Single station monitoring of volcanoes using seismic ambient noise

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1 Abstract

Seismic ambient noise cross correlation is increasingly used to monitor volcanic activity. However, this method is usually limited to volcanoes equipped with large and dense networks of broadband stations. The single station approach may provide a powerful and reliable alternative to the classical “cross-stations” approach when measuring variation of seismic velocities. We implemented it on the Piton de la Fournaise in Reunion Island, a very active volcano with a remarkable multi-disciplinary continuous monitoring. Over the past decade, this volcano was increasingly studied using the traditional cross-correlation technique and therefore represents a unique laboratory to validate our approach. Our results, tested on stations located up to 3.5 km from the eruptive site, performed as well as the classical approach to detect the volcanic eruption in the 1-2 Hz frequency band. This opens new perspectives to successfully forecast volcanic activity at volcanoes equipped with a single 3-component seismometer.

Key points:

- Volcanic monitoring with seismic noise cross correlation techniques can be achieved with a single three component seismic station
- The Single station Cross component approach provides more stable results than Auto correlation
- Environmental perturbations have a different signature from volcanic perturbations and can therefore be discriminated
2 Introduction

Volcanoes are studied and monitored using diverse instrument networks. Data recorded by the instruments are generally turned into proxies to assess the state of the volcano. The ultimate goal of volcano observatories is to identify changes in a volcanic edifice as far ahead in time of an eruption as possible. The efficiency of this monitoring is critically related to the detection and understanding of slight changes in the edifice long before the start of the magma transport during active volcanic episodes. During volcanic eruptions, magma transport causes gas release, pressure perturbations in the plumbing system [Patane, 2006] and potential surface deformation that can be detected using geodetic techniques [e.g., Peltier et al., 2009, 2016; Staudacher et al., 2009]. However, the sensitivity of these techniques to deep changes can be limited [e.g., Chaussard et al., 2013], leaving room for a better early warning solution.

Alternatively, deep mechanical processes associated with magma pressurization and/or migration and their spatial-temporal evolution can be monitored with volcanic seismicity and yield precise locations and mechanisms for earthquakes and volcanic tremor [e.g., Battaglia et al., 2005; Massin et al., 2011; Lengliné et al., 2016]. Yet, seismicity only provides information on short-term phenomenon (few-seconds to few days) and is inadequate to expose early aseismic processes such as magma pressurisation [Brenquier et al., 2008b].

Seismic interferometry uses the multiple scattering of seismic vibrations by heterogeneities in the crust. Implemented on coda waves of earthquakes or seismic ambient noise, this technique allows to retrieves the Green’s function for surface waves between two stations by cross-correlating these diffuse wavefields [e.g., Lobkis and Weaver, 2001; Derode et al., 2003; Snieder, 2004; Wapenaar, 2004]. This technique is increasingly used as a non-destructive way to continuously monitor small seismic velocity changes (~0.1%) associated with variations of heat, pressure, or water saturation in the sub-surface [Grêt et al., 2006; Sens-Schönfelder and Wegler, 2006]. Seismic velocity changes are typically measured from the cross correlation functions (CCF) for each pair of stations and eventually averaged over the whole network to yield more stable results. The Cross Correlation (CC) technique has been extensively described over the past decade, with many available reviews [e.g., Larose et al., 2006; Bensen et al., 2007; Wapenaar et al., 2010a, 2010b; Sens-Schönfelder and Wegler, 2011]. Seismic velocity variation measurements using the CC technique are used for monitoring [Snieder and Hagerty, 2004; Brenquier et al., 2008b; Hadziioannou et al., 2011] with applications for volcanoes [e.g., Sens-Schönfelder and Wegler, 2006; Duputel et al., 2009; Mordret et al., 2010; Caudron et al., 2015], large magnitude earthquakes in the far field [Wegler and Sens-Schönfelder,
2007; Brenguier et al., 2008a; Ohmi et al., 2008; Wegler et al., 2009; Hobiger et al., 2012; Minato et al., 2012] and smaller magnitude earthquakes at smaller distances [Maeda et al., 2010; D’Hour et al., 2015]. In most cases, relative velocity changes have been evidenced using a large number of stations and of station pairs. The technique is also applicable to single stations, using one (or all) components of a one- (or three) component seismometers. The single station approach has been successfully applied to study earthquakes [Sens-Schönfelder and Wegler, 2006; Wegler and Sens-Schönfelder, 2007; Hobiger et al., 2014; Nakahara, 2014; D’Hour et al., 2015]

