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Abstract

We present results from Multi-scale InSAR Time Series (MInTS) [1] analysis of L-band (ALOS PALSAR) and C-band (ERS and Envisat) interferometric data sets over Central California. MInTS was designed to exploit the correlation of phase observations over space to improve the signal-to-noise ratio in the estimated deformation time-series compared to the traditional time-series InSAR techniques [2,3]. Traditional time-series analysis techniques assume the statistical independence of InSAR phase measurements over space and time when estimating deformation. However, existing atmospheric phase screen models [4,5] clearly show that noise in InSAR phase observations is correlated over the spatial domain. The MInTS technique reduces the set of InSAR observations to a set of almost uncorrelated coefficients at various spatial scales using wavelets. Traditional inversion techniques can then be applied to the wavelet coefficients more effectively, thus significantly improving the signal-to-noise ratio.

Our results represent the first study of inter-seismic deformation across the Central San Andreas using L-band data. We clearly observe the transfer of offsets from the San Andreas to the Calaveras Fault network just south of Hollister, CA and observe that the offset across the Calaveras Fault persists as far north as San Jose, CA. The region around the Central Calaveras Fault is characterized by heavy decorrelation and previous InSAR studies have failed to reliably estimate the creep rate across this section. The ability of our analysis technique to detect sub cm/yr deformation rates reliably is also demonstrated with example profiles across the Hayward Fault in the Bay Area. We also present evidence for a change in deformation rate in the vicinity of the southern end of the creeping section after the Parkfield, CA possibly driven by a long temporal scale post-seismic mechanism.

