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Monitoring of fields using body and surface waves reconstructed from passive seismic ambient noise
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Summary
There are important economic, environmental and societal reasons for monitoring production from oil, gas and geothermal fields. Unfortunately, standard microseismic monitoring is often not useful due to low levels of microseismicity. We propose to use body and surface waves reconstructed from ambient seismic noise for such monitoring. In this work, we use seismic data recorded from a dense sensor array at the Groningen gas field in northern Holland and show how direct P-waves can be extracted from the ambient noise cross correlations and then used to monitor seismic velocity variations over time. This approach has advantages over the use of coda wave interferometry due to the ability to localise such changes in the subsurface. We show how both direct and refracted (head) P-waves as well as Rayleigh surface waves can be used for such field monitoring, with changes of ~1% being resolved. Both fundamental and first overtone Rayleigh waves are used to localise such changes, which correspond nicely to known geology to within 100 m.

Introduction
Monitoring of oil, gas and geothermal fields during production is important for economic and societal reasons. While microseismic monitoring is widely used for monitoring of the stimulations and early production [Shapiro, 2008], the typical drop in recordable microseismicity afterwards reduces the value of this technique [Shoenball et al, 2014]. In particular, the aseismic deformations that can indicate potentially unwanted leakage and in some cases corresponding contamination of aquifers are difficult to monitor [Rutqvist, 2012].

Noise-based passive seismic methods can potentially add much value for reservoir monitoring. These methods are based on the proportionality of the Green’s function with the cross-correlation noise seismograms recorded in a diffuse seismic wavefield [Lobkis and Weaver, 2001; Shapiro and Campillo, 2004; Roux et al, 2005; Campillo, 2006; Snieder, 2007].

Different studies have showed that body-wave extraction from noise correlations is possible at various scales (Roux et al. 2005, Draganov et al. 2009, Poli et al. 2012). Nakata et al. (2015) were able to implement the first passive 3-D P-wave velocity tomography from continuous ground motion recorded on a dense array of more than 2500 seismic sensors installed at Long Beach (California, USA). Recently Brenguier et al. (2016) and Nakata et al. (2016) showed the temporal stability of direct virtual body-waves between dense arrays on Piton de la Fournaise volcano thus opening the way for continuous, passive ballistic wave monitoring.

Method
The new approach is based on measuring temporal changes of apparent slowness of specific ballistic waves that have been reconstructed from noise correlations using dense arrays of seismic sensors (Boué et al., 2013, Mordret et al. 2014, Nakata et al., 2015, Nakata et al., 2016). The underlying requirement is for a high number of seismic sensors (> 100) and if satisfied the noise correlation receiver pairs allow for the reconstruction of a virtual shot-gather of sufficiently high quality to be able to isolate and measure the apparent velocity of ballistic waves such as direct P or S-waves, refracted waves or surface waves with clear mode separation.

In this study, we gather all possible noise cross-correlations from every sensor pair in a dense array into a single data set – propagation time vs virtual source offset – and thus assume a 1D horizontal velocity model. Then we measure the arrival time variations of identified body wave types at each offset, and use a linear regression to measure the velocity variation across the array for each time interval. In this way time lapse monitoring is achieved for each identified body wave, as illustrated in Figure 1.

Data
The Groningen gas field located in the northeast of the Netherlands is one of Europe’s largest natural gas fields. The reservoir is located at 3 km depth and is thought to be 40 by 50 km wide and 250 m thick. It is sealed by an overlying Zechstein salt layer up to 1 km thick. Above the
salt layer lies a ~1 km thick Cretaceous Chalk formation capped with a 800 m thick Tertiary and Quaternary sediment cover, up to the surface (van Thienen-Visser & Breunese 2015).

We are using continuous seismic data recorded by a network of 417 short period triaxial (3C) geophones deployed in the Groningen area (Figure 2) from 11 February (day 42) to 12 March (day 71) 2017. The network forms a grid array with aperture of the order of 8 km and typical inter-sensor distance of about 300 m.