In this study, we apply this approach to volcano monitoring using a small network. In addition, we explore the possibility to exploit higher frequencies than traditional ambient noise monitoring approaches and eventually opening the path to systematically make use of short-period seismometers for seismic velocity monitoring on volcanoes. We focus on the Piton de la Fournaise (PdF), a basaltic shield volcano located in the ESE part of Reunion Island (France) in the Indian Ocean. It is one of the world’s most active volcanoes with, on average, one eruption every year. The PdF is extensively monitored with a broad range of instruments and many studies use the seismic noise cross correlation method with multiple stations to measure seismic velocity changes associated to its activity [Brenguier et al., 2008b, 2011; Duputel et al., 2009; Clarke et al., 2013; Rivet et al., 2014, 2015; Sens-Schönfelder et al., 2014]. We studied data acquired by three seismic stations in 2014, containing one eruption in June and several environmental perturbations recorded throughout the year. This background provides a consistent base of comparison to test the robustness of the single station approach.

3 Data and method

3.1 Data

We used seismic data from the Piton de la Fournaise Volcano Observatory Network (OVPF, Institut de Physique du Globe de Paris), Reunion island, acquired in 2014 (Figure 1). At that time, the monitoring network of the PdF was composed of 25 continuously recording stations, i.e. 8 short-period and 17 broadband. This dataset covers a 1-day eruption on 21 June 2014 that followed 4 years of quiescence. We focused our analysis on broadband stations CSS, FJS and FOR, which are at similar distances from the crater Dolomieu, the main crater of the PdF. Among them, station FOR is the closest to the 2014 fissure eruption (Figure 1). In parallel, we analysed a catalogue of earthquakes located manually along with the rainfall measured at a meteorological station next to station FOR.
3.2 Method

Seismic velocity changes are measured from seismic noise cross correlation following a workflow similar to Lecocq et al. 2014: Seismic records for all components are pre-processed by carefully checking for their timing (sample alignment), gaps (interpolating or tapering between gaps), then bandpass pre-filtered between 0.01 and 8.0 Hz and resampled to 20 Hz prior to whitening and cross-correlation.

Traditionally, ambient seismic noise is cross correlated between pairs of stations. This approach was adopted in an earlier study by Rivet et al. (2015) and is compared with our results in section 4.4. In contrast, we performed two types of processing based on the single station approach: Single station Cross component (SC), and each component with itself or Auto Correlation (AC). The spectral whitening that sets the amplitude of the signal to 1 for all frequencies was not applied for AC since only the phase of the signal would remain and the auto-correlation of such a signal does not contain information on the medium anymore. For both AC and SC the data was then filtered in different frequency ranges (0.01-1.0Hz, 0.5-1.0Hz, 1.0-2.0Hz and 2.0-4.0Hz) and the performance of 1-bit normalisation as well as clipping at 3 times the RMS of each time window were tested as time domain normalisation. Clipping eventually provided the most stable results and was therefore chosen for this study.

The cross and auto correlations were then computed for each individual day, between all the possible combinations of components (3 SC and 3 AC for each station) and for all the different filters, before being averaged with a 5-days linear stacking to maximise the signal-to-noise ratio (Figure 2).

The plot of CCFs over time shows coherent phases in the late part of the coda for different days (e.g., +10 seconds time lag). Shifts in time from these coherent phases, even if not visible on Figure 2, are interpreted as changes of seismic velocity in the crust that can be measured. With the assumption of a homogeneous change in the medium, we considered that the relative differences in travel time $dt$ are due to the change in the seismic velocity $dv$ as $-dt/t = dv/v$ [Ratdomopurbo and Poupinet, 1995].

Temporal velocity variations in the medium are measured both on the negative and the positive sides of the CCF, for time lags between 5 and 35 seconds, preventing direct wave contamination. Each individual CCF (daily) is compared to a reference CCF that averages the results for the whole period of study. The travel time changes are measured in the frequency domain using the Multiple Window Cross Spectral Analysis method with a quality control using coherency and the error of the linear regression in the time domain [Poupinet et al., 1984; Brenguier et al., 2011]. The velocity variations calculated with the SC and the AC are then ultimately averaged by station for each frequency band.
4 Results and discussion