Data and Methodology

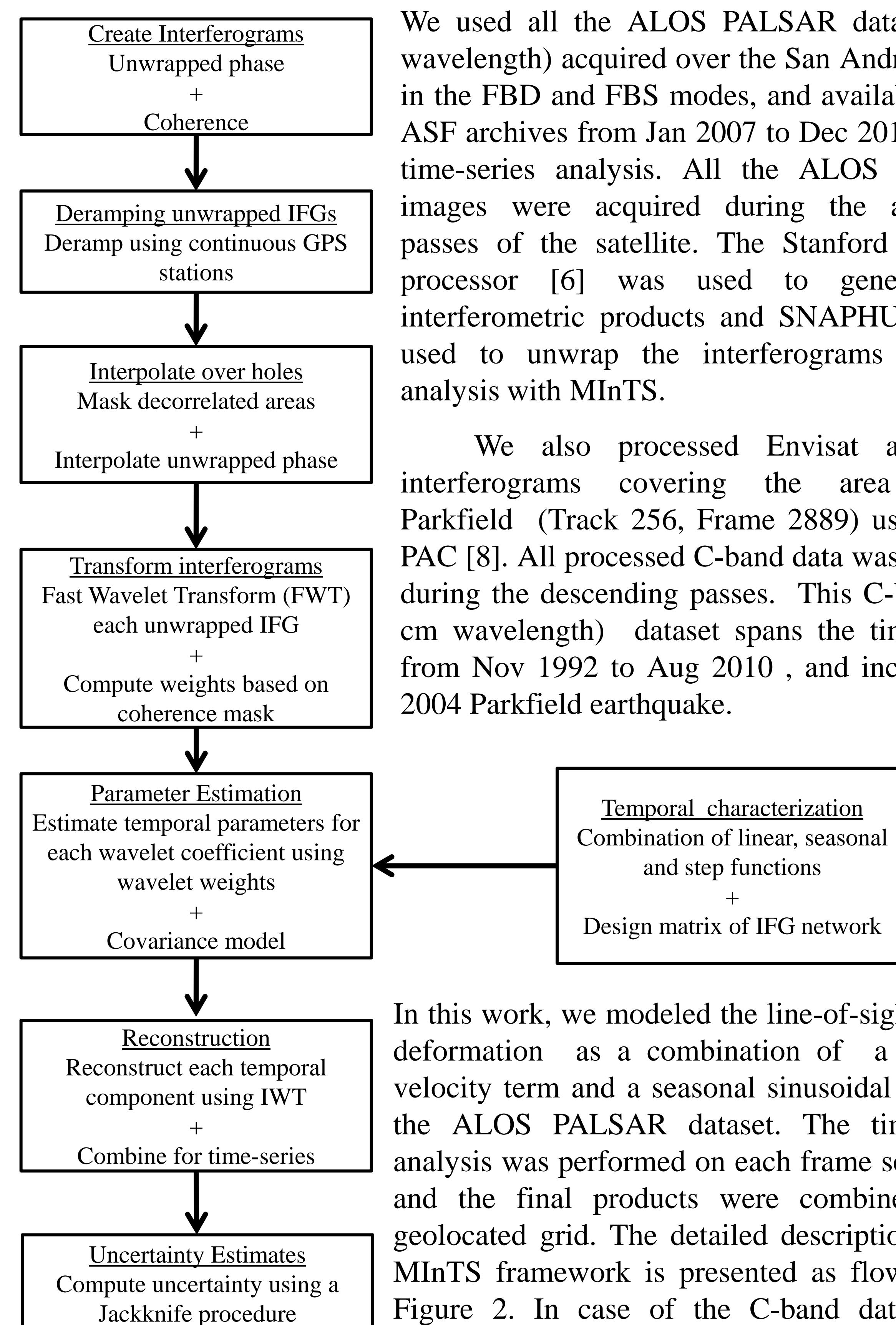


Figure 1. Detailed flowchart of the various processing stages in our MInTS analysis of the InSAR data sets [1].

