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Temporal change of near-surface shear wave velocity associated with rainfall in Northeast Honshu, Japan

Yu Miao, Yang Shi and Su-Yang Wang^{*}

Abstract

We apply seismic interferometry to earthquake records to investigate temporal change of near-surface shear wave velocity associated with rainfall in Northeast Honshu, Japan. We observe there exists a threshold (around 100 mm of 24 h rainfall total) of the temporal influence of rainfall on near-surface shear wave velocity. The velocity does not change significantly under low-intensity rainfall, while it is reduced by 1–3% with one exception of 10% when the rainfall intensity exceeds the threshold. Compared with the travel time of no rain, the travel time increment of low-intensity rainfall is no more than 3×10^{-4} s, while that of high-intensity rainfall ranges between 10^{-3} and 10^{-2} s. Based on the results of observation, we propose a one-dimensional model to compute the travel time increment, which can give a preliminary estimation of the temporal influence of rainfall on near-surface shear wave velocity.

Keywords: Earthquake ground motions, Seismic interferometry, KiK-net, Rainfall

Introduction

Near-surface shear wave velocity has been recognized as an important mechanical parameter in the fields of earthquake engineering and geophysics. It has a wide range of application such as classifying site classes (Borcherdt 1994), evaluating liquefaction potential (Ahmadi and Paydar 2014) and estimating local site response (Choi and Stewart 2005) for seismic design of buildings, structures, and underground utilities (Lee et al. 2014). And its relevance to evolving stress and effective rheology at depth for seismic monitoring of faults (Hillers and Campillo 2016; Wegler and Sens-Schönfelder 2010), volcanoes (Brenguier et al. 2014; Wegler et al. 2006), landslides (Mainsant et al. 2012; Pilz et al. 2014; Xu et al. 2017) and hydrocarbon reservoirs (Bakulin and Calvert 2006; Miyazawa et al. 2008) is well documented.

Hardin and Black (1968) expressed the shear wave modulus of soil as a function of eleven factors, mainly including effective stress, void ratio, saturation, strain history, temperature, grain properties and soil structure.

*Correspondence: suyangwang@whu.edu.cn

Huazhong University of Science and Technology, Luoyu Road 1037, Wuhan. China And the shear wave velocity of rock can be influenced by porosity, fluid saturation, fractures, pressure and temperature (Toksoz et al. 1976). Hence the near-surface shear wave velocity is subjected to a variety of environment conditions, such as crustal stress, ground surface perturbation and atmospheric forcing. Many researchers have observed change of near-surface velocity caused by strong motion (Hobiger et al. 2012; Nakata and Snieder 2011; Niu et al. 2008), plate motion (Nakata and Snieder 2012a; Takagi and Okada 2012), volcanic eruption (Brenguier et al. 2008, 2011), earth tide (Hillers et al. 2015a, b), seasonal variations, including temperature, precipitation, groundwater level; sea level, wind and so on (Clements and Denolle 2018; Meier et al. 2010; Richter et al. 2014; Sen-Schönfelder and Wegler 2006; Tsai 2011; Wang et al. 2017).

Seismic interferometry is a powerful noninvasive tool to image the structure of shallow crust from earthquake records or ambient noise (Hillers et al. 2015a; Mehta et al. 2007; Obermann et al. 2013; Yamada et al. 2010). It is a technique based on the correlation of signals recorded at different sensors (or stations). For seismic vertical arrays, seismic interferometry is used to estimate the Green's function between two seismic sensors of vertical arrays



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and extract the travel time of seismic wave propagating between them. The near-surface shear wave velocity can be estimated by dividing the distance between two seismic sensors (borehole depth in this study) by the travel time. In most cases, the borehole depth ranges from tens to hundreds of meters (only few larger than 1000 m), while the shear wave velocity changes within 6–8 km in the upper crust are considered by applying interferometry to station pairs at the surface. Hence, compared with applying interferometry to station pairs, applying interferometry to vertical arrays can obtain estimations of shear wave velocity with much higher depth resolution.

