The 2014 M_w 6.0 Napa Earthquake, California: Observations from Real-time GPS-enhanced Earthquake Early Warning

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Recently, progress has been made to demonstrate feasibility and benefits of including real-time GPS (rtGPS) in earthquake early warning and rapid response systems. Most concepts, however, have yet to be integrated into operational environments. The Berkeley Seismological Laboratory runs an rt-GPS based finite fault inversion scheme in real-time. This system (G-larmS) detected the 2014 M_w 6.0 South Napa earthquake in California. We review G-larmS performance during this event and 13 aftershocks and present rt-GPS observations and real-time modeling results for the main shock. The first distributed slip model and magnitude estimates were available 24s after the event origin time, which, after optimizations, was reduced to 14s ($\approx 8s$ S-wave travel time, $\approx 6s$ data latency). G-larmS' solutions for the aftershocks (that had no measurable surface displacements) demonstrate that, in combination with the seismic early warning magnitude, M_w 6.0 is our current resolution limit.

Key Points:

1. GPS-based distributed fault slip model derived in real-time for early warning 2. Reasonable first order approximation of South Napa slip distribution at $M_w \approx 5.9$

3. $M_w 6.0$ is detection limit for current N-California rtGPS network and processing

1. Introduction

The importance of including real-time GPS (rtGPS) into earthquake early warning (EEW) systems has been recognized for a few years and found wide acceptance after the 2011 Tohoku-oki earthquake. Much progress has been made to demonstrate the feasibility and benefits of either GPS-only EEW, or seismic and GPS EEW integrations [Crowell et al., 2009; Allen and Ziv, 2011; Melgar et al., 2012; Wright et al., 2012; Ohta et al., 2012; Colombelli et al., 2013; Minson et al., 2014; Grapenthin et al., 2014]. The biggest contribution GPS brings to seismic P-wave detection algorithms is the near instantaneous measurement of permanent surface displacements during and after an earthquake. These data can be used to constrain slip on finite faults and hence derive a geodetic magnitude of the event. Tests of proposed concepts rely on simulated real-time replay of either real data from a different location, which implies a different station geometry, or synthetic data, which currently lack the dynamics of real events. Operational real-time analysis, on the other hand, provides the benefits of testing algorithms in their production location with realistic noise and data gaps, data latencies, network resolution, and resource requirements.

The California Integrated Seismic Network ShakeAlert [*Böse et al.*, 2014] is a real-time EEW demonstration system for California. Currently three algorithms triggering on Pwave arrivals in seismic data feed event detections into a Decision Module, which combines magnitude, location, and origin time estimates from the algorithms and sends alarms to its subscribers. During this demonstration phase ShakeAlert subscribers range from users in science and industry to triggered algorithms.

Grapenthin et al. [2014] describe a partially triggered, least-squares based rtGPS staticoffset inversion algorithm (G-larmS), which has been tested in real-time at the Berkeley Seismological Laboratory (BSL) since the beginning of May 2014. G-larmS detected the M_w =6.0 South Napa earthquake that nucleated on August 24, 2014 at 10:20:44 UTC near Napa, California (Figure 1). The event was recorded by a network of 58 real-time high-rate (1 Hz) GPS stations in the greater San Francisco Bay Area; a combination of stations from the Bay Area Regional Deformation [*BARD*] network, operated by the BSL, the Plate Boundary Observatory (PBO) operated by UNAVCO, Inc. and those operated by the USGS, Menlo Park. The BSL generates real-time displacement time series for a network of these stations (Figure 1) from which G-larmS estimates permanent surface displacements upon receipt of a ShakeAlert to infer a geodetic magnitude for the triggering event.

Here, we review the real-time online performance of G-larmS during this event and 13 aftershocks. This sequence provides a unique opportunity to study system performance based on random temporal sampling, but virtually stationary spatial sampling in a real-time environment. We compare the real-time analysis results of the main shock, which induced permanent surface displacements (up to 2.9 cm, 6 real-time stations within ≈ 25 km of the epicenter show more than 1 cm of permanent displacement [courtesy of Bill Hammond, UNR; based on GPS processing at Nevada Geodetic Laboratory]), to analysis results for the aftershocks. The aftershocks were too small to induce measurable motion at the surface, which gives us an opportunity to investigate the impact of real-time noise on solution quality.