L-band analysis

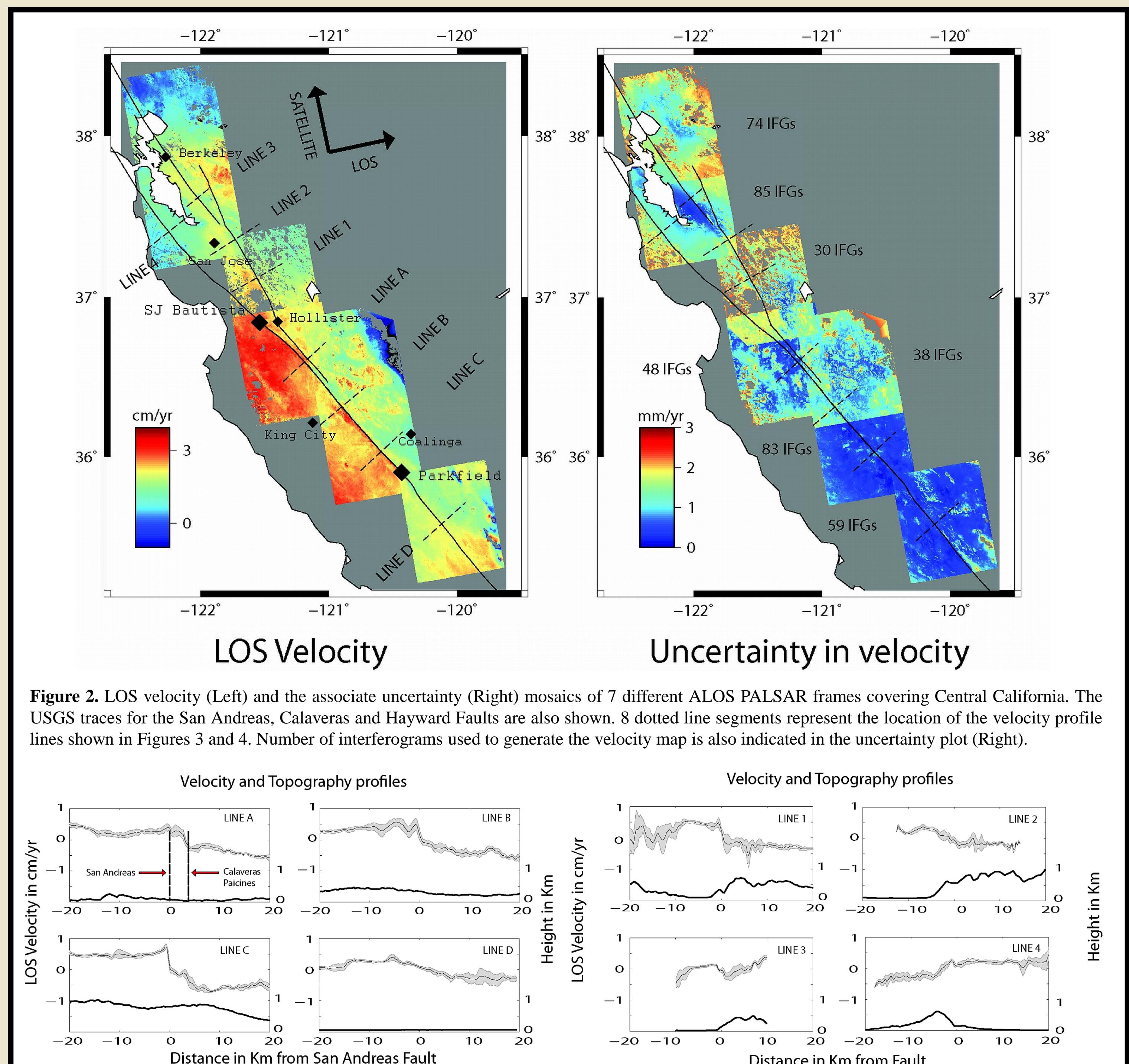


Figure 2. LOS velocity (Left) and the associate uncertainty (Right) mosaics of 7 different ALOS PALSAR frames covering Central California. The USGS traces for the San Andreas, Calaveras and Hayward Faults are also shown. 8 dotted line segments represent the location of the velocity profile lines shown in Figures 3 and 4. Number of interferograms used to generate the velocity map is also indicated in the uncertainty plot (Right).

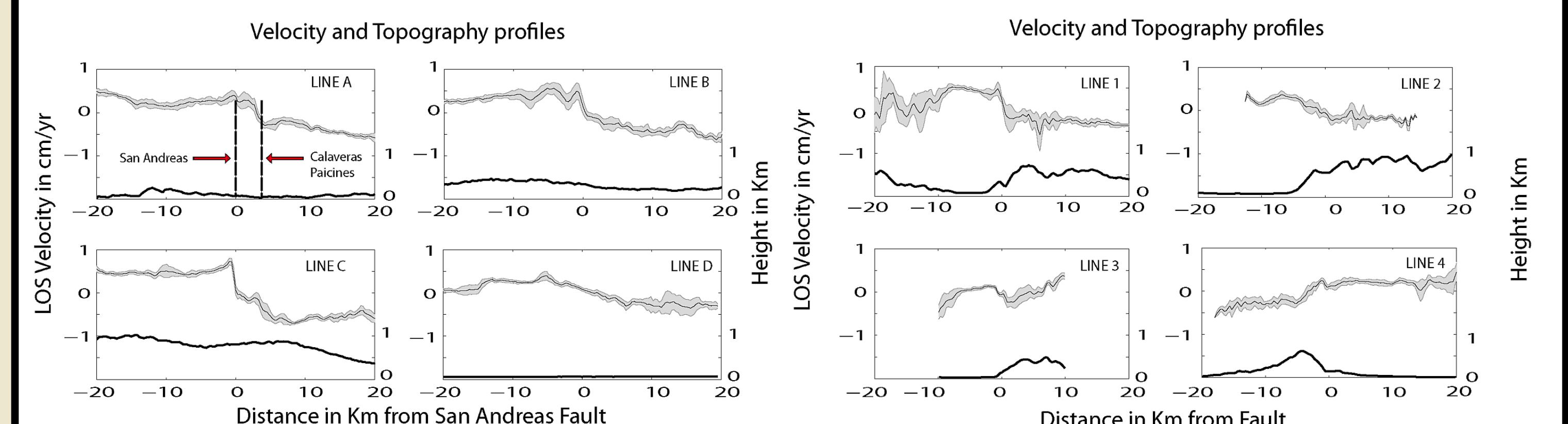


Figure 3. LOS velocity profiles (shaded) and topography profiles (solid) plotted along the dotted lines across the San Andreas Fault as shown in Figure 2. A region of 6 km was averaged along the profile lines.