Many researchers (Hillers et al. 2014; Ohmi et al. 2008; Rivet et al. 2015; Sen-Schönfelder and Wegler 2006; Tsai 2011; Voisin et al. 2017) used cross-correlations of longterm ambient noise to investigate the seasonal change of average shear wave velocity in upper crust (free surface to several kilometers depth) caused by precipitation, and their results can be interpreted by hydrological models of crustal fluid pore pressure in upper crust or groundwater level. In contrast, in this paper, we focused on the temporal change of near-surface shear wave velocity associated with rainfall by applying seismic interferometry to strong motion data of seismic vertical arrays. The near-surface shear wave velocity is estimated between ground surface and borehole bottom with a depth of 100-200 m, and the temporal change occurs in the period that rainwater is penetrating through the top meters of the near surface right after the rain. For above differences, the study of this paper is necessary.

In this study, we apply seismic interferometry to KiKnet data to investigate the temporal change of near-surface shear wave velocity associated with rainfall based on earthquake and weather logging data in Northeast Honshu. Then we present the relationship between velocity temporal change and rainfall, and discuss the dependency of corresponding travel time increment on rainfall. Finally, we propose a one-dimensional model to interpret the temporal influence of rainfall on near-surface shear wave velocity.

Data and method

Strong motion and meteorologic observation net in Japan

The Kiban-Kyoshin Network (KiK-Net) is composed of 697 vertical arrays with an uphole/downhole pair of strong motion seismometers. These arrays were built by the National Research Institute for Earth Science and Disaster Resilience (NIED) after the 1995 great Hanshin-Awaji earthquake. The depth of most boreholes range between 100 and 200 m, and the depth of boreholes for 27 stations which are located on thick sediment is greater than 1000 m. The sampling frequency of accelerograms was 200 Hz initially and has been changed to 100 Hz since 2008. The total length of the accelerograms is 60–300 s which includes 15 s pre-trigger data (Aoi et al. 2004; Okada et al. 2004).

In Japan, a nationwide meteorologic observation network, which consists of more than 1400 weather stations, has been operated by Japan Meteorological Agency (JMA) for over 30 years. Air temperature, wind direction and speed, precipitation, sunshine duration and snow depth are automatically recorded every 10 min at a majority of these weather stations.

Rainfall intensity classification

In this study, the rainfall totals within 24 h recorded at the nearest weather station of the seismic station before the earthquake event were chosen to be the representative of intensity of rainfall, while hourly totals are more commonly used to quantify rainfall intensity in many regions (including Japan, Europe and America). Comparing with hourly totals, 24 h totals have a more reasonable consideration of the process of rainfall infiltration in the top meters of the near surface to investigate the temporal influence on shear wave velocity. We also tested different accumulated periods of rainfalls (from 24 h to 2 weeks) and found that rainfall amounts accumulated over 24 h (i.e., 24 h totals) could balance the stability and resolution of the seismic velocity changes well (Additional file 1: Fig. S1).

According to rainfall intensity classification from China Meteorological Administration (CMA), 24 h rainfall totals are linearly binned into six classes in this study. The six rainfall classes and their corresponding CMA rainfall classes are given in Table 1. As we introduced before, the near-surface shear wave velocity is subjected to a variety of conditions (such as groundwater level, incidence angles of incoming waves, temperature, earth tide and so on). Comparing with a scatter plot of the single measurements of seismic velocity versus the rainfall intensity, the rainfall intensity classification is designed for producing an averaging estimation of the seismic velocity to

Table 1 Rainfallclassesusedinthisstudyand corresponding CMA rainfall classes

24 h rainfall totals (mm)	Rainfall classes			
	This study	CMA No rain		
0	Level I			
0.1–25	Level II	Light to moderate rain		
25–50	Level III	Heavy rain		
50–75	Level IV	Torrential rain		
75–100	Level V	Heavy torrential rain		
≥100	Level VI	Downpour		

Station	Seismic stations				Weather stations		
	Lat.	Lon.	Alt. (m)	Depth (m)	Lat.	Lon.	Alt. (m)
FKSH14	37.03	140.97	3	147	37.07	140.88	12
IBRH11	36.37	140.14	67	103	36.40	140.24	72
IBRH13	36.80	140.58	505	100	36.74	140.59	395
IBRH14	36.69	140.55	330	100	36.74	140.59	395
IWTH21	39.47	141.93	13	100	39.45	141.96	24
IWTH26	38.97	141.00	125	108	39.05	141.05	75
MYGH04	38.79	141.33	35	100	38.68	141.45	39
MYGH06	38.59	141.07	20	100	38.63	141.19	5
SITH06	36.11	139.29	80	200	36.11	139.18	128
TCGH10	36.86	140.02	203	132	36.84	140.04	188
TCGH16	36.55	140.08	105	112	36.61	140.00	148
TCGH17	36.99	139.69	635	104	36.92	139.70	620
YMTH03	38.10	140.16	278	114	38.12	140.21	270
YMTH07	37.90	140.03	352	200	37.91	140.14	245
YMTH14	38.39	139.99	465	103	38.39	139.99	440

Table 2 Locations of the selected seismic stations and corresponding weather stations

separate the temporal influence of rainfall from velocity changes caused by other variables.