2. Data Analysis

The rtGPS data are streamed into the BSL and analyzed in a network of 169 baselines (Figure 1). For each baseline, one station is assumed static (base station) while the motion of the other site (rover) is given relative to the base station. Positioning solutions for each baseline are generated by individual trackRT processes with ultra-rapid (predicted) orbits provided by the International GNSS Service [*Dow et al.*, 2009]. TrackRT is distributed as part of GAMIT/GLOBK [*Herring et al.*, 2010]. Further details on the GPS processing are given by *Grapenthin et al.* [2014].

The displacement time series generated by trackRT are streamed into G-larmS, which performs continuous quality analysis on the data. When triggered by ShakeAlert, G-larmS estimates static offsets along the baselines and inverts these for distributed slip on a finite fault from which geodetic magnitude is calculated. ShakeAlert currently consists of several algorithms that generate EEW messages based on P-wave detection in seismic data. As rtGPS alone is too noisy for P-wave detection even for large events [e.g., *Ohta et al.*, 2012, applies STA/LTA picker to static offset detection for the 2011 Tohoku earthquake], we turn G-larmS into a ShakeAlert subscriber and use the time between alarm receipt and S-wave arrival to set up the processing (e.g., Green's function generation). Details of the individual processing steps are described by *Grapenthin et al.* [2014], we provide only a summary.

Upon receipt of a ShakeAlert alarm G-larmS is intended to select a subset of baselines within a radius, $r \leq max(1.5 * 2^{M_w}, 50)$ (e.g., $r(M_w 6.0)=96$ km), to reduce the processing load (M_w is initial ShakeAlert magnitude). However, in the current test phase G-larmS

uses all baselines for all triggering events to test computational resource needs. The circle in Figure 1 encloses stations and associated baselines that would be used in a large production network (e.g., California-wide).

G-larmS calculates pre-event positions for these baselines by averaging buffered position solutions up to the ShakeAlert event origin time. The estimation of an average post-event position begins with the expected S-wave arrival at the site of the baseline that is closest to the event. The post-event position is an average over a time window that increases with new data arrivals. Static offsets are the difference of post-event and pre-event positions, and are used in a least-squares inversion for distributed slip on a finite fault.

In the inversion for slip, we center a vertical 50 km long fault (5 segments, 10 km length each) on the ShakeAlert epicenter. The strike is currently prescribed to be San Andreas Fault parallel (320° N). In width, the fault reaches from the surface to 12 km, the bottom of the seismogenic zone in this region. The analytical expressions for strike-slip and dip-slip by *Okada* [1985] provide Green's functions. The inversion routine (currently no weighting based on solution quality) estimates slip as soon as static offsets are available and repeats at every epoch. The solution is regularized through Laplacian smoothing with a constant smoothing factor.

3. Real-time Results

During the event data from 58 out of 61 real-time stations streamed into the BSL (except P189, P262, P298). This gives us 159 of 169 baselines with displacement solutions. Problems with sites P189 and P298 were related to the local configuration, which is now

corrected. P262 was not operational during a time period that includes this event, which created a hole in our triangulated processing network.

G-larmS received a ShakeAlert trigger at 10:20:49.5 UTC with an estimated event origin time at 10:20:44.4 UTC, a location about 3 km from the location in the ANSS catalog, and an initial magnitude of M_w =5.7 (r=78 km). The station closest to the event origin is OHLN at about 23 km south of the epicenter (Figure 1). Assuming an effective Swave velocity of 3 km/s, the arrival of static offsets was expected at 10:20:52 UTC for baselines involving OHLN (Figure 1). However, G-larmS produced the first static offset estimates and magnitude solutions 16 s later than that (24 s after origin time, 10:21:08 UTC). In addition to 8 s of S-wave travel time we observed 6 s of data latency and 10 s of additional latency due to a (now corrected) miscalculation of the wait-time for S-wave arrival (Figure 2). Much of the data latency is due to local buffering (4 s) in a BKG Ntrip Client [Weber and Mervart, 2009] to mediate data loss.