4.1 Frequency dependence

Comparing $dv/v$ curves for station FOR in Figure 3 shows that the SC (A.) is significantly more stable than the AC (B.) in the lowest frequencies. The AC exhibits a very strong variability with a higher signal to noise ratio starting from the 0.5-1.0 Hz frequency band. The SC already shows a distinct pre-eruptive seismic velocity drop in the 0.1-1.0 frequency band (A.). In the 1.0-2.0 Hz frequency band the SC (E.) is sensitive to both heavy rainfall events and the volcanic eruption (blue and red vertical lines, respectively) with a larger drop before the eruption (~0.3%, station FOR) than for the rainfall (~0.15%, station FOR). The AC (F.) is also sensitive to rainfall and the volcanic eruption in that frequency band. However, a larger amplitude perturbation affects the observation between 10 and 20 May and in early July. This perturbation is consistent for all three stations and seems to have little effect on the SC processing. No rainfall or seismic activity (blue and yellow vertical lines, respectively) seem to correspond to this seismic velocity change. The contamination of the AC by high amplitude events such as earthquakes is a known issue since no spectral whitening can be used to mitigate them. Here, the perturbation was probably caused by storms and very strong winds which were recorded those days [Meteo - France Direction Interrégionale Océan Indien, 2014] and are known to affect high frequencies [Withers et al., 1996]. Differences between AC and SC are again more pronounced in the 2.0-4.0 Hz frequency band with an increased variability for the SC (G.) opposed to the AC (H.), which exhibits less variability but also a smaller signal-to-noise ratio.

The SC clearly provides better results than AC in terms of stability and clarity for the pre-eruptive decrease of seismic velocity. For all three stations, the pre-eruptive velocity drop is clearly detected. The best results are obtained in the 0.5-1 Hz and the 1-2 Hz frequency bands (C., E.), with the highest signal to noise ratio being reached in the 1-2 Hz frequency band. The best performance of these frequency bands could be caused by a higher power at these frequencies of the ambient noise field (Figure S3).

4.2 Seismicity

Figure 4 shows our results (A. and B) along with the rainfall (C.) and the seismic activity (D.). For SC (A.) and AC (B.) the 1.0-2.0 Hz and 0.5-1.0 Hz frequency bands respectfully show the best signal to noise ratio and sensitivity to the noteworthy events that occurred during the studied period. Considerable rainfall (blue vertical lines, the value exceeds 3 times the RMS) caused a drop in seismic velocity of similar magnitude at all three stations which are always observed, except in April with the
AC approach. Another drop precedes the 21 June eruption (red vertical line) and could indicate a pre-eruptive decrease in seismic velocity.

Before 9 June, the seismicity remained at a very low level with less than 30 earthquakes per day, a vast majority of which were identified as rockfalls. On 11 June the drop in seismic velocity seems to have begun independently from the number of earthquakes (less than 5/hour) and their magnitudes. The seismicity eventually increased due to summit activity and as the seismic velocity dropped, reached a peak the day before the eruption (955 earthquakes, 846 from summit activity). After the eruption, the seismicity became very low, dominated by rockfalls again (Figure 4, S1).

A large number of volcano-tectonic earthquakes at PdF volcano were relocated by Lengliné et al. [2016] for 2014 and 2015. They identified a persistent shallow (~700 m above sea level) pre-eruptive ring shaped cluster under the summit crater associated with a westward migration of its southern part before the eruption. They interpreted the repetitive occurrence of earthquakes along this structure as possible pre-existing zones of weakness within the edifice that are triggered by static stress changes linked to an over-pressurization of the magma chamber or dike intrusions in the volcanic edifice. The same process likely caused the pre-eruptive seismic velocity drop, which was therefore concomitant to the increase of seismic activity. Although we interpreted the pre-eruptive velocity drop as a result of the build-up of pressure in the subsurface, we cannot completely preclude an influence of the simultaneous increase of seismic activity or migration of earthquakes.

The seismicity could ultimately contaminate our observation in the microseismic frequency band, but only affecting direct waves when we are looking further in the coda. It could also be assumed that high seismicity is required to perform good observations of seismic velocity variation with this technique. However, lower seismicity did not prevent us from observing rainfall-associated velocity drops (blue vertical lines, Figure 4), ruling out this assumption.

4.3 Station by station comparison

The station geographically closest to the fissure eruption (FOR) clearly displays a significantly larger drop (~0.3%) than the two others (<0.1%) for both processing approaches. By contrast, the three stations display the same amplitude for the rainfall-associated drop (0.05-0.1%). We interpreted this as a larger scale distribution of rainfall which, unlike the volcanic eruption, affects the three stations in the same way. This noteworthy difference could provide a way to discriminate the causes of seismic velocity change when the single station approach is used.
4.4 Comparison with Rivet et al. [2015]

Rivet et al. [2015] studied the temporal variation of seismic velocity using seismic noise correlation in the traditional CC approach with all the 27 stations of the OVPF network. They used only vertical components band-pass filtered between 0.25 and 2 Hz combined with 1-bit normalisation, 8-days stacking, “long-term variation removal” (sic.) and network-wide averaging to observe the pre-eruptive drop of seismic velocity. They extended their observation to the preceding low activity period and highlighted seismic velocity variations highly correlated with rainfall episodes and subsequent pore pressure perturbation.