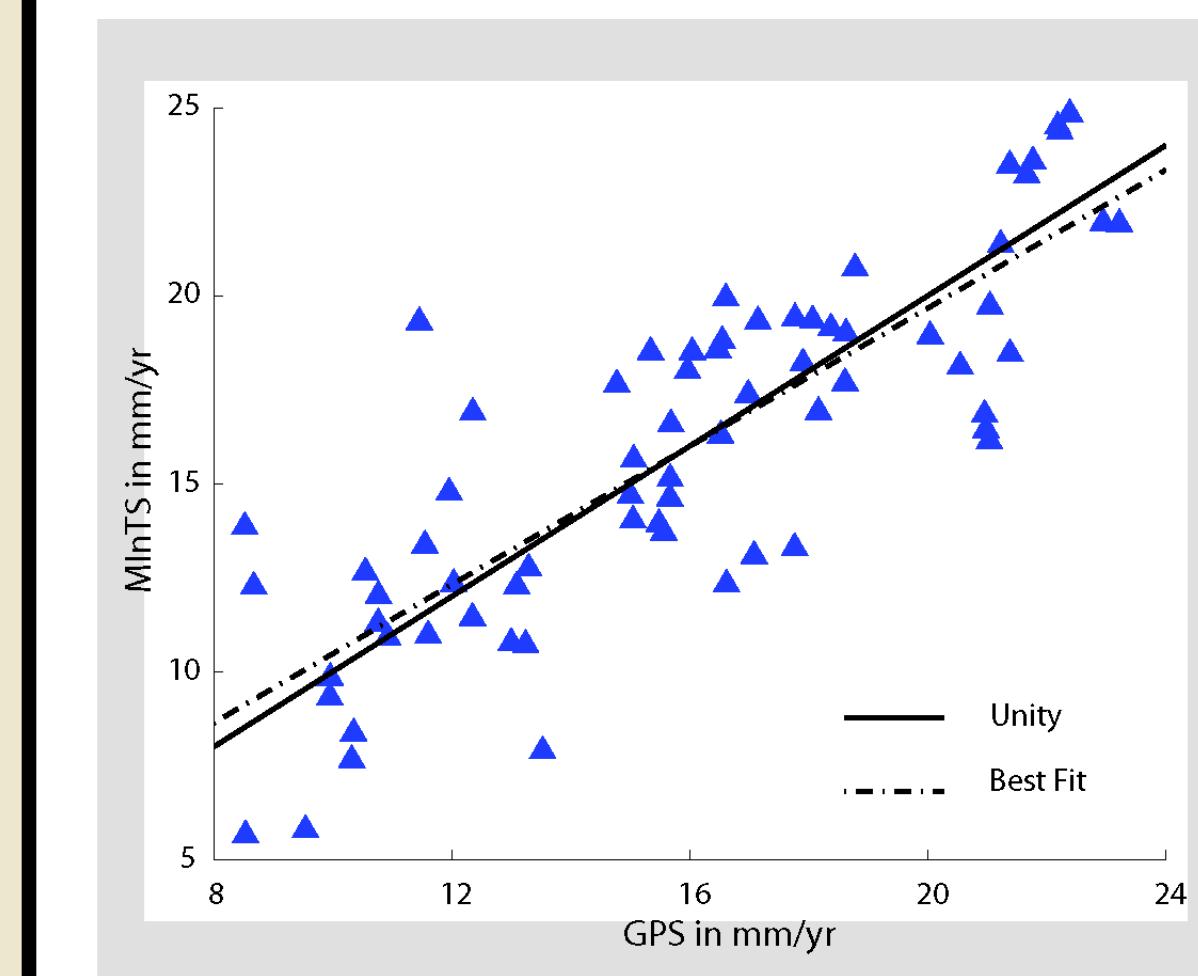


Figure 4. LOS velocity profiles (shaded) and topography profiles (solid) plotted along the dotted lines across the Calaveras and Hayward Faults as shown in Figure 2. A region of 6 km was averaged along the profile lines.

