Preliminary damage mapping following the 2016 Kumamoto Mw7.1 Earthquake using pre- and post-event PALSAR-2 images

http://www.regid.irides.tohoku.ac.jp/response/

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BACKGROUND

- M_w6.4 earthquake occurred on April 14, 2016 21:32 (JST) in Kumamoto City, Kyushu, Japan. This event was followed by two big earthquakes of M_w6.4 on April 15, 2016 00:09 (JST) and Mw7.1 on April 16, 2016 01:30 (JST). It was reported that the latest event generated most of the damage in Kumamoto city and surrounding areas.
- The ReGiD Lab. (Laboratory of Remote Sensing and Geo-informatics for Disaster Management), IRIDeS, Tohoku University conducted a preliminary building damage estimation using two different approaches. First, conventional change detection approach using pre- and post-event Synthetic Aperture Radar (SAR) images (Liu et al., 2013). Second, a new approach of machine learning using only post-event SAR image (Bai at at., 2016).
- In this preliminary damage assessment, we employed a two post-event SAR images acquired after the main shock of Mw 7.1 (third significant earthquake) on 16th April. The image from Kumamoto city was acquired by the Advanced Land Observing Satellite 2 (ALOS-2) / L-band Synthetic Aperture Radar (PALSAR-2) of JAXA.

JAXA ALOS-2/PALSAR-2 Data Acquisition

Area	PALSAR-2 data	Acquisition	Туре	Polarization	Indecent angle
	Post-event	20160417	L1.5	VH-VV	45.1
	Pre-event	20150517	L1.5	VH-VV	45.1
	Post-event	20160417	L1.5	нн	45.1
	Pre-event	20150517	L1.5	НН	45.1
	Post-event	20160415	L1.5	нн	28.7
	Pre-event	20141114	L1.5	НН	28.7
	Post-event	20160417	L1.5	VH-VV	28.8
	Pre-event	20141114	L1.5	VH-VV	28.7
	Post-event	20160416	L1.1	нн	18.8
	Pre-event	20160330	L1.1	HH	38.2

JAXA ALOS-2/PALSAR-2 Data Coverage



Change Detection Approach

Change Detection Approach

Pre-processing

The PALSAR-2 data was converted from digital number to backscattering coefficient (sigma-naught [dB]) using the equation provided by JAXA website (http://www.eorc.jaxa.jp). Then, the PALSAR-2 data was filtered using a Lee filter (Lee 1980) using a 3 x 3 pixel window.

Damage estimation

- 1. First, using the available VH-polarization images, we calculated the difference in backscattering and correlation coefficients using a 15 x 15 pixel window (37.5m x 37.5m) (Matsuoka et al., 2004).
- 2. The change detection was conducted using the Change Factor (*z*-coefficient) introduced by Liu et al., (2013). Theoretically, high values of *z*-coefficient indicates high probability of changes (Liu et al., 2013).

PALSAR-2 Post-event Images

Sigma-naught (σ°) images

- Acquisition mode: StripMap (SM1)
- Ascending Orbit direction (azimuth)/ Right side observation (range)
- VH- and VV-polarizations
- Data format Level 1.1 / Pixel Spacing: 2.5 m

2015-05-17



2016-04-17



Preliminary Damage Mapping

Kumamoto city (熊本市)





Z-factor (Liu et al., 2013) $z = \left| \frac{d}{\max |d|} \right| - 0.5 \cdot r$

1.0

-0.4

- d = difference of backscattering coefficients, r = correlation coefficient.
- ↔ High values of the change factor (z ≈ 1.0) show high changes around the Kumamoto city (red spots).

Liu, W., F. Yamazaki, H. Gokon and S. Koshimura, Extraction of Tsunami-Flooded Areas and Damaged Buildings in the 2011 Tohoku-Oki Earthquake from TerraSAR-X Intensity Images, Earthquake Spectra, Vol. 29, No. S1, pp. S183-S200, 2013. doi: 10.1193/1.4000120

広域被害把握研究分野 (ReGiD Lab.)

Preliminary Damage Mapping

Kumamoto Castle(熊本城)







Machine learning approach

Machine Learning Approach

Pre-processing

The PALSAR-2 data was converted from digital number to backscattering coefficient (sigma-naught [dB]) using the equation provided by JAXA website⁽¹⁾. Then, the PALSAR-2 data was filtered using a Lee filter (Lee 1980) using a 3 x 3 pixel window.

Damage estimation

- 1. We calculated 37 texture parameters based on the Grey-Level Co-occurrence Matrix (Haralick et al., 1973), and the HH-polarization information.
- 2. We conducted the damage estimation in block-scale of 50 m x 50 m, the average value of each texture parameter is calculated in each block.
- 3. Finally, using training sample of structural damage observed in the 2015 Nepal earthquake, a random forest algorithm (Breiman et al., 2001) was applied to classify the building damage into two categories in Kumamoto city.

Method (Bai et al., 2015)



PALSAR-2 Post-event Image (2016/04/16)

Sigma-naught (σ°) image



- Acquisition mode: StripMap (SM1)
- Ascending Orbit direction (azimuth)
- Right side observation (range)
- HH-polarization
- Data format Level 1.5
- Pixel Spacing: 2.5 m

Preliminary Damage Mapping (Mashiki town)



Preliminary Damage Mapping







32°47'10"N

Post-event Image (GSI)

Pre-event Image (GSI)

Summary

- We have conducted a preliminary damage mapping in Kumamoto city and Mashiki town using pre- and post-event ALOS-2 images with VH-polarization and post-event HH-polarization, respectively.
- According to damage estimation, the change detection approach applied in Kumamoto city showed spots of high changes, which can be interpreted as damaged buildings. However, several misdetections were due to new construction found in the post-event SAR image.
- However, the change detection analysis successfully detected parts of damages around Kumamoto Castle. Based on the comparison of aerial photos (GSI), these damage is consistent with reported building collapse.
- On the other hand, machine learning approach applied in Mashiki town showed that small 20 % of built-up area were detected as damaged. But the results need to be verified with in-situ data.

Acknowledgement

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