Station and record selection

We used the earthquake and weather logging data to investigate the temporal change of near-surface shear wave velocity associated with rainfall in Japan. Fifteen seismic stations were selected for further study from 697 stations in KiK-net according to following criteria. The detailed locations of the selected seismic and weather stations are given in Table 2.

Firstly, in order to obtain enough earthquake records, Northeast Honshu (the main island of Japan) was selected as the study area because this area is prone to earthquakes. For each seismic station, we used more than 10 years of seismic records from the time when the sampling frequency of the seismographs is changed to 100 Hz to June 2017. The preliminary condition for a seismic station to be selected is to record at least 500 events.

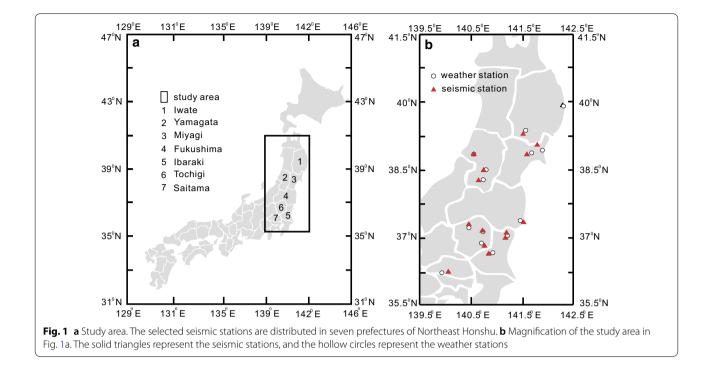
Secondly, climatic character was considered in this study. In the study area, several prefectures, such as Aomori, Akita and Niigata, are located at the seaside of Sea of Japan. Due to the water vapor of the warm current in winter, it often snows heavily in these prefectures. To avoid the effect of snowfall, these prefectures were eliminated from the study area. As shown in Fig. 1, we only selected seismic stations from seven prefectures in Northeast Honshu.

Lastly, seismic stations should satisfy the requirement that the corresponding weather stations had recorded adequate high-intensity rainfall events. For a selected seismic station, the distance between seismic station and its corresponding weather station should be less than 20 km, and the weather station should record at least two level VI events.

After station selection, because strong motion has an obvious impact on near-surface shear wave velocity (Takagi and Okada 2012), we then screened the seismic records for each selected station to eliminate the interferences of strong motion. For selected seismic stations, the effect of strong motion can be eliminated by the following two steps. We first selected earthquake records of weak motions with peak ground acceleration (PGA) < 50 cm/ s², and then removed records which were recorded in a period of time after strong motions with PGA > 200 cm/ s^2 . In this study, the influence time of strong motion depends on its PGA. Based on previous studies, for the strong motions whose PGA are between 200-400, 400-800 and larger than 800 cm/s^2 , the influence times are 30, 90 and 180 days, respectively (Nakata and Snieder 2012b; Wu and Peng 2012). Figure 2 shows an example of record selection for IBRH13. The total number of seismic records for IBRH11 is reduced from 1911 to 1113 after record selection. The number of seismic records with different rainfall classes for the selected seismic stations can be found in Table 3.

Seismic interferometry

Both cross-correlation and deconvolution can be used to retrieve the Green's function representations