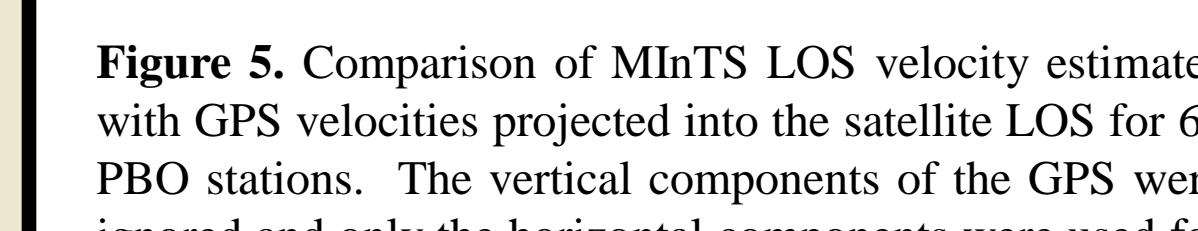


Figure 5. Comparison of MInTS LOS velocity estimates with GPS velocities projected into the satellite LOS for 69 PBO stations. The vertical components of the GPS were ignored and only the horizontal components were used for the comparison. Best fit line has a slope of 0.98 and rms value of 2mm.

- The creeping section of the San Andreas Fault is clearly visible in the ALOS stacks (Figure 3). Surface creep across the Hayward and Calaveras Faults are also observed (Figure 4).
- This represents the first InSAR surface observations of creep across the Central Calaveras section (Lines 1 and 2 in Figure 4).
- The discrepancy between InSAR observations and LOS projected GPS velocities is on the order of 2 mm/yr. This is approximately twice that observed at C-band.
- Single LOS observation prevents us from separating horizontal and vertical deformation, particularly close to the fault.

Estimating Uncertainty

We developed a variant of the statistical Jack-knife procedure to compute uncertainties in the various estimated parameters. Set of InSAR phase observations in a network of interferograms, is correlated due to the inclusion of a single SAR image in many interferograms. Hence, conventional statistical techniques, that assume independent set of observations are no longer applicable. Hence, we determine uncertainties as follows:

1. We create sub-networks from the main interferogram network by removing all interferograms corresponding to individual SAR scene, for each of the SAR scenes.
2. We re-estimate the parameters of interest for each of these sub-networks.
3. We realize that the a single interferogram participates in many of these sub-networks and hence, these sets of estimated parameters are correlated.
4. We compute an approximate correlation matrix, taking into account the number of common interferograms between each of the sub-networks.
5. We reduce the problem to one of determining the mean and variance of a correlated set of observations, using our correlation matrix from Step 4.

All the uncertainties shown in this poster, were derived using the procedure described above.

