 incorrect fundamental mode picks at high frequencies. By modeling the dispersion effect for linear velocity 279 gradient models with different average Vs, we find that for Vs30 < 360 m/s, the road surface has a significant 280 effect on the dispersion data and should be accounted for by incorporating a high-velocity road layer into the 281 inverse model. For high-velocity sediments (Vs30 > 760 m/s), the road surface has a less significant effect, 282 but incorporating a high-velocity road layer in the inverse model can still improve the Vs model. 283

### <sup>284</sup> Data and Resources

- R/T dispersion forward modeling code https://github.com/yiran06/mat\_disperse (last accessed August 2019)
- Waveforms modeling using "Basic Seismic Utilities Software" https://scholarworks.boisestate. edu/geo\_data/3/ (last accessed August 2019)
- SCPT data provide by Utah Geological Survey https://geology.utah.gov/about-us/geologic-programs/
- 200 geologic-hazards-program/for-consultants-and-design-professionals/community-velocity-model-cvm-and
- shallow-shear-wave-velocity-data/ (last accessed August 2019)
- Seismic data uploaded as IRIS assembled data set 19-019: Salt Lake 2017 Land Streamer profiling
   DOI: 10.7914/SN/XQ\_2017 https://doi.org/10.7914/SN/XQ\_2017 (last accessed August 2019)

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## 367 Author Address

- 368 Gabriel Gribler
- <sup>369</sup> 1910 University Drive
- Boise, Idaho 83725
- 371 USA
- <sup>372</sup> Lee M. Liberty
- <sup>373</sup> 1910 University Drive
- <sup>374</sup> Boise, Idaho 83725
- 375 USA
- 376 T. Dylan Mikesell
- <sup>377</sup> 1910 University Drive
- <sup>378</sup> Boise, Idaho 83725
- 379 USA

# <sup>380</sup> Figure Captions

381	1	Vs estimates from SCPT measurements at borehole 146 (black) and surface-seismic MASW;	
382		the solid blue line represents the average Vs profile from the six one dimensional MASW	
383		linearized inversions near borehole 146. The surface seismic data are located within about	
384		10 m of borehole 146. Dotted blue lines are the $\pm$ one standard deviation. Note that the	
385		smoothed linearized inversion does not accurately reconstruct velocity inversions. Dashed	
386		(red) line represents the simplified seven-layer SCPT-derived model used in the forward models	
387		(Figures 2–4)	17
388	2	Summed power spectra plot between 3 to 40 Hz for modeled data with a native material road	
389		surface (solid black), a 25 cm rigid (1300 m/s) road surface (dot-dashed red), and a 10 cm $$	
390		flexible (800 m/s) road surface (dashed cyan)	18
391	3	MASW-derived dispersion images for the seven-layer velocity model (dashed-red line in Fig-	
392		ure 1) with different road surface properties: (a) native material seven layer model, (b) an	
393		added 25 cm thick stiff road surface layer (Vs= $1300 \text{ m/s}$ ), (c) an added 10 cm thick flexible	
394		road surface (Vs= $800 \text{ m/s}$ ). Lighter colors represent higher coherence and lines represent	
395		theoretical dispersion curves, which include the fundamental mode and the first two higher	
396		mode Rayleigh waves (dashed-black lines). (d) Dispersion image from seismic data collected	
397		at the location of SCPT borehole 146. The fundamental mode dispersion curves from (a), (b),	
398		and (c) are overlain on (d). Black asterisks indicate automatic dispersion picks tracking the	
399		highest coherence for each frequency. The eight-layer road surface dispersion model curves	
400		show higher phase velocities across all frequencies when compared to the seven-layer native	
401		material model curve.	19
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403		lines in a-c is the true model, solid colored lines are the average results from 12 individual	
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407		Vs at 30 meters depth corresponds to Vs30. (f) Percent difference in average Vs with depth	
408		between the seven-layer starting model and inverse models. Line styles and colors in (e) and	
409		(f) correspond to those in (d). The table summarizes the differences between inversion results	
410		and SCPT model.	20

411	5	(a) Topographic map showing seismic profile location (solid red line) along 700 South in Salt
412		Lake City, Utah, USA. Black diamonds indicate SCPT measurement locations for boreholes
413		146 and 147. Black dots near borehole 146 represent the geophone array length for our
414		(Figure 3d) comparison. (b) Vs30 estimates from shot gathers along the 700 South seismic
415		profile (a). Dotted (blue) line with circular markers represents inversion results with no road
416		layer included in the inversion. Error bars represent one standard deviation calculated from 5-
417		7 adjacent dispersion curves. Dotted (red) line with triangular markers represent GA inversion
418		results with a rigid road surface included in the inversion. Two diamonds represent SCPT-
419		derived Vs30 at boreholes 146 and 147. We observe a systematic overestimation of Vs30 when
420		when the layer is ignored compared to the inversion where the road layer is included $21$
421	6	Dispersion curve percent overestimation due to 25 cm thick rigid (1300 m/s) road surface for
422		different Vs30 measurements (NEHRP classes B–E). Velocity profiles constitute a constant
423		Vs gradient of $+4 \text{ m/s}$ per meter for the upper 100 meters with different surface velocities
424		$\mathbf{V}_o$ . Two dot-dashed (red) lines denote dispersion differences due to a 10 cm thick flexible
425		(800  m/s) road surface layer. Table shows the NEHRP classification of Wills et al. (2000) and
426		velocity parameters for each model



Figure 1: Vs estimates from SCPT measurements at borehole 146 (black) and surface-seismic MASW; the solid blue line represents the average Vs profile from the six one dimensional MASW linearized inversions near borehole 146. The surface seismic data are located within about 10 m of borehole 146. Dotted blue lines are the  $\pm$  one standard deviation. Note that the smoothed linearized inversion does not accurately reconstruct velocity inversions. Dashed (red) line represents the simplified seven-layer SCPT-derived model used in the forward models (Figures 2–4).



Figure 2: Summed power spectra plot between 3 to 40 Hz for modeled data with a native material road surface (solid black), a 25 cm rigid (1300 m/s) road surface (dot-dashed red), and a 10 cm flexible (800 m/s) road surface (dashed cyan).



Figure 3: MASW-derived dispersion images for the seven-layer velocity model (dashed-red line in Figure 1) with different road surface properties: (a) native material seven layer model, (b) an added 25 cm thick stiff road surface layer (Vs=1300 m/s), (c) an added 10 cm thick flexible road surface (Vs=800 m/s). Lighter colors represent higher coherence and lines represent theoretical dispersion curves, which include the fundamental mode and the first two higher mode Rayleigh waves (dashed-black lines). (d) Dispersion image from seismic data collected at the location of SCPT borehole 146. The fundamental mode dispersion curves from (a), (b), and (c) are overlain on (d). Black asterisks indicate automatic dispersion picks tracking the highest coherence for each frequency. The eight-layer road surface dispersion model curves show higher phase velocities across all frequencies when compared to the seven-layer native material mode curve.



Figure 4: Inversion results for native material (a), no road included (b) and road included (c). Black solid lines in a-c is the true model, solid colored lines are the average results from 12 individual inversions and dash colored lines represent one standard deviation of the inversion results. (d) Comparison between SCPTderived and inverted Vs profiles using a genetic algorithm inversion of the synthetic dispersion curves. (e) Depth versus time-averaged Vs. For instance, Vs at 30 meters depth corresponds to Vs30. (f) Percent difference in average Vs with depth between the seven-layer starting model and inverse models. Line styles and colors in (e) and (f) correspond to those in (d). The table summarizes the differences between inversion results and SCPT model.



Figure 5: (a) Topographic map showing seismic profile location (solid red line) along 700 South in Salt Lake City, Utah, USA. Black diamonds indicate SCPT measurement locations for boreholes 146 and 147. Black dots near borehole 146 represent the geophone array length for our (Figure 3d) comparison. (b) Vs30 estimates from shot gathers along the 700 South seismic profile (a). Dotted (blue) line with circular markers represents inversion results with no road layer included in the inversion. Error bars represent one standard deviation calculated from 5-7 adjacent dispersion curves. Dotted (red) line with triangular markers represent GA inversion results with a rigid road surface included in the inversion. Two diamonds represent SCPT-derived Vs30 at boreholes 146 and 147. We observe a systematic overestimation of Vs30 when when the layer is ignored compared to the inversion where the road layer is included.



Figure 6: Dispersion curve percent overestimation due to 25 cm thick rigid (1300 m/s) road surface for different Vs30 measurements (NEHRP classes B–E). Velocity profiles constitute a constant Vs gradient of +4 m/s per meter for the upper 100 meters with different surface velocities  $V_o$ . Two dot-dashed (red) lines denote dispersion differences due to a 10 cm thick flexible (800 m/s) road surface layer. Table shows the NEHRP classification of Wills et al. (2000) and velocity parameters for each model.