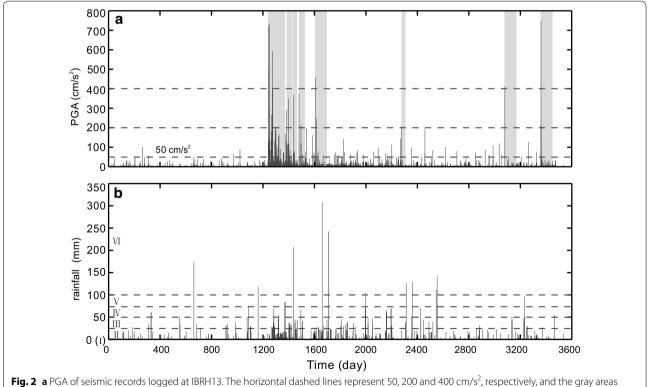


Fig. 2 a PGA of seismic records logged at IBRH13. The horizontal dashed lines represent 50, 200 and 400 cm/s², respectively, and the gray areas represent the influence time of strong motions. **b** Daily rainfall for IBRH13. The horizontal dashed lines represent 25, 50 and 100 mm, respectively. The zeroth day of the horizontal axis indicates the start date when sampling frequency of seismographs at IBRH13 is changed to 100 Hz (17 October 2007)

Station	I	II	111	IV	v	VI	Total
FKSH14	1067	477	47	12	5	5	1613
IBRH11	978	373	55	6	4	4	1420
IBRH13	672	331	45	8	5	5	1066
IBRH14	705	369	58	15	4	5	1156
IWTH21	764	298	23	13	5	3	1106
IWTH26	667	418	32	7	2	2	1128
MYGH04	1013	378	31	8	7	3	1440
MYGH06	1148	416	24	5	3	2	1598
SITH06	580	219	21	2	1	3	826
TCGH10	1142	546	76	19	6	5	1794
TCGH16	1010	394	53	15	7	6	1485
TCGH17	343	247	17	5	0	2	619
YMTH03	259	227	5	0	2	2	497
YMTH07	640	567	20	9	0	2	1247
YMTH14	473	490	43	8	2	6	1022

Table 3 Number of seismic records with different rainfall classes

between different receiver locations in seismic interferometry. These two types of Green's function representation have been verified as effective methods for vertical array to extract shear wave velocity in many previous studies (Miyazawa et al. 2008; Nakata and Snieder 2012a; Takagi and Okada 2012). Moreover, Takagi and Okada (2012) reported that the cross-correlation of the coda waves is stable in time as the deconvolution of direct waves used by Nakata and Snieder (2012a). In this study, we preferred deconvolutionbased interferometry because it can eliminate the effect of the incident wavefields to get a more reasonable estimation of shear wave velocity (Nakata and Snieder 2012a). We also compared the results from coda waves and those from direct waves for three stations (i.e., IBRH11, IBRH13 and TCGH17) using deconvolutionbased interferometry as shown in Additional file 2: Fig. S2. It can be found that the results from coda waves are similar to those from direct waves, because deconvolution interferometry can eliminate the effects of the incident wavefields for vertical arrays as we mentioned above.

The deconvolution function in frequency domain can be calculated by Eq. (1) and it can be converted into time domain by inverse Fourier transform.

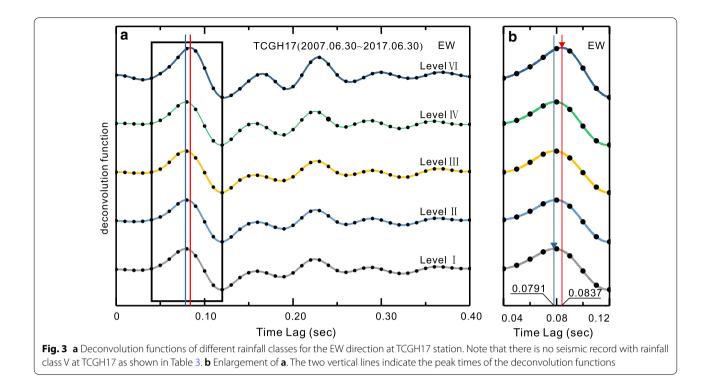
$$G(\mathbf{r}_{s}, \mathbf{s}, \omega) = \frac{u(\mathbf{r}_{s}, \mathbf{s}, \omega)}{u(\mathbf{r}_{b}, \mathbf{s}, \omega)} \approx \frac{u(\mathbf{r}_{s}, \mathbf{s}, \omega)u^{*}(\mathbf{r}_{b}, \mathbf{s}, \omega)}{|u(\mathbf{r}_{b}, \mathbf{s}, \omega)|^{2} + \varepsilon}$$
(1)

in which *s* is source location, $u(r_b, s, \omega)$ is the incoming wavefield recorded at the bottom sensor at location r_b ,

 $u(\mathbf{r}_{s}, \mathbf{s}, \omega)$ is the corresponding wavefield recorded at the surface sensor at location \mathbf{r}_{s} , * is the complex conjugate and ε is a regularization parameter, set as 1% of the average power spectrum of the borehole receiver (Nakata and Snieder 2012a), to enhance the stability of deconvolution function.