Direct and Refracted P-waves

After the cross correlation of the vertical components of the seismograms and the binning procedure previously outlined, we obtain images shown in Figure 3 for different frequency bands. The dominant waves present, particularly at low frequency, are the surface waves but P-waves are also clearly present – especially at higher frequencies. The P-waves are present on the vertical components due to the velocity gradient near to surface which causes the P-waves to travel horizontally as a series of ‘smile’ curves with vertices from reflections off the free surface.

The high frequency (3-12 Hz) window reveals both ‘direct’ P-waves travelling at 1700 m/s (which is roughly the known P-wave velocity in the top 700m layer) and refracted P-waves travelling with a velocity of about 3000 m/s. We interpret this refracted wave as a head wave from the P-waves travelling beneath the interface, and thus this observed velocity is the P-wave velocity of that layer.

Having identified the inter-station distance window where the direct and the refracted P-waves are clearest, we build a reference image by stacking the virtual gathers from days 42-50. Then we use a sliding 10-day window to make similar images and compare against the reference to extract velocity changes for each inter-station distance. A linear fit to the velocity changes at each inter-station distance yields the velocity change at that time. This procedure is repeated for both direct and refracted P-wave: the images for the refracted wave is shown in Figure 4, and resulting velocity change over time for both P-waves is shown in Figure 5.
As previously mentioned, the cross correlated waveforms exhibit strong surface (Rayleigh) waves on the vertical components. Actually, both the fundamental and first overtones are clearly visible (Figure 6) and thus used in this analysis, after a suitable F-K filter is used to separate them.

As before, a reference cross-correlation is used to compare against the daily cross-correlations, and then used to extract phase velocity variations for each day as functions of inter-sensor distance and frequency. As we use all sensor-pairs, we are implicitly assuming a 1D velocity model. A linear fit is applied to the changes as a function of inter-sensor distance, in a window of between 3 and 7 wavelengths to allow the surface waves to fully develop and to avoid high distances where data is sparse and measurement less reliable. In this way we obtain surface wave changes for each frequency band for each day.

We then make a depth inversion for each day, using the frequency-dependent changes and both fundamental and first overtone data. It’s worth noting that the sensitivity kernels (Figure 7) for these frequency bands show different depth dependencies and so the first overtone data
Ambient Noise Passive Seismic Monitoring

significantly improve the inversion. After the inversions, we have a 1D S-wave velocity model for each day, which we present in Figure 8.

Observations

The direct and refracted body waves (Figure 5) show that while the near-surface P-wave velocity is not significantly changing in the 30 day recording period, the velocity of the Chalk layer beneath 800 m depth does increase by +1.5% around day 54-56. This level of change is more than double the measurement error. After day 55, the Rayleigh wave analysis indicates a significant decrease of around 1.5% in the zone below 800 m depth. It should be noted that these two methods are quite different, using the same seismic noise data but different waves and processing. With that in mind, the common timing and location of the indicated subsurface changes is remarkable.

The recorded noise characteristics (directionality and spectrum) are stable in the 30-day period, ruling this out as a cause of the indicated velocity change. INSAR and GPS data do not suggest any causes, and neither does gas production data and modelling - these induced pressure variations are of the order of less than 0.1 MPa locally and are thus too small to potentially induce a 1.5 % velocity increase 2.3 km above in the carbonate layer. We also rule out the effects of local induced earthquakes due to the low level of seismicity during the studied time period. The only significant weather event was heavy rain (40 mm) over days 54-56, and so this is suggestive as the origin of the surface changes (P-wave +1.5%, S-wave -1.5%) at 800 m depth in the days following it.

Conclusions

We have demonstrated how ambient seismic noise – vibrations from trains, cars, ocean, etc. – recorded by a dense array of surface sensors can be used to reconstruct both body waves and surface waves, which can then be used to monitor subsurface changes in a gas (or oil or geothermal) field during production. These results suggest a new, cost-effective way of monitoring such fields.

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