C-band analysis

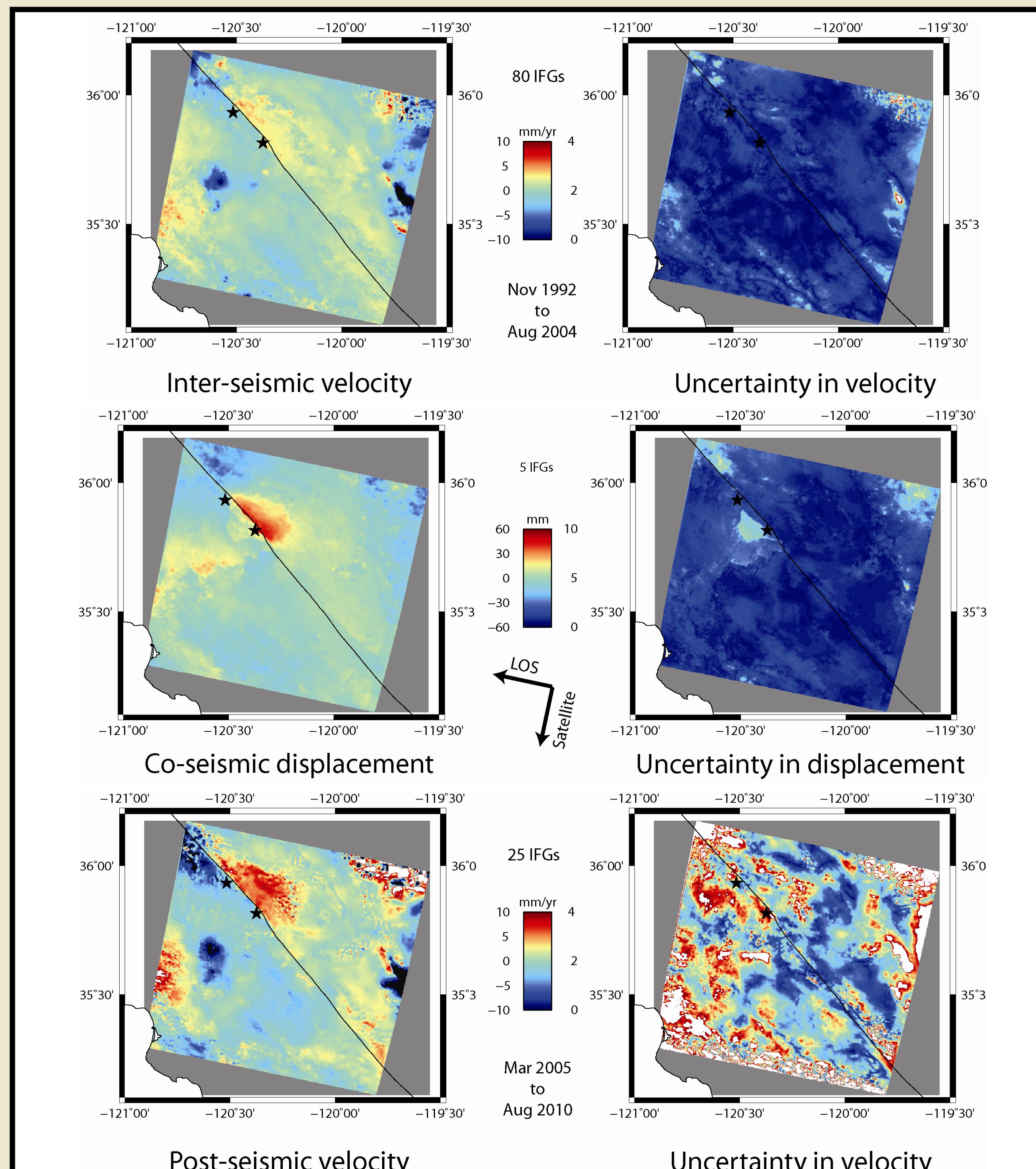


Figure 6. Estimated MInTS parameters and associated uncertainties for the C-band stack around Parkfield, CA. The epicenters of the 1966 and 2004 earthquakes are also shown. The number of interferograms and time-span of each of the stacks is also shown. The co-seismic displacements shown includes the rapid deformation (1 month time-scale) following the 2004 earthquake.

- The C-band stacks clearly show the transition zone between the creeping and locked sections of the San Andreas Fault around Parkfield, CA.
- C-band dataset spans a longer time-period (1992-2010) but is less coherent than that L-band stacks generated over the same region.
- L-band interferogram networks are highly redundant, whereas C-band networks are sparse. Most C-band SAR scenes participate in only one or two interferograms.
- The average LOS velocity after the 2004 Parkfield EQ is significantly higher than the inter-seismic rate, possibly due to the effect of long temporal scale post-seismic deformation mechanisms.

Things To Do

1. Process more C-band stacks acquired on descending passes.
2. Use the L-band and C-band stacks with GPS observations to generate a 3D interseismic displacement map for Central California.
3. Include sparse parameter estimation in MInTS for better modeling of EQ time-series.
4. Model the estimated 3D velocities, taking into account the associated uncertainties, using a Bayesian approach.

References

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