Figure 3 shows the displacement field 85 s after the event onset and hence can be considered 'final', i.e., only the large displacements at some sites in the far field (e.g., NW vectors) may still be impacted by dynamic displacements. The maximum real-time static offset estimates range from 1.0-2.7 cm (outlier removed, Figure 3 red vectors, some of that might be common mode signal). As we would expect for real-time data and the size of this event, these are over-estimates compared to post-processed daily positioning solutions at these sites (compare blue and red vectors in Figures 2 [dynamic], 3 [static in near field]). Real-time displacements (red vectors) in Figure 3 mostly agree with post-processed results (blue vectors, based on 2 days of post-event data) in azimuth (Supplement S1 animates

displacement field evolution). At that time the amplitudes in the near field have already decreased significantly from those in Figure 2, but are still overestimates compared to the post-processed results. The time series in the right panels in Figures 3 show this more clearly. The offset estimation begins during the last phase of the dynamic displacements, resulting in an initially much larger amplitude which gets averaged out over about 10-15 s. The magnitude time series fluctuates accordingly.

The middle panel in Figures 2, 3 shows the finite fault slip model and the fit of its predictions to the data at, respectively, 26 and 85 s past the event origin time. The maximum slip during this time is 6.1 cm. The slip maximum first gets pulled toward the south, where we have the initial observations to constrain the model. Towards the end of the process maximum slip is assigned to a fault patch slightly more north of the epicenter. The time series of the magnitude estimate is shown in the top right panel of Figure 3. We see slight variations in estimated magnitude at the beginning due dynamic shaking. The median over the first 60 s of solutions is M_w =5.86 with a generally good fit to the data (WRSS=0.05 m, Figure 3).

4. Event Replay Results

The bug-related wait time during the real-time analysis prevents us from analyzing the impact of dynamic motion on the real-time solutions. To gain an understanding of what can be expected for such events, we replayed true real-time displacements through G-larmS in simulated real-time. A debugged version of G-larmS estimated static offsets and slip models just like in real-time, but only for baselines within our magnitude based

selection area (Figure 1). The simulation does not add real-time latencies, so 6s should be added to time values in offset and magnitude time series to approximate real-time scenario (given in parenthesis below).

Figure 4 shows estimated offsets, slip model and magnitude 16 (22) seconds and 78 (84) seconds after the event origin time (time series animations in Supplements S3 and S4). These times correspond roughly to times in Figures 2 and 3. The displacement time series in the right panels clearly show dynamic motion due to S-wave and surface waves. While the dynamic motion causes overestimates of static offsets, they are damped quickly (≈ 10 s). The maximum magnitude is $M_w 6.18$ with maximum slip of 17 cm during dynamic motion. The impact of the dynamic motion on the magnitude decays at about 27 s (33 s) after the event onset when the magnitude reaches $M_w 5.99$ and decays slightly from there. Similarly to the real-time results, the final slip model shows most of the high-slip on patches slightly to the north and at the epicenter and gives a good fit to the observations.

5. Discussion

ShakeAlert did an excellent job alerting for this event: 5.1 s after the origin time an alert was issued, which delivered, for example, a 5 s S-wave alert time at the BSL. ShakeAlert's initial magnitude estimate was M_w 5.7, which briefly dropped to M_w 5.4 and then stabilized at M_w 5.8-6.0. The point source approximation inherent to seismic algorithms holds for events of this size and finite source solutions are usually not required. Hence, the South Napa earthquake was an excellent test case for our system and the seismic results provide validation.

The real-time (and replay) observations show permanent offsets induced by the earthquake in the displacement time series. However, when extracted, these static offsets are slight overestimates in most places; especially at the beginning of the magnitude estimation (compare left panels in Figures 2 and 3). As this still leads to reasonable magnitude estimates, mostly due to large fault surface area over which slip can be smoothed, it motivates the question on how our results for the South Napa earthquake compare to inversions of background noise.