After computing deconvolution function, we grouped them by rainfall classes (Table 1) and then averaged them for each group. The shear wave velocities of different rainfall classes can be extracted using cubic spline interpolation to find peak times (Fig. 3b). These times are the travel times for shear waves that propagate between the bottom and surface sensors. By rotating the wavefields from 10° to 180° in 10° interval, we computed the shear wave velocity as a function of polarization angle. As described in Nakata and Snieder (2012a), the velocity can be separated into the isotropic and the anisotropic terms by a Fourier series expansion. The anisotropic component of the velocity is caused by shear wave splitting, which is beyond the scope of this study. The isotropic component can be calculated by averaging velocities of the eighteen directions. We assumed the splitting time to be much smaller than the travel time, and employed the isotropic component to represent the near-surface velocity.

Figure 3 shows an example of deconvolution functions for the EW direction in different rainfall classes at TCGH17 station. The deconvolution functions have clear peaks near the measured travel times of shear wave based on velocity profiles. The peak time of the level VI lags 4.6 ms from that of level I (Fig. 3b). It could



be clearly observed that high-intensity rainfall has an obvious influence on near-surface shear wave velocity for TCGH17 station.

Results

We computed the shear wave velocities of different rainfall classes from deconvolution functions and calculated the relative velocity change $(d\nu/v)$ for these 15 selected stations. The relative velocity change is defined as

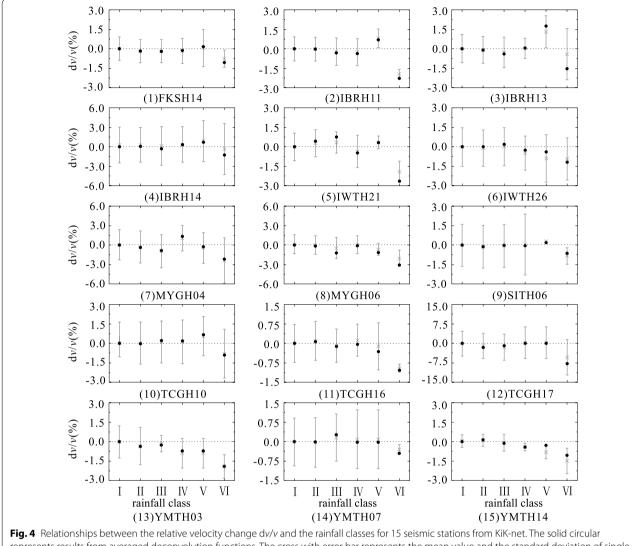
$$d\nu/\nu = (\nu_i - \nu_{\rm I})/\nu_{\rm I} (i = {\rm II, III, IV, V, VI})$$
(2)

in which v_i is the near-surface shear wave velocity for level *i*.

Figure 4 depicts the relationship between the relative velocity change $d\nu/\nu$ and the rainfall classes. It can be found that the results from averaged deconvolution function are comparable with the mean values of single measurements. Meanwhile, the relative high standard deviations of single measurements indicate the necessity of the averaging process. As discussed in Wang et al. (2017), seismic velocity changes in Japan are subjected to nonidentical environmental factors in different areas. According to local geological, climatic and geographical conditions, we infer that the seismic velocity variations in Northeast Honshu are mainly controlled by the strong motion, precipitation and sea level changes. It also should be noted that Northeast Honshu is one of the regions with the smallest seasonal seismic velocity changes in Japan, which allows detection of temporal change of seismic velocity associated with rainfall that might be masked by the strong seasonal perturbations in other areas.

It can be found that the surface shear wave velocities do not change obviously when rainfall class varies from level II to V, and the velocity decreases significantly when rainfall class reaches level VI. It can be concluded that there exists a threshold (rainfall class level VI) of the temporal influence of rainfall on near-surface shear wave velocity.

