

INTEGRATION OF INSAR TECHNIQUE, GOOGLE EARTH IMAGES AND EXTENSIVE FIELD SURVEY FOR LANDSLIDE INVENTORY IN A PART OF CAMERON HIGHLANDS, PAHANG, MALAYSIA

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(Received 30th Aug 2018; accepted 7th Nov 2018)

Abstract. Landslide is among hazardous kind of disasters, which claiming many of lives and billions of dollars of damages in property every year. Detecting the spatial distribution of landslides is highly important in decision making. Interferometry synthetic aperture RADAR (InSAR) is a technique have been employed in many studies, including land surface deformation. The Google Earth (GE) with images in high spatial resolution provide a potential for regional landslide mapping, especially in vegetated areas, where landslides creates bare lands. This study aims to detect landslide locations using integration of InSAR technique, Google Earth images and extensive field survey in a part of Cameron Highlands, Pahang, Malaysia. A pair of synthetic aperture RADAR (SAR) images were acquired from scihub.copernicus.eu website. In order to extract surface deformation, either imageries were pre-processed and processed through sentinel application platform (SNAP) software. A total number of 152 landslide locations were extracted from phase, coherence and unwrapped bands with the help of the Google Earth images, where they were overlaid on the Google Earth and by zooming in deformed surfaces the landslide locations were identified, extracted and converted into layer format using Geographical Information System (GIS). Geographical Positioning System (GPS) was utilized for validation purposes.

Keywords: *remote sensing, geographical positioning system, landslide detection, Cameron Highlands, Malaysia*

Introduction

Landslide is one of major type of geo-hazards claiming thousands of lives and causing billions of dollars of damages in property every year (Shahabi et al., 2013). Landslide is triggered by some extreme events, including earthquakes, precipitations, volcanic eruptions, anthropogenic-based actions and etc. However, identifying the spatial distribution of landslides is crucial for managing and planning in situ investigations (Yang and Chen, 2010; Shahabi and Hashim, 2015). Interferometry synthetic aperture RADAR (InSAR) is a technique, which can be employed in numerous fields, including tectonic (Massonnet et al., 1993), volcanic (Stevens and Wadge, 2004; Wright et al., 2006), subsidence (Tomás et al., 2005, 2014; Herrera et al., 2007), landslides (Sun et al., 2015; Tomás et al., 2016), ice flow (Goldstein et al., 1993), digital elevation model (DEM) generation (Pauciullo et al., 2015). The Google Earth (GE) images have been widely developed and used in many sectors. With free

images in high spatial resolution the Google Earth images provide high potential for regional landslide mapping (Van Den Eeckhaut et al., 2012; Pham et al., 2016), especially in vegetated areas, where the landslides create bare lands, therefore can be easily detected through visual interpretation.

Due to the weak soil structure and heavy rainfall, Cameron Highlands is highly prone to landslide and has faced many landslides to date, which led to significant damages in properties (Jebur et al., 2015). Even for areas of a few km², RADAR imageries are remarkably expensive and acquiring such data are quite challenging. The study area is covered by cloud almost a whole year and on the other side is highly vegetated. However, the only alternative for landslide detection in such areas is RADAR remote sensing, which can penetrate through cloud coverage, vegetation foliage and work day and night (Chitroub et al., 2002; Lillesand et al., 2014).

RADAR imageries are costly, but SENTINEL-1 satellite products are among the most important and applicable source of RADAR data, which can be obtained online. SAR image is independent of solar energy and information can be taken day and night (Chitroub et al., 2002). Interferometry synthetic aperture RADAR (InSAR) technique is a RADAR technique used in many areas, including land surface deformation. Satellite synthetic aperture RADAR images are taken in separate days, therefore changes that have occurred between RADAR images are then recorded. This technique can monitors millimetre to centimetre scale deformation resulting from landslides, earthquakes, volcanoes and human induced factors (Van Baelen and Richmond, 1991). However, with millions of users, GE imagery has been recognized as one of the important sources of scientific data (Potere, 2008).

Many studies have been focused on landslide inventory by using different models, methods and imageries worldwide, including aerial photo interpretation (Wieczorek, 1984; Brardinoni et al., 2003; Fiorucci et al., 2011; Guzzetti et al., 2012; Strozzi et al., 2013; Li et al., 2016; Garcia-Urquia and Yamagishi, 2017; Chen et al., 2018; Marchesini et al., 2018), satellite imageries (Singhroy et al., 1998; Nichol and Wong, 2005; Nichol et al., 2006; Marcelino et al., 2009; Fiorucci et al., 2011; Ardizzone et al., 2013; Shahabi et al., 2014, 2015; Pradhan et al., 2016; Wang et al., 2018), the Google Earth images (Costanzo et al., 2012; Van Den Eeckhaut et al., 2012; Li et al., 2015; Conoscenti et al., 2016; Maki Mateso et al., 2016; Pham et al., 2016, 2017; Broeckx et al., 2018; Yamagishi and Moncada, 2018), interferometry synthetic aperture RADAR (InSAR) technique (Singh et al., 2005; Greif and Vlcko, 2012; Motagh et al., 2013; Bellotti et al., 2014; Tang et al., 2015; Tomás et al., 2016; Bayer et al., 2018; Tien Bui et al., 2018), differential interferometry synthetic aperture RADAR (DInSAR) technique (Cascini et al., 2010; Herrera et al., 2013; García-Davalillo et al., 2014; Ferlisi et al., 2015; Infante et al., 2016; Calvello et al., 2017; Nobile et al., 2018).

However, convergence of above mentioned methods was the strong point of the current study compared with other studies, where only one or two methods used for landslide inventory (Cascini et al., 2010; Bellotti et al., 2014; Ferlisi et al., 2015; Bayer et al., 2018). On the other hand, for a better result the Google Earth imagery as well as extensive field checks were highly needed in this study, which concern with the hilly situation of Camron Highlands, another