Figure 5 shows the spread of magnitude estimates over the first 60 solutions for the real-time (151878) and replay (REPLAY) events and some of the aftershocks ($M_w 2.2$ -3.9, ShakeAlert IDs 15184-15210) until August 29. It is obvious that an $M_w 6.0$ event at the given distance (23 km) is at the lower limit of resolution of our current setup for the region. The magnitude estimates for the aftershocks shown in Figure 5 are based solely on noise in the real-time positioning solutions (ANSS magnitudes given in black on top of the horizontal axis, real-time ShakeAlert magnitudes given above that in gray). The medians for most events range between $M_w 5.2$ -5.8 with two events at $M_w 5.9$ and $M_w 6.0$. Clearly, the seismic system gives very reliable solutions in this magnitude range and is the primary way to decide whether G-larmS should send out an alarm. The final G-larmS production setup will send out magnitudes only if the seismic magnitude is greater than \approx 5.5. Another means to automatically evaluate solution quality is the model misfit (here: weighted residual sum of squares, WRSS). High magnitude estimates for noisy data should give higher misfit of model predictions, but exceptions exist (event 15209).

Generally, these observations suggest that the solutions for the $M_w 6.0$ Napa earthquake are reasonable and not solely based on noise in the network (compare WRSS).

The replay of the event, including code changes and exclusion of far field data fares only slightly better than the original event (15178). Some of the outliers at the high magnitude range are created during the time dynamic displacements traverse the network. This is inevitable and corrects itself quickly; the overestimates are still within ± 0.3 magnitude units, which is a reasonable goal for EEW applications. Implementing a low-pass filter to correct for the impact of dynamic motion would result in longer build-up times to the final magnitude [e.g., *Melgar et al.*, 2012].

During this sequence G-larmS had to process multiple alarms simultaneously and demonstrated that rapid foreshock-aftershock sequences as well as near-simultaneous, independent events (e.g., Northern California and Southern California) are handled well. However, additional work is required to ensure each event has well defined pre-event and post-event positions (currently the entire buffer is averaged, which ignores cases of multiple discrete static offset accumulations due to multiple earthquakes in rapid sequence). Furthermore, G-larmS depends on seismic detection which caused problems during aftershocks of the 2011 Tohoku earthquake [*Ohta et al.*, 2014], and may require implementation of rtGPS-based offset detection algorithms [e.g., *Allen and Ziv*, 2011; *Ohta et al.*, 2012]. A comparison of the real-time slip model to post-processed slip models, which are available in the hours to weeks after an event, must be very qualitative for two reasons: (1) the available data are inherently different: not all GPS in this region transmit data in realtime and post-processed GPS data are generally more precise (final orbit estimates are

available, multiple iterations for parameter estimation are possible); InSAR data are not available in real-time, and (2) real-time inversions operate under tight time-constraints, which limit the level of detail of parameter space exploration. Post-processed slip models (http://comcat.cr.usgs.gov/earthquakes/eventpage/nc72282711#scientific_finitefault, seismic: Doug Dreger, UC Berkeley; GPS+InSAR: William Barnhart, USGS) put most of the slip of up to 1-1.2 m on small regions up to 10 km north of the epicenter. While our final solutions also put the bulk of the slip north of the epicenter (Figure 3, middle panel), the fault patches are each 10 km long and Laplacian smoothing distributes the slip to the adjacent patches. Having the initial slip mostly south of the epicenter (Figure 2) is due to a bias in data distribution: early data are from that region. Due to the size of our patches (10 km long, 12 km wide), our peak slip of 6.1 cm is much smaller than those of post-processed models as it is integrated over a much larger area. Given our coarse discretization, our peak slip location gives a first order approximation of the more refined post-processed models.

6. Summary & Conclusions

We present rtGPS observations and real-time modeling results for the $M_w 6.0$ Napa earthquake that occurred on 24 August 2014 in Northern California. Almost all aspects of the G-larmS system running at the BSL worked as expected and produced finite fault slip models in real-time. A bug in the handling of time in the code caused a delay of 10s before the first results were produced (additionally to 6s data latency). Due to this the real-time system provided first results 24s after the origin time (8s is the estimated S-wave travel time). We show in a simulated real-time solution that this time can be

reduced to 14 s (S-wave wait time plus data latency, virtually no delay inside G-larmS). As ≈ 4 s of the 6 s latency are due to data buffering, we will explore in the near future how much this wait time can be reduced while still providing high data completeness.