Nakata and Snieder (2012a) reported that the stations located at soft rock sites are influenced more by rainfall than the stations at hard rock sites. We followed the definition of the soft and hard rock sites of their study. The estimated shear wave velocity of hard rock sites is greater than 600 m/s, while that of soft rock sites is less than 600 m/s. Among the 15 stations, there are four stations (IBRH14, MYGH04, TCGH17 and YMTH03) are located at hard rock sites. However, in this study, it did not show that stations at soft rock sites are influenced more by rainfall than stations at hard rock sites. Moreover, the relative velocity change caused by high-intensity rainfall of TCGH17 station is much higher than that of 12 stations at soft rock sites (Fig. 4). One possible interpretation is that the relative velocity change $(d\nu/\nu)$ highly depends on the borehole depth (Takagi and Okada 2012). More detailed explanation can be found in "Discussion" section.



represents results from averaged deconvolution functions. The cross with error bar represents the mean value and the standard deviation of single measurements. The velocity does not change obviously when the rainfall class ranges from level II–V, while it is reduced by 1–10% when the rainfall class reaches level VI

Discussion

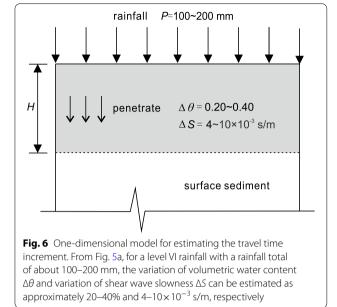
First, we discussed the threshold of the temporal influence of rainfall on the near-surface shear wave velocity. In a recent study, Dong and Lu (2016) used bender element technique to measure the dependencies of shear wave velocity on saturation for seven different types of unsaturated soils. They reported there are two thresholds of the influence of water content on shear wave velocity. The laboratory test results show that, as the volumetric water content varies from zero to the first threshold of around 0.15, the shear wave velocities of different soils decrease sharply. When the water content is in the range of approximately 0.15–0.40, the shear wave velocities gradually decrease with a mild rate as the water content increases. After the water content exceeds the second threshold of about 0.40, the shear wave velocities decrease significantly with the water content increases. The velocity reduction is mainly caused by the decreased effective stress (Dong and Lu 2016; Snieder and Beukel 2004). The detailed properties of tested soils and the classifications can be found in Dong and Lu (2016).

Generally speaking, the initial water content of the in situ soil is higher than the first threshold. The water content of the surface sediment is influenced by rainfall directly, thus the relationship between near-surface velocity and rainfall is similar to that between shear wave velocity and water content. It is not difficult to understand the thresholds of rainfall for different stations are quite similar because thresholds of water content for different soils are about 0.4, ranging from 0.35 to 0.45.

Compared with wave velocity, wave slowness (the inverse of wave velocity) is a direct indicator of wave propagation time (the travel time is proportional to the wave slowness) mathematically, and has been widely used in geophysics and earthquake engineering (Boore and Asten 1983; Brown et al. 2002; Eisner et al. 2009; Hao and Stovas 2016; Steck 1995; Steck and Prothero 1993; Xiang et al. 1999). As shown in Fig. 5a, the threshold of the temporal influence of rainfall on near-surface shear wave velocity could be more clearly explained by the relationship between shear wave slowness and water content.

Since major velocity change is localized in shallow layer, the travel time increment is independent of the borehole depth. Thus we presented the relationship between travel time increment and rainfall class in Fig. 5b. Compared with the travel time corresponding to the rainfall class level I (no rain), the travel time increment is no more than 3×10^{-4} s when rainfall class varies from level II to V, and it ranges between 1×10^{-3} and 7×10^{-3} s (the mean value and standard deviation are 3.29×10^{-3} and 1.36×10^{-3} s, respectively) when the rainfall class reaches level VI.

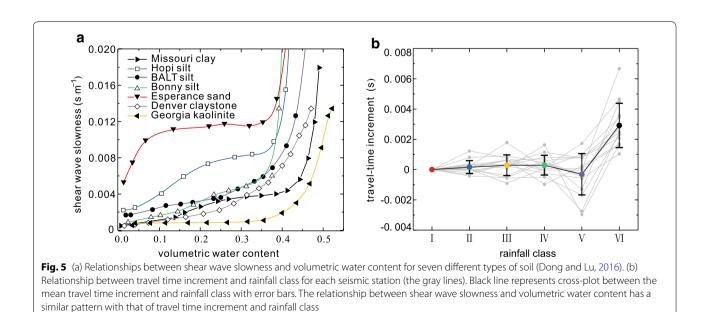
To get a deeper insight into the temporal influence of rainfall on near-surface shear wave velocity, we proposed a one-dimensional model to estimate travel time increments. As shown in Fig. 6, *P*, *H*, $\Delta\theta$ and ΔS are rainfall totals, permeate depth, the variation of surface volumetric water content and variation of shear wave slowness after rain, respectively. The travel time increment Δt can be obtained as follows:



$$\Delta t = \frac{H}{\nu_2} - \frac{H}{\nu_1} = H \cdot \Delta S \tag{3}$$

in which v_1 and v_2 are surface shear wave velocities before and after rain, respectively. In the calculation model, we assumed the rainfall is completely and evenly absorbed by the surface sediments. Hence we can get:

$$P = H \cdot (\theta_2 - \theta_1) \tag{4}$$



in which θ_1 and θ_2 are surface volumetric water content before and after rain, respectively. Combining Eqs. (3) and (4) yields:

$$H = \frac{P}{\Delta\theta} \tag{5}$$

$$\Delta t = \frac{P \cdot \Delta S}{\Delta \theta} \tag{6}$$

Equations (5) and (6) can be used to estimate the permeate depth and the travel time increment, respectively. From Fig. 5a, when the rainfall class reaches level VI, the variation of volumetric water content $\Delta \theta$ and variation of shear wave slowness ΔS can be estimated as approximately 20–40% and $4-10 \times 10^{-3}$ s/m, respectively. And the rainfall total of level VI is about 100-200 mm according to the meteorological records. Hence we can get H=0.25-1.0 m and $\Delta t=1-10\times10^{-3}$ s for level VI. The result of H is consistent with the previous analytical solutions for one-dimensional, transient infiltration (Srivastava and Yeh 1991) and it reveals that the temporal influence of rainfall on travel time increment is only related to surface sediments (around 1 m, which is far less than the borehole depth). The estimated travel time increments are basically consistent with our results of observation (Fig. 5b), which validate the presented formula of Eq. (6).

Conclusions

In this study, we investigated temporal change of nearsurface shear wave velocity associated with rainfall using earthquake and weather logging data in Northeast Honshu. The relative velocity reduction is observed as several percent under high-intensity rainfall. The relationship between near-surface velocity and rainfall is similar to that between shear wave velocity and water content. Based on the results of observation, we proposed a onedimensional model to explain the temporal influence of rainfall on near-surface shear wave velocity. The conclusions of this study are summarized as follows:

- 1. The near-surface shear wave velocity does not change obviously under low-intensity rainfall, while it decreases significantly when the rainfall intensity exceeds a threshold (around 100 mm of 24 h rainfall total).
- 2. Compared with the travel time of no rain, the travel time increment of low-intensity rainfall is no more than 3×10^{-4} s, while that of high-intensity rainfall ranges between 1×10^{-3} and 7×10^{-3} s (mean value and standard deviation are 3.29×10^{-3} s and 1.36×10^{-3} s, respectively).

3. The velocity change caused by rainfall is directly related to the dependency of velocity on saturation of surface sediments (around 1 m). According to the one-dimensional model proposed in this study, the travel time increment can be estimated by multiplying rainfall *P* and the ratio of the variation of the shear wave slowness ΔS to variation of volumetric water content $\Delta \theta$. This model can give a preliminary estimation of the temporal influence of rainfall on near-surface shear wave velocity.

Additional files

Additional file 1: Fig. S1. Relationships between the relative velocity change and the rainfall classes for different accumulated periods of rainfall amounts.

Additional file 2: Fig. S2. Relationships between the relative velocity change and the rainfall classes for direct waves and coda waves.

Abbreviations

CMA: China Meteorological Administration; JMA: Japan Meteorological Agency; NIED: National Research Institute for Earth Science and Disaster Resilience; PGA: Peak ground acceleration.

Authors' contributions

YM and YS participated in acquisition of seismic and meteorologic data. YS processed the obtained data. YS and SYW drafted the manuscript. YM, YS, and SYW contributed to the planning of this study and interpretation of the data. All authors read and approved the final manuscript.

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Competing interests

The authors declare that they have no competing interests.

Availability of data and materials

The earthquake and weather data were provided by National Research Institute for Earth Science and Disaster Prevention and Japan Meteorological Agency, respectively. The earthquake records are available on http://www. kyoshin.bosai.go.jp/kyoshin/quick/index_en.html. The weather records are available on http://www.data.jma.go.jp/obd/stats/etrn/index.php.

Consent for publication

Not applicable.

Ethics approval and consent to participate

Not applicable.

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