Figure 4 shows the results from Rivet et al. [2015], before and after they were corrected for the rainfall effect, along with our results for SC and AC processing. Their results unsurprisingly do not seem to exhibit the same temporal resolution as our results because they averaged the pairs from all the network stations, used 8-days stacking and possibly also some extra fitting/sliding mean [Brenguier et al., 2008b; Clarke et al., 2013].

There is a striking correlation between their results obtained with the CC and all the pairs from the 27 stations network and our results obtained with one station at a distance up to 3.6 km from the eruptive vent. The single station approach clearly appears as a promising alternative when the CC cannot be efficiently implemented. These unfavourable scenarios include volcanoes equipped with only one or too few seismic stations as well as analysis where the CC provides across-correlation coefficient between the CCFs and the reference that is too low. Additionally, short period instruments are not always used for cross-correlation analysis due to strong attenuation of high frequencies between far apart station pairs that affect the measurement of seismic velocity variations. The frequency bands used in our analysis are dominated by microseismic energy and would therefore likely work everywhere for seismic velocity monitoring, including with short period instruments.

5 Conclusions

Like the traditional Cross Correlation approach implemented by Rivet et al. [2015], both the Single station Cross component (SC) and the Auto Correlation (AC) approaches successfully detected the pre-eruptive seismic velocity drop along with other extreme climatic perturbations in 2014. The good performance of the single station approaches opens the possibility to use noise cross-correlation techniques on volcanoes equipped with only one of too few instruments, or poorly correlated station pairs.
The AC exhibits poorer results than the SC in terms of stability, a lower pre-eruptive velocity drop and a sensitivity to strong amplitude events that the SC does not have. More work is still necessary to better mitigate the contamination of strong amplitude events on the AC with solutions such as using phase cross-correlation over the classical cross-correlation [Schimmel, 1999; Schimmel and Gallart, 2007; Schimmel et al., 2011; D’Hour et al., 2015] which will be left to a future study. The best performance for the SC and the AC are obtained in the 1-2 Hz and the 0.5-1 Hz frequency bands, respectively. The good performance at high frequencies could be associated to the higher amplitude of those frequencies in the ambient noise content that “illuminates” the change in the medium, providing a clear, stable observation of the velocity drop. These results also show that short period seismometers could probably be used with the single station approach.

The volcanic eruption and the rainfall have a different effect on the seismic velocity measured at distinct stations. The rainfall has a similar impact on all the stations while the volcanic eruption has a greater effect on the closest station. Still, it should be noted that even the most distant station (station CSS, 3.65 km from the eruptive site) clearly detected the pre-eruptive velocity drop using both the SC and the AC. These results open new perspective to monitor volcanoes using seismic velocity variations where the traditional cross-correlation analysis cannot be performed.

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7 References


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Figure 1: Map of the stations used for this study (red triangles) and the OVPF seismic network (triangles). The yellow triangles represent Short period seismometers, the remaining are broadband seismometers. Station FOR is closer to the 2014 fissure eruption (circle).
Figure 2: Plot of the Cross correlation functions (CCF) with time for station FOR, components pair ZE and 5-day stack. The data was filtered between 1 and 2 Hz and clipped in the time domain at 3 times the standard deviation. Stable phases can clearly be identified up to about 30 sec of lag time.
Figure 3. Variation of seismic velocity for 3 stations (CSS, FJS, FOR) and 4 frequency ranges (from top to bottom, 0.1-1.0 Hz, 0.5-1.0 Hz, 1.0-2.0 Hz and 2.0-4.0 Hz) measured using Single station Cross components (SC, left) and Auto Correlation (AC, right). The vertical lines represent high rainfalls (blue), increasing seismicity (yellow) and the eruption day (red). The 95% confidence limits are shown in Figure S2.
Figure 4. Seismic velocity variations calculated using Single station Cross components (A., SC, 1.0-2.0 Hz) and Auto correlation (B., AC, 0.5 – 1.0 Hz) at Piton de la Fournaise volcano (5-days stacking) along with the rainfall (C.) and the seismic activity (D.) between March and July 2014. The vertical lines represent high rainfalls, increasing seismicity and the day of the eruption in blue, yellow and red, respectively. Results digitized from Rivet et al. 2015 are shown for comparison.