The magnitudes produced in real-time and in replay mode capture the event well and a first-order distributed slip model is produced. When compared to background noise it is obvious that displacements induced by this event just barely stand out above the noise in this network. The model fit to the data provides an additional metric to automatically evaluate the rtGPS solution quality for $M_w \approx 6.0$ estimates. For the future, this means G-larmS will only publish solutions for events with a ShakeAlert magnitude greater than $M_w \approx 5.5$. Aggregator algorithms like the Decision Module should implement a magnitude threshold at which they consider an rtGPS contribution relevant.

The assumption of fixing the geometry to San Andreas Fault parallel worked well in this case, but work is required to parallelize the solution algorithm and test additional strike and slip orientations in real-time. If fully parallelized, this should not add much to the actual solution time as only picking of the best solution of many is required.

In summary, this was an excellent test for the G-larmS implementation at the BSL. The system ran stably, produced the first real-time finite fault slip distributions during an earthquake and was able to process multiple events in parallel. Observing the many aftershocks and investigating the solutions gives a clear picture on the rtGPS threshold, considering the station-event geometry, on the lower end of the magnitude spectrum in the Bay Area at $M_w \approx 6.0$.

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Figure 1. Network of high-rate GPS station baselines (dot color indicates operator) in the greater San Francisco Bay Area processed at the BSL. High-rate PBO stations that are not yet real-time capable (as of April 2014) are shown as white dots. Black star marks event epicenter, black circle represents G-larmS' baselines selection area for a M_w =5.7 event (78 km radius). OHLN is closest site. Baselines that delivered first offsets are colored red.



Figure 2. Real-time solution produced 26 s after the event origin time (first solution was at 24 s, event location given by black star). **(Left Panel)** Offsets with respect to site P256 (large red dot). Blue offsets are static horizontal offsets from rapid daily time series (courtesy of UNR). Red vectors give real-time offsets from $\approx 5 \text{ min}$ of pre-event data and 3 s of post-event data. Real-time data uncertainties are large and omitted. **(Middle Panel)** Model at 26 s after the event using offsets shown as red vectors in left panel. White to yellow colored baselines indicate model misfit. Projection of vertical fault is shown in map view. Pink colors indicate slip amplitude. N-S (left to right) fault cross section is at the bottom of the panel: vectors give rake (right lateral) normalized to maximum rake of the final solution. **(Right panels)** Top: Time series of GPS-based magnitude, black circle shows initial ShakeAlert magnitude; bottom four panels show north (blue) and east (black) displacement time series for bold, colored baselines in middle panel. Crosses mark offsets derived along these baselines (time shift between GPS solutions and offsets is due to 6 s data acquisition and processing latency). Supplement S1 and S2 animate the time series.



Figure 3. Same as Figure 2, but showing 'final' real-time solution 85 s past event origin. Large outlier pointing west is station MCCM (poor sky view). Supplement S1 and S2 animate the time

series.



Figure 4. Replay of rtGPS solutions in simulated real-time with corrected time handling. Snapshots at 16 and 78 s after the event origin time. Figure setup similar to Fig. 2 and 3. When started at predicted S-wave arrival time the impact of dynamic shaking on the slip model becomes obvious. Offset estimation and magnitudes stabilize after about 10 s.



Figure 5. G-larmS results for M_w =6.0 main shock processed in real-time (15178) and replayed (REPLAY), and 13 aftershocks identified by ShakeAlert ID. Each box includes the first 60 results G-larmS produced for each event. Upper panel shows range of inferred magnitudes. Lower panel shows misfit of slip model (weighted residual sum of squares). Line in each box is median, boxes extend from 25th to 75th percentile, whiskers cover 1.5 times interquartile range, outliers are plotted individually. Vertical arrow in WRSS panel for event 15200 indicates that misfit is large (median \approx 140 cm). ANSS (black) and ShakeAlert (grey) magnitudes for each event are given on the upper horizontal axis.