ACQUISITION OF CROSSWELL SEISMIC MONITORING DATA

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1. INTRODUCTION

Seismic velocity (both compressional and shear) is a fundamental property of materials, a property that varies with changes in conditions both external (stress, temperature) and internal (fluid saturation, crack density). Monitoring of changes in these external or internal conditions is a goal of geophysical investigations such as earthquake prediction (via stress change monitoring) and reservoir exploitation (via fluid saturation monitoring). Using an active source and fixed receivers, the simplest means of observing a time-dependent stress change is to measure the delay time between subsequent source pulses for the same path. The precision at which this measurement can be made determines the precision to which stress changes can be detected, and the interval between measurements determines the temporal resolution. Active source, continuous monitoring of seismic velocity can provide high-precision in situ measurement of temporal changes. Using a crosswell geometry for acquisition (i.e., having source and sensors, at depth, in separate boreholes) provides a methodology

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for spatially localizing the monitored region. Ideally, the source and sensors would both be in a formation of interest, and only changes in that formation would affect the data. This minimizes the effects of near-surface changes, which are often unwanted “noise”, and simplifies data analysis.

An important component of in situ monitoring with active-source seismic measurement is calibrating the change in the seismic P- or S-wave velocity with a known stress variation, such as earth tides or atmospheric pressure change. This would allow the identification of that part of seismic velocity changes that is due to stress change in a given rock volume, and allow an in situ determination of the stress-velocity relationship. Changes in seismic velocity induced by solid earth tidal loading and variations in barometric pressure have been estimated in some studies. The fractional change in seismic velocity with respect to stress change (i.e., the pressure sensitivity) is reported to be in the range of $10^{-5}$ Pa to $10^{-6}$ Pa (De Fazio et al., 1973; Yukutake et al., 1988; Sano et al., 1999).

Previous active source monitoring efforts had variable resolution of in situ velocity changes; however, recent work has shown dramatic improvement. The major advances, compared to the earlier studies, are twofold. First, the characteristics of available seismic sources have significantly improved, primarily in repeatability. Detection of travel-time changes is maximized for highly repeatable sources that can operate at high frequency and for extended periods of time. Previous work used sources such as surface vibrators or air guns, which were difficult to keep running for extended periods and had limited repeatability. Modern piezoelectric borehole sources are highly repeatable and dependable for millions of source excitations. Second, the precision in measuring differential travel time has increased, made possible in large part by the greatly increased sample rates available with modern high-speed data acquisition systems using high-speed internal clocks, and by the computational capability of performing massive waveform stacking. Additionally, the development of seismic sensors, both coil/magnet type geophones and piezoelectric-type hydrophones, has improved the sensitivity and dependability over recent decades. Studies with permanent installations have demonstrated the improvement in monitoring resolution.

A logical next step is to take continuous monitoring measurements to field sites and use equipment designed for temporary deployment in wells of opportunity, allowing in situ monitoring to assess the sensitivity of a given site and then redeployment when and if desired. In this paper, we present techniques used for in situ travel-time monitoring using shallow (10-100 m) field site boreholes, with equipment designed for crosswell seismic acquisition, recorded for time periods of 6 and 35 days at two separate sites. The analysis is focused on correlation of barometric pressure changes to velocity changes, allowing calibration of the stress sensitivity.
2. **Optimal Acquisition Parameters**

For a given acquisition geometry, we would like to optimize acquisition parameters such as frequency content and source-receiver distance, which tend to scale (i.e., lower frequencies at larger distances due to attenuation). In travel-time monitoring, we measure $\tau$, the lag time between an arriving wavelet (usually the first) on two seismograms, with a total travel-time $T$, and a center or characteristic frequency $f_0$. If we express both $\tau$ and $T$ in terms of the characteristic frequency, $f_0$, $\tau = \varepsilon/f_0$, $T = N/f_0$, where $N$ is the number of wavelengths between the source and receiver and $\varepsilon$ is a normalized time delay. Then the velocity perturbation, $\delta \hat{v} = \delta V/V$ can be simply rewritten as the ratio of two dimensionless parameters:

$$\delta \hat{v} = \frac{\varepsilon}{N}. \quad (1)$$

This particular parameterization provides a simple way of scaling experiments both in terms of frequency and spatial extent. From Eq. (1), it is clear that achieving the lowest possible detection threshold for $\delta \hat{v}$ can be broken into two separate tasks: (1) maximizing the number $N$ of wavelengths between the source and receiver, and (2) increasing the precision in measuring the normalized delay time, $\varepsilon$ (effectively the phase). The first task, maximizing $N$, is a property of the medium, the geometry, and the source characteristics, since $N$ can be written as $N = L f_0/V$. Thus, the acquisition geometry (path length = $L$) and maximum frequency (a function of source output and the medium’s attenuation) will factor into determining $N$. The second task is a function of frequency content and signal-to-noise. The attenuation, $Q$ (Toksoz and Johnston, 1981) will affect the signal-to-noise and therefore the precision. It can be shown that the optimal $N \sim Q/\pi$ (Silver et al., 2007).

For $Q = 60$, a typical upper crust value, we have $N \sim 20$, and assuming the $P$ velocity is 1.5 km/s, the optimal frequencies for source-receiver distances of 3 m, 30 m, 100 m and 1,000 m (corresponding to the distance range that we expect to encounter) are 10 kHz, 1.0 kHz, 330 Hz, and 33 Hz, respectively. To the extent possible, we use the optimal frequency in the data acquisition design. For our two test sites, we were able to use 10 kHz and 1 kHz at 3 and 30 m, respectively.

3. **Acquisition Hardware**

Data acquisition for continuous crosswell monitoring builds on development of crosswell seismic work of the previous 10–20 years, which was usually designed for tomographic imaging (e.g., Daley et al. (2004),
Majer et al. (1997), Rector III (1995)), but more recently has been used for time-lapse monitoring (e.g., Vasco (2004), Hoversten et al. (2003)). Three major data acquisition components are involved: seismic source, seismic sensor, and the recording system. Our acquisition was conducted with a combination of commercial and special-built equipment. The special-built components are the piezoelectric source and the high-voltage amplifier used to power it. The source is made up of piezoelectric-ceramic (lead zirconate titanate) cylindrical rings epoxied together and wired for positive and negative voltage on the inner and outer surfaces. This type of source has been used in many seismic crosswell surveys (e.g., Majer et al. (1997), Daley et al. (2004)) and is known to be repeatable and dependable. The tests reported here used a single-cycle square wave generated by a source waveform generator (in this case, a programmable analog signal generator), which also sends a trigger signal to the recording system. Other source waveforms are known to provide greater signal-to-noise ratio. The sensors were commercial hydrophones designed for large bandwidth and high sensitivity, using piezoelectric film and built-in amplifiers with analog data transmission to the surface recording system.

The commercial recording system, a “Geode” manufactured by Geometrics, has a 24-bit analog-to-digital converter using a sigma-delta conversion algorithm. Our data was collected at a sampling rate of 48,000 samples per second. The triggering of the recording system for each shot is an important consideration. The digitizer is continually sampling the data, and receives a trigger that will generally be between two digitized samples. Including a section of pre-trigger data, the time series is interpolated and resampled, so that the time series begins at the time of the trigger. This start time is not exact, and, at a sampling rate of 48,000/s, this time is computed to within a 20th of a sample (Geometrics engineering, personal communication). Thus, there is a delay-time measurement error that will be a 40th of a sample (halfway between samples), and the average error will be an 80th of a sample, assuming that the errors are uniformly distributed. This corresponds to an average error of 260 ns per trigger. The error in the stacked data is smaller by a factor of $N^{1/2}$, assuming they are uncorrelated. For $N = 36,000$ traces, as in the case of our experiments, the error is thus 1.4 ns for one-hour sampling, which is below other sources of error.

4. Field Tests

Our first test was conducted at a distance of about 3 m between two 15-m deep holes at the Lawrence Berkeley National Laboratory (LBNL). The piezoelectric source was placed at the bottom of one hole. The source power supply for this test was a commercial high-voltage pulser, made by Cober Electronics. The sensor cable had 24 sensors at 0.5 m spacing, spanning most
of the second hole. All 24 sensors were recorded. The source was pulsed every 100 ms with a record length of 10 ms, which yields 10 traces per sensor per second. These individual traces were stacked in the acquisition system, with a stacked set of 24 traces output every minute. A one-minute record is thus a stacked recording consisting of a real-time stack of 600 single traces for each sensor. An example of a single shot record (no stacking) for all 24 sensors is shown in Figure 1. The signal-to-noise ratio is very high even without any stacking, especially for those sensors in the middle of the receiver string. The sensors near the surface (channels 18–24) had a much lower signal level, because they were above the water table in a much more attenuative medium. One advantage of using multiple receivers is that we can select the channels to be analyzed to obtain $\varepsilon$ as low as possible.

In this 3 m experiment, we continuously recorded a total of 160 h starting from November 12, 2003. The SNRs average was $\sim4,000$ for the stacked one-minute records. Figure 2(a and b) shows that SNR for this experiment continues to increase for up to $10^4$ stacks, implying that a nonrandom noise “floor” has not been reached.

Figure 2(c) shows the histogram of the measured delay time between two adjacent one-minute records, which follows a normal distribution with a standard deviation of $\sim50$ ns. Since the P-wave velocity at the test site is $\sim1.5$ km/s, the travel time, $T$, is $\sim2$ ms. For one-hour stacked records, we can achieve a standard deviation, or precision, of up to 6 ns, which
corresponds to a resolution of $3 \times 10^{-6}$ on the velocity perturbation, $\delta \hat{v}$. Our actual precision is possibly better than this, since there is a contribution from actual stress-induced velocity perturbations in the histogram. This nanosecond precision is the key to successful stress monitoring.

In this experiment, we measured the delay time using two different time windows: a short time window, which includes the first arrival only (1.4 cycles), and a long time window, which contains both the first arrival and coda (10 cycles). The measured delay times with respect to the trace using the two time windows are plotted in Figure 3(a) and b. Both show a time variation of $\sim 3$ μs (for a travel time $\sim 2$ ms), corresponding to $\delta \hat{v}$ of $10^{-3}$. We also show the variations in barometric pressure (Figure 3(c)) and dilatation (Figure 3(d)) recorded by a nearby strain meter. During the period of observation, we see a large pressure excursion ($\sim 10^3$ Pa) that continues for 3 days (Figure 3(c)). The delay times closely track this barometric pressure change. Using the variations in delay time and the barometric pressure change, we obtain a value of stress sensitivity of $\sim 10^{-6}$/Pa. Cross correlations between delay time with barometric pressure look very similar to the autocorrelation of the delay time with itself, demonstrating that at least over a long period, there is an excellent correlation (Figure 3(e-g)).

The sign of the travel-time fluctuation relative to the barometric pressure (increasing travel time for increasing pressure) is the opposite of the standard expectation and requires some explanation. An increase in barometric pressure acting directly on the borehole will raise pore pressure and decrease
the effective stress in the near-field region next to the borehole. In the far field, the same increase in barometric pressure will act on unchanged pore pressure and increase the effective stress. The change in seismic velocity along a particular ray path thus depends on the net change in effective stress. When two boreholes are relatively close, near-field effective stress can decrease and cause a decrease of seismic velocity.
After the LBNL site work, we wanted to deploy in deeper wells and at a longer interwell distance, hopefully moving out of the near-well conditions at the LBNL site. We chose to deploy in wells at the Richmond (California) Field Station (RFS), a test site on San Francisco Bay near the University of California, Berkeley. The RFS test site has several boreholes approximately 70 m deep and 30 m apart from each other (Daley and Gritto, 2001).

At RFS, using equipment shown in Figure 4, the source waveform was a 1 ms wide pulse (half cycle of a square wave). The output waveform had 1.2 kV peak-to-peak amplitude. The source was repeated 10 times per second, with a 50 ms record length, and these pulses were stacked by the recording system to give recordings every 1 minute (600 pulses). For this study, we attempted to reduce any movement of the sensors with “centralizers” (flexible bands of metal acting like springs), which kept three of the sensors fixed, and with hard foam attached to the source. Figure 5 shows the sensors with centralizers and the source with foam. Unfortunately, at some point during deployment, the source foam was compressed, apparently by the water pressure.
The dominant frequency observed in the RFS data was about 1 kHz; the travel time was about 20 ms (a Vp of 1.5 km/s). Compared to the LBNL test, the data frequency is about 1/10, but the travel time is 10 times as large; therefore, we achieved a resolution for \( \delta\), essentially the same as the value we obtained at LBNL, \( \sim 3 \times 10^{-6} \). The signal-to-noise ratio of 1-minute stacks is 604, with a standard deviation of 39.2 ns, while 1-hour stacks have a SNR of 3185 with 5.2 ns standard deviation. We recorded 35 days of continuous data in the RFS test.

The RFS data, shown in Figure 6(a), required some initial preprocessing. Bad shots were deleted and the long-term trend was removed. The cause of the long-term trend is not known, but it has a period larger than our signals of interest. During the 30 m experiment, we found that ambient surface temperature (Figure 6(b)) had a measurable effect on the data. We observed that delay time correlates very strongly with temperature. Recording of temperature inside the recording container allowed correction for temperature variation. The scaling is about 0.1 s/°C (as shown in Figure 6(d)) where swings of 10 °C correspond to oscillations in delay time of about a microsecond.

This is a large error, which we treated in two ways. First, we can solve for a temperature correction coefficient and subtract out this effect, assuming it is linear. A linear scaling constant between the measured temperature and the delay time variation was estimated, and the scaled temperature data was then subtracted from the delay time data. This was effective in minimizing the 24-hour variation in delay time caused by temperature. Second, we can minimize the effect by maintaining the electronics within
Figure 6  (a) (top) Measured raw delay time in microseconds for the RFS experiment for first 25 days. Excursion from 14 to 20 days is during heavy rainfall. (b) Measured temperature, degrees centigrade, inside the recording container. (c) Measured barometric pressure, millibars. (d) Comparison of raw delay time (microseconds) and temperature for initial 250 h at RFS.

a temperature range, which was done midway through the experiment by installing thermostat-controlled heating and air conditioning in the recording container, which stabilized the temperature. The temperature control began on Day 17 and the minimization of temperature variation can be seen in Figure 6(b). We were successful in maintaining temperature variation within $\sim 2$ °C using standard commercial thermostats to control heating and cooling.

The data collection coincided with a period of heavy rain, and we found that the correlation to barometric pressure had changed following this rainfall event. The data from the rainfall period is difficult to interpret. At least two possible effects are involved, an increase in surface load (increasing effective stress) resulting from the mass of the rainwater, and an increase in pore pressure (decreasing effective stress) resulting from the infiltration of rain water. The effect on travel time due to rain is therefore difficult to interpret and we do not attempt to do so here.

Figure 7 shows the delay time and barometric pressure for two time periods, before and after the rain, corresponding to days 1-14 and 24-36. In the first time period, a delay time change of 1 microsecond approximately
corresponds to a barometric pressure change of 15 millibar, giving a stress sensitivity of about $5 \times 10^{-8}$/Pa. The second time period had a stress sensitivity of about $12.5 \times 10^{-8}$/Pa. Importantly, the sign of travel-time fluctuation relative to the barometric pressure is now the expected decrease in travel time for increasing pressure, indicating that we are measuring the “far field” effect away from the boreholes.

5. Conclusions

Modern crosswell data acquisition systems are capable of high-precision subsample delay time measurement. We have acquired two crosswell monitoring data sets over time periods of 7 days and 36 days. A large sensitivity to surface temperature in the recording instruments was observed and corrections were applied. In the second experiment, centralizers were used to stabilize the position of the source and sensors. In these two separate experiments, we demonstrated travel-time measurement precision on the order of $10^{-7}$ to $10^{-8}$ s. This precision allows us to monitor barometric pressure as a calibration signal to measure the stress sensitivity of velocity at each field site. Near-field effects dominated our first test at 3 m spacing between wells. These effects caused the velocity to decrease with increasing pressure, with a sensitivity of $10^{-6}$/Pa. In a larger-scale 30-m experiment, the velocity was observed to increase with increasing barometric pressure, with a sensitivity of 5-12.5 $\times 10^{-8}$/Pa.

Addendum

Since the completion of this manuscript, we have successfully deployed a crosswell stress monitoring experiment at 1 km depth, for multiple weeks.
in two deployments, using the San Andreas Fault Observatory at Depth (SAFOD) site boreholes. As well as measuring the stress sensitivity, pre- and coseismic stress changes were observed. This work is described in Niu et al. (2008), and the reader is referred to this recent experiment. In 2009, our coauthor Paul Silver died unexpectedly. We would like to acknowledge his friendship and leadership in this work.

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