A California Statewide Three-Dimensional Seismic Velocity Model from Both Absolute and Differential Times

by Guoqing Lin,* Clifford H. Thurber, Haijiang Zhang,† Egill Hauksson, Peter M. Shearer, Felix Waldhauser, Thomas M. Brocher, and Jeanne Hardebeck

Abstract We obtain a seismic velocity model of the California crust and uppermost mantle using a regional-scale double-difference tomography algorithm. We begin by using absolute arrival-time picks to solve for a coarse three-dimensional (3D) $P$ velocity ($V_P$) model with a uniform 30 km horizontal node spacing, which we then use as the starting model for a finer-scale inversion using double-difference tomography applied to absolute and differential pick times. For computational reasons, we split the state into 5 subregions with a grid spacing of 10 to 20 km and assemble our final statewide $V_P$ model by stitching together these local models. We also solve for a statewide $S$-wave model using $S$ picks from both the Southern California Seismic Network and USArray, assuming a starting model based on the $V_P$ results and a $V_P/V_S$ ratio of 1.732. Our new model has improved areal coverage compared with previous models, extending 570 km in the SW–NE direction and 1320 km in the NW–SE direction. It also extends to greater depth due to the inclusion of substantial data at large epicentral distances. Our $V_P$ model generally agrees with previous separate regional models for northern and southern California, but we also observe some new features, such as high-velocity anomalies at shallow depths in the Klamath Mountains and Mount Shasta area, somewhat slow velocities in the northern Coast Ranges, and slow anomalies beneath the Sierra Nevada at midcrustal and greater depths. This model can be applied to a variety of regional-scale studies in California, such as developing a unified statewide earthquake location catalog and performing regional waveform modeling.

Online Material: Smoothing and damping trade-off analysis, a priori Moho depth, resolution tests, and map-view slices and cross sections through the 3D $V_P$ and $V_S$ models.

Introduction

Numerous studies of velocity structure in California have been done with varying scales in different areas (Table 1). The largest and most complete crustal tomography models in California are the recent southern California model by Lin et al. (2007) and the northern California model by Thurber et al. (2009). These studies have revealed key features of the crustal structure of California. The integration of the Northern and Southern California Seismic Networks (NCSN and SCSN) into a unified statewide network, the California Integrated Seismic Network (CISN; e.g., Hellweg et al., 2007), has motivated the development of a statewide seismic velocity model for locating earthquakes in California. In this study, we take advantage of the regional-scale double-difference (DD) tomography algorithm (Zhang and Thurber, 2006) to develop $P$- and $S$-wave velocity models for the entire state of California. Our $P$ velocity model is derived from both first-arrival absolute and differential time picks obtained from the California seismic networks and has a horizontal grid spacing as fine as 10 km. The $S$ velocity model is solved using the SCSN and USArray picks with a horizontal grid spacing of 30 km. There are numerous tomographic studies in California and it is impractical to compare our model with all of them; we focus on a comparison with the recent regional-scale tomographic studies for southern and northern California by Lin et al. (2007) and Thurber et al. (2009). Because our $V_P$ and $V_S$ models are solved using different sets of data and the $V_P$ model has better resolution, we present

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them separately with more emphasis on the $V_p$ model. Figure 1 shows selected geological and geographic features in our study and the positions of three profiles for the velocity cross sections.

**P-Wave Velocity Model**

**Data and Inversion Method**

The data sets for our $V_p$ model are the first-arrival absolute and differential times of 8720 earthquakes recorded by the seismic networks in California, consisting of 4325 events from the Northern California Seismic Network, 3668 events from the Southern California Seismic Network, and 727 events from the Pacific Gas and Electric seismic network (blue, pink, and green dots in Fig. 2a, respectively). These earthquakes were selected based on having the greatest number of $P$ picks among those events within a 6 km radius, with a magnitude threshold of 2.5. The total number of $P$ picks in our data set is 551,318 with an average of 63 picks per event. In order to improve constraints on the shallow crustal structure, we assembled first-arrival times from 3110 explosions and airguns (red circles in Fig. 2b) recorded on profile receivers and network stations. The principal active-source data sets and sources are listed in Table 2.
Quarry blasts, which have known locations but unknown origin times, are also valuable to be included in tomographic inversions because they provide constraints that are almost as good as the active-source data. We include data from 44 quarry blasts (blue circles in Fig. 2b), with 19 in southern California (see Lin et al., 2007) and 25 in northern California. Figure 2c shows the locations of temporary and network stations used in our study.

The model presented in this study is obtained by using a regional-scale DD tomography algorithm (tomoFDD; Zhang and Thurber, 2006), which is a generalization of DD location (Waldhauser and Ellsworth, 2000). It maps a spherical-Earth coordinate system into a Cartesian coordinate system (a sphere in a box; Flanagan et al., 2007) and incorporates a finite-difference travel time calculator and spatial smoothing constraints. This algorithm is designed to solve jointly for 3D velocity structure and earthquake locations using both first-arrival times and differential times, leading to improved resolution in the seismically active areas where the differential data provide dense sampling.

Figure 2. Event and station distributions in our study area and starting inversion grid nodes for the 3D coarse model (30-km horizontal spacing). (a) earthquakes; (b) controlled sources; (c) stations; (d) inversion grid nodes (small panel shows the 1D starting velocity model).
3D Coarse Model

Because of the large spatial scale and amount of data in our study, we first solve for a coarse 3D $V_P$ model starting with a one-dimensional (1D) velocity model (shown in the small panel of Fig. 2d) for the entire state. This 1D model is based on standard regional 1D velocity models used to locate earthquakes by the seismic networks in northern and southern California. The starting model nodes (shown in Fig. 2d) are uniformly spaced at 30 km intervals in the horizontal directions and extend 570 km in the SW–NE direction and 1320 km in the NW–SE direction. In the vertical direction, the nodes are positioned at $0.0001, 1, 4, 8, 14, 20, 27, 35,$ and $45$ km (relative to mean sea level). We only use absolute arrival times for this 3D coarse model. An a priori Moho is not included at this stage, but is introduced later for the finer-scale model. Preliminary inversions were carried out using the tomography algorithm simul2000 (Thurber and Eberhart-Phillips, 1999). This algorithm simultaneously solves for 3D velocity structure and earthquake locations using the first-arrival times employing an iterative damped-least-squares method. This step was taken for data quality control purposes (i.e., identifying poorly constrained events and picks with very high residuals), and to provide formal but approximate estimates of velocity model resolution and uncertainty. After the data quality control step using simul2000, we applied the regional-scale DD tomography algorithm, which is more suitable for the large-scale area.

<table>
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<th>Study Area</th>
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<td>Coalinga</td>
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Table 2

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in this study. To make the inversion more stable, some regularization method is required, such as smoothing and damping. Smoothing regularization provides a minimum-feature model that contains only as much structure as can be resolved above the estimated level of noise in the data (Zhang and Thurber, 2003). Damping and smoothing are often selected empirically, by running a series of single-iteration inversions with a large range of values, and plotting the data variance versus model variance trade-off curves (e.g., Eberhart-Phillips, 1986, 1993). We explored a wide range of damping (from 25 to 1000) and smoothing (from 0 to 1000) to make sure that we looked at the entire trade-off curve instead of a portion of it. The smoothing constraint weighting of 100 (same for the horizontal and vertical directions) and the damping parameter of 350 were chosen by examining these trade-off curves, which produced good compromises between data misfit and model variance (see Figure S1 in the electronic edition of BSSA).

3D Starting Model Adjustments

Our preliminary modeling work did not include an a priori velocity increase at the Moho, but first-arrival body-wave tomography by itself was not capable of imaging sharp discontinuities. Thus, after we obtained the 3D coarse velocity model, we introduced an a priori Moho interface (see Figure S2 in the electronic edition of BSSA) from the results of Fuis and Mooney (1990), which was a modification of Mooney and Weaver (1989). We set the velocity to 8 km/sec in the model layer right below the Moho and 8.2 km/sec at the deepest layer with a linear gradient for the layers between these two layers. In order to start with a conservative 3D model, we removed the low-velocity anomalies in the 3D coarse model, that is, we require that velocity is initially a monotonically increasing function of depth. The resulting adjusted 3D model is the starting model for our final P velocity model. Figure 3a shows the map view of this model at 4 km depth, with the layer-average velocity values in the inset of Figure 3b. In order to check the effects of the inclusion of the Moho interface on our final results, we compared our final model with a model obtained without a Moho (low-velocity anomalies are still removed). The two models are quite similar over well-resolved areas and the model differences between the initial models are reduced after the inversion, indicating that our process converges. Table S1 in the electronic edition of BSSA shows a comparison of models with and without the inclusion of a Moho interface.

In order to use differential times to obtain a finer-scale model given our computer memory limitations, we split the entire state into five subregions (Fig. 3a). The adjacent subregions overlap by about 30 km. We use the same depth layers as the coarse model. Figure 3b shows the event and station distributions for subregion 1 as an example. For each subregion inversion, we use absolute and differential times from events inside the subregion (blue circles in Fig. 3b) that are recorded by all the stations (black triangles) and only absolute times from events outside of the subregion (pink circles) that are recorded by stations inside of the subregion. In this way, we improved the resolution of the resulting velocity model for

Figure 3. (a) Map view of the 3D coarse velocity model at 4 km depth and the boundaries of the 5 subregions. (b) Event and station distribution in the subregion 1 (small panel shows the 1D layer-average velocity). Yellow squares: finer inversion nodes inside of subregion; green squares: nodes with fixed velocities; blue circles: earthquakes inside of subregion 1; pink circles: earthquakes outside of subregion 1 but recorded by stations inside of the subregion; black triangles: permanent and temporary stations; red stars: active sources (shots and quarry blasts). Please refer to the text for more details.
the deeper layers due to the inclusion of substantial data at large epicentral distances. We also include all available explosion and quarry data for each subregion (red stars). Inside of each subregion, the horizontal nodes are spaced at 10 km intervals in the areas with dense data coverage and 20 km in other areas (yellow squares). The node spacing outside of each subregion is 30 km from the coarse model (green squares). The initial velocity value at each node is computed from the velocity values at the surrounding eight nodes of the coarse initial model using trilinear interpolation as described in Thurber and Eberhart-Phillips (1999). The velocities outside of each subregion are fixed during the inversion of the inside-subregion velocities. We inverted five local models separately; our final statewide velocity model is a stitched version of all the five subregion models. The velocities in the areas of overlap are computed as the average velocities of the two adjacent subregions. In order to test the robustness of our stitching approach, we inverted a model that includes subregions 1 and 2 and compared the resulting model with the stitched model. The two models are quite similar with some minor differences that are likely caused by different regularization parameters, which are determined individually for each inversion by examining data variance versus model variance trade-off curves.

Model Quality and Resolution

The quality of our model can be evaluated by its ability to (1) fit the observed arrival-time data, and (2) produce accurate locations for onland controlled-source explosions that have known coordinates. Figure 4 shows a comparison of the arrival-time residual distribution before (a) and after the coarse (b) and final (stitched) (c) 3D velocity inversions for the entire data set (including controlled sources). The root-mean-square (rms) misfit is reduced by over a factor of 3, from 1.26 sec to 0.37 sec, after the 3D coarse model inversion, and then to 0.32 sec after the final model inversion. The reason that the fit to absolute times in the subregion model inversions is only slightly better than the coarse model fit is because the purpose of this step is to improve fine-scale velocity resolution by using differential times. A hierarchical weighting scheme is applied with greater weight to the absolute data for the first two iterations and greater weight to the differential data for the next four iterations. Note that most of the improvement of arrival-time fit after the 3D final model inversion is mainly due to the lower differential time residuals, where the rms misfit of the differential times is reduced from 0.28 sec to 0.11 sec.

We independently located the onshore explosions using the starting 1D and the coarse and final 3D velocity models and then calculated the horizontal and vertical location differences between the relocations and the known true locations. Figure 5 shows histograms of shot location accuracy relocated using the starting 1D model compared with the two 3D models for both horizontal and vertical coordinates. The horizontal location errors are all positive; the vertical errors are positive when the assigned location is deeper than the true location and negative when the assigned location is shallower than the true location. For the 1D model, the error distributions are quite broad, with a mean error of 1.23 km, and a standard deviation of 1.08 km horizontally. The vertical error distribution has peaks at about 0.6 and 4.5 km, a mean (absolute) error of 2.19 km, and a standard deviation of
2.38 km. In contrast, the 3D coarse model error distributions are peaked between 0 and 1 km, with mean errors of 0.60 and 0.32 km, and standard deviations of 0.59 and 0.82 km for the horizontal and vertical errors, respectively. Although this model is coarse, the 3D shot relocations are significantly improved, especially in depth. This is because a single 1D-velocity model cannot account for lateral heterogeneity in velocity structure across all of California. Further, the 3D final model location error distributions are peaked around 0.2 km, with mean errors of 0.35 and 0.15 km and standard deviations of 0.39 and 0.41 km for the horizontal and vertical errors, respectively. The poorly located shots fall into several

Figure 5. Histograms of the differences between relocations and known true locations for onland explosions. The two columns are for horizontal and vertical location errors, respectively. (a) and (b) 1D model; (c) and (d) 3D coarse model; (e) and (f) 3D final model.
categories. One group is distant shots recorded on permanent network stations for cases in which we were not able to obtain the corresponding refraction profile picks. An example is the shots from the PACE 1992 project (Fliedner et al., 1996). Others are cases for which data for a particular shot were split into two separate events, due to the fact that the data were obtained and entered into the database separately. Examples are a number of shots from the Parkfield area and several LARSE shots. Finally, there are a few shots for which only profile picks are available, and the recording geometry is too poor to constrain the locations adequately. The reduction in relocation errors of about a factor of 2 over the 3D coarse model indicates that our final model significantly improves resolution for the lateral heterogeneities in the 3D velocity structure, especially at shallow depths.

To assess the model quality, we performed a restoration and a checkerboard resolution test similar to those in Thurber et al. (2009). In the restoration test, event hypocenters, station locations, and synthetic travel times, calculated from the final inverted model, have the same distribution as the real data. We followed the same inversion strategies as those for the real data and examined the recovering ability of our algorithms. The inverted final model is similar to the true model over well-resolved areas (see Figure S3 in the electronic edition of BSSA). In the checkerboard test, the synthetic times are computed through the 1D starting velocity model with ±5% velocity anomalies across three grid nodes. The results are shown in Figure S4 (see the electronic edition of BSSA). Note that in this test, we did not include the Moho interface, but still removed low-velocity anomalies as was done for the real data inversion. Some smearing is still seen, part of which is likely due to the interpolation of velocities when obtaining the starting velocities for subregion models.

Final P-Wave Velocity Model

Map Views. Figure 6 shows map view slices through the resulting tomographic P velocity model. Pink dots in each figure represent earthquakes relocated within ±1 km of each layer depth. The white contours enclose the areas where the derivative weight sum (DWS; Thurber and Eberhart-Phillips, 1999) is greater than 300. Derivative weight sum measures the sampling of each node and serves as an approximate measure of resolution (Zhang and Thurber, 2007). Areas with DWS values above 300 correspond to well-resolved areas in the synthetic tests. In the following, we show average velocities at each layer computed for these areas. In order to quantitatively compare our model with previous tomography models, we interpolated the southern and northern California models by Lin et al. (2007) and Thurber et al. (2009) onto our inversion grids and calculated the correlation coefficients at each layer depth. We will refer to these two models as the SC and NC models in the following.

Figure 6a,b shows the P-wave velocities in the top two layers of our model. The average velocity values are 5.26 km/sec at 1 km and 6.0 km/sec at 4 km depth. The velocities in these shallow layers generally correlate with the surface geology. Lower values are observed in basin and valley areas, such as the Great Valley, southern San Joaquin Valley, Ventura basin, Los Angeles basin, and Imperial Valley, whereas relatively higher velocities are present in the mountain ranges, such as the northern Coast Ranges, Transverse Ranges, Peninsular Ranges, and Sierra Nevada. The correlation coefficients for the well-resolved areas of these two layers between our model and the NC model are 0.49 and 0.56, respectively. The relatively low correlations for these two layers are mainly due to the low-velocity anomalies in the Great Valley and fast anomalies in the Sierra Nevada in our model. The lowest velocity anomalies (about 2.9 km/sec) appear in the Great Valley and southern San Joaquin Valley. However, these slow anomalies are at the edge of our well-resolved areas because of the sparse event distribution in this region. Fairly high-velocity anomalies (∼6.0 km/sec) at 1 km depth in the Klamath Mountains and Mount Shasta area are observed that are consistent with the results from seismic-refraction and gravity data in this area (Zucca et al., 1986; Fuis et al., 1987), but are not seen in the recent northern California P-wave velocity model by Thurber et al. (2009). This high-velocity body extends to 14 km depth in our model, reaching ∼6.5–6.7 km/sec at 4 km, ∼6.7–7.1 km/sec at 8 km, and ∼7–7.1 km/sec at 14 km depth, with relatively little structural variations along the north-south direction. These velocities are consistent with the conclusion by Fuis et al. (1987), who argued that an imbricated stack of oceanic rock layers underlies the Klamath Mountains. Another high-velocity anomaly zone is apparent at 1 and 4 km depth in the Lake Oroville area. The ∼6.8 km/sec velocity at 4 km depth is generally consistent with the observations by Spieth et al. (1981) that the velocity is of the order of 7.0 km/sec at a depth of 5 km. This high-velocity anomaly (∼6.9 km/sec) extends to 8 km depth in our model. The high velocities at 1 km depth in the southern Sierra Nevada area, ranging from 5.2 km/sec to 5.8 km/sec, are consistent with the results of Fliedner et al. (1996, 2000). The velocities at 4 km depth are generally higher than those estimated by Thurber et al. (2009) (∼6.0 km/sec compared with ∼5.3 km/sec); our model is more consistent with the results based on the active seismic refraction experiment by Fliedner et al. (1996, 2000).

In southern California, the correlation coefficients for these two layers between our model and the SC model are 0.65 and 0.56, respectively. Note that these coefficients are computed over the resolved areas in the model of Lin et al. (2007), corresponding to the area of X = −100 to 200 km and Y = −620 to −100 km. Near-surface velocities in our model are also relatively high in the western Mojave Desert in our model. The anomalies are slightly higher than previous results (e.g., Hauksson, 2000; Lin et al., 2007). We think this may be due to the inclusion of the active-source data in this area, which were not used before. In the Imperial Valley area, the slowest velocity at 1 km depth is 3.07 km/sec in this study, but about 3.6 km/sec at the surface in Lin et al. (2007), who concluded that their model slightly overestimates
Figure 6. Map views of the $P$-wave velocity model at different depth slices. The white contours enclose the areas where the derivative weight sum is greater than 300. The average velocities are computed over these areas. Pink dots represent relocated earthquakes. Black lines denote coast line and lakes, gray lines rivers and surface traces of mapped faults. (Continued)
Figure 6. Continued.
the near-surface velocity compared with seismic refraction results (Fuis et al., 1984). The reduction of this overestimation indicates that our model has better resolution for near-surface structure. The southern San Joaquin Valley is better resolved in this new model, which is at the northern boundary of the study area in Lin et al. (2007).

Figure 6c,d shows map views for 8 and 14 km depths, with average velocity values of 6.26 km/sec and 6.46 km/sec, respectively. These layers are the two best-resolved layers in our model because of the abundant seismicity at these depths, and the results are generally quite compatible with previous tomographic results. At 8 km depth, a strong velocity contrast is apparent between the Great Valley and the Sierra Nevada. At 14 km depth, some of the features we imaged for the shallow layers are reversed, that is, the basin and valley areas show relatively high-velocity anomalies and lower values are present under the mountain ranges. The reversal of the velocity anomalies associated with most of the major basins is also observed in previous southern and northern California tomography studies (Lin et al., 2007; Thurber et al., 2009). The correlation coefficients for these two layers between our model and the NC model are 0.61 and 0.47, and 0.35 and 0.49 with the SC model, respectively. The relatively poor correlation with the NC model is mainly due to the slower velocities (~5%) in the northern Coast Ranges than what is observed in the Thurber et al. (2009) model. The correlation coefficients with the SC model are reduced compared with the shallower layers. For these two layers, our model is generally faster than the Lin et al. (2007) model by about 5% in the basin areas, such as the Ventura basin, Los Angeles basin, and Imperial Valley.

Map views for the 20 and 27 km depth layers are shown in Figure 6e,f, with average velocity values of 6.81 km/sec and 7.28 km/sec, respectively. The correlation coefficients for these two layers between our model and the NC model are 0.65 and 0.72, respectively. The resolution of the southern California model by Lin et al. (2007) is poor below 17 km depth, so we focus on the comparison in northern California. At 20 km depth, the model is consistent with the results of Thurber et al. (2009), but is slightly slower in the center of the Great Valley. At 27 km depth, the Sierra Nevada area shows about 6.0 km/sec low-velocity anomalies, but in the same area, the velocity in Thurber et al. (2009) is about 6.5 km/sec. Our model extends to 45 km depth. Figure 6g,h shows the map views of the last two layers at 35 km and 45 km depths. Although the model is not resolved nearly as well as the shallower layers, we are able to see the low-velocity anomalies in the Sierra Nevada region; the correlation coefficients for these two layers between our model and the NC model are 0.70 and 0.46, respectively.

Cross Sections. We present three cross sections to illuminate the large-scale features of the model. One is parallel to the San Andreas fault (SAF; X = 0 km in the Cartesian coordinate system), and the other two are perpendicular to the SAF (Y = 210 km and Y = −30). A complete set of cross sections is provided in the BSSA electronic edition of BSSA. In Figure 7 we show the velocity cross sections through the resulting model along the three profiles whose locations are shown in Figure 1.

The X = 0 km section in Figure 7a starts in the northern Coast Ranges where intermediate velocities ($V_p < 6.2$ km/sec) extend into the lower crust. At depths greater than 20 km, the seismicity and high velocities of the subducting Gorda Plate are visible. From Y ∼ 350 to −210 km, the low near-surface velocities of the Great Valley and southern San Joaquin Valley sediments and sedimentary rocks are evident, extending to depths of ∼10 km in the northwest and to ∼4 km in the southeast. High-velocity rocks ($V_p > 6.5$ km/sec) of the underlying Great Valley ophiolite body are present throughout this part of the section. The section crosses the Garlock fault (Y ∼ −210 km) and the SAF (Y ∼ −255 km), where upper and midcrustal velocities are relatively low ($V_p < 6.3$ km/sec), and then cuts through the San Gabriel Mountains (SGM) and Peninsular Ranges where the upper crust velocities are relatively high ($V_p > 6.2$ km/sec) at shallow depths. Beneath the SAF and SGM, a strong low-velocity zone is apparent, as identified in

![Figure 7](image_url)
previous studies in this area, which has been interpreted to indicate fluids (e.g., Ryberg and Fuis, 1998; Fuis et al., 2000).

The section in Figure 7b cuts across the seismically quiet southern San Francisco (SF) Peninsula and SF Bay ($X = -120\, \text{to} \, -90\, \text{km}$) and then reaches the seismically active Hayward, Calaveras, and Greenville faults beneath the East Bay ($X = -70\, \text{to} \, -30\, \text{km}$). The section then enters the Great Valley, where the high-velocity basement, thought to be ophiolite (e.g., Godfrey et al., 1997), shallows to the northeast ($X = -30\, \text{to} \, +50\, \text{km}$). After that, the section enters the Sierra Nevada where a thicker crust with a velocity of $\sim 6.2\, \text{km/sec}$ extends to $32\, \text{km}$ depth. The section in Figure 7c passes through the seismic activity of San Simeon ($X = -120\, \text{km}$), Parkfield ($X = -75\, \text{km}$), and Coalinga ($X = -30\, \text{km}$). Even with the 10 km model gridding, the velocity contrast across the San Andreas at Parkfield is evident (southwest side faster, e.g., Thurber et al., 2006). In this section as well, the high-velocity Great Valley ophiolite body is evident with a predominantly southwestern dip of its upper surface, consistent with potential field data (Jachens et al., 1995). At $X \sim 50\, \text{km}$, we see a transition to the slower, thicker crust of the Sierra Nevada.

S-Wave Velocity Model

Although S-wave velocity models in northern California are available from ambient noise and surface wave data (e.g., Yang et al., 2008), there is no 3D model based on regional network data. In this study, we use the S first-arrival times from the SCSN and USAArray to solve for a $V_S$ model. Note that these data are different from those used for the $V_P$ model. Figure 8a shows the 1020 SCSN and 1292 USAArray events with at least 4 $P$ and 4 $S$ picks. Due to the sparse distribution of the data, we use the velocity inversion nodes of the 3D coarse $V_P$ model (i.e., 30 by 30 km horizontal node spacing). The starting $S$ velocities are derived from our resolved $V_P$ model and a constant $V_P/V_S$ of 1.73. The resolution estimated by the DWS values is quite poor. In order to test the robustness of the $S$ model, we also start with the $S$ velocity values from the ambient noise and teleseismic multiple-plane-wave tomography results by Yang et al. (2008). The well-resolved part agrees with the results starting with the constant $V_P/V_S$, indicating that the model is relatively robust. Figure 8b shows the map view of our resolved $V_S$ model at 8 km depth. The white contours enclose the area where the derivative weight sum is greater than 100. We also compared our $V_P$ and $V_S$ models to the empirical $V_P/V_S$.

Figure 8. (a) Event distribution for $V_S$ model. The red and blue circles represent the events from the USAArray and SCSN, respectively. (b) Map view of our resolved $V_S$ model at 8 km depth. Areas where the derivative weight sum is greater than 100 are shown. The average velocity is computed over these areas. The pink dots represent relocated earthquakes.
relation of Brocher (2005). This relation is derived from a diverse dataset, including wireline borehole logs, vertical seismic profiles, laboratory measurements, and seismic tomography models, and can be used to infer $V_S$ for the entire Earth’s crust from $V_P$. Using the layer-average $P$ velocities as inputs, we obtained the corresponding empirical $S$ velocities. Our comparison indicates that the tomography-based $V_S$ results are faster than predicted by the empirical relation at 1 and 35 km depth, and lower than expected at 8 km depth. Due to the poor resolution of this model, we do not attempt to solve for Poisson’s ratio and other parameters that depend on $V_P/V_S$ values. A complete set of map views and cross sections of our $S$-wave model is provided in the electronic edition of BSSA).

Discussion

Our model is the first 3D seismic velocity model for the entire state of California based on local and regional arrival-time data. It has improved areal coverage compared with the previous northern and southern California models, and extends to greater depth due to the inclusion of substantial data at large epicentral distances. The combination of northern, southern, and central California data sets results in better-resolved velocity structure at the study boundaries of previous tomographic models, such as the San Joaquin Valley and southern Sierra Nevada. Because of the 10 km horizontal grid spacing in our model inversion, which is larger than the distance cutoff of most waveform cross-correlation calculations ($\leq 5$ km), we did not apply any differential times from cross-correlation in this study. There may be some finer-scale structures that are not resolved due to the data and grid spacing used in our model. We compared our model with some results based on refraction and/or reflection data. Our model generally agrees with most of the studies, such as in the Diablo and Gabilian Ranges (Steppe and Robert, 1978; Walter and Mooney, 1982), the Coyote Lake (Mooney and Luetgert, 1982), the Long Valley (Luetgert and Mooney, 1985), the San Francisco Bay area (Holbrook et al., 1996), the southern Sierra (Fliedner et al., 1996), and the Mojave Desert (Fuis, Ryberg, et al., 2001); but slightly overestimates near-surface velocity values in some basins and valleys, such as in the Imperial Valley (McMechan and Mooney, 1980; Fuis et al., 1984), the Great Valley (Colburn and Mooney, 1986), the Livermore area (Meltzer et al., 1987), and the greater Los Angeles basin (Fuis, Ryberg, et al., 2001b).

The differences between this statewide velocity model and previous regional-scale models are due to several factors, such as data sets, grid spacing (cell size), tomographic algorithms, and inversion parameters (e.g., damping, smoothing, and residual weighting). The model is very similar to the recent northern California model by Thurber et al. (2009) for the middle to lower crust because the two studies use the same type of data sets (both absolute and differential times) and inversion algorithm (tomofDD), whereas the southern California model by Lin et al. (2007) is derived by applying the SIMULPS algorithm (Thurber, 1983, 1993; Eberhart-Phillips, 1990; Evans et al., 1994) to absolute arrival times for composite events. We also calculated the correlation coefficients, which are 0.67, 0.56, 0.36, 0.54, and 0.69 for the top five layers, between our model and the SCEC unified velocity model (Version 4, Magistrale et al., 2000). Note that the SCEC model is based on geotechnical borehole seismic velocity data and the regional tomographic model of Hauksson (2000), which is also obtained by applying the SIMULPS algorithm. The slightly better correlation at 1 km depth is due to the low-velocity anomalies in the basin areas. Our new model is generally consistent with these previous results. The improved resolution of our model in near-surface layers over the previous California tomographic models is mainly due to the large amount of active-source data in this study.

The goal of this study is not to replace the previous tomographic models in California that have more detail than can be resolved by our data and grid spacing, but to image the entire state of California at a regional scale, to reveal some features that are difficult to resolve in local studies, and to provide the geophysical community with a velocity model that should be useful for regional-scale studies, such as regional waveform modeling. The model is available in the electronic edition of BSSA.

Conclusions

We have developed statewide body-wave tomography models ($P$ and $S$) for California using absolute and differential arrival times from earthquakes, controlled sources, and quarry blasts. By merging the data sets from networks in northern, southern, and coastal central California and USArray, we have achieved relatively complete coverage of the entire state for $V_P$. By including a large amount of active-source data in this study, we obtained improved resolution in near-surface layers over the previous California tomographic models, especially in the largest sedimentary basins, such as the Great Valley, the Imperial Valley, and the Los Angeles basin. At 8 km depth, there is a clear north to south increase in the average velocity of the Coast Range Mountains extending from the Mendocino Triple Junction to the border with Mexico. At 14 and 20 km depths, the basin and valley areas show relatively high-velocity anomalies, with lower velocities present under the mountain ranges. The Great Valley ophiolite body and the subducting Gorda Plate are evident from the cross sections. Low-velocity anomalies in the Sierra Nevada exist from midcrustal to greater depths; the slow velocity root over this area is the largest anomaly at 35 km depth. Our model provides a reasonable fit to the data and relocates explosions, treated as earthquakes, with an absolute accuracy of better than a kilometer. Thus, it should be useful for producing a statewide earthquake location catalog based on a single velocity model.
Data and Resources

Active-source data used in this study were collected from published studies listed in the references. Catalog picks were obtained from the USArray, the Northern California Earthquake Data Center (NCEDC), and the Southern California Earthquake Data Center (SCEDC) and originate principally from the Northern California Seismic Network (NCSN) and Southern California Seismic Network (SCSN). Some figures were made using the public domain Generic Mapping Tools software (Wessel and Smith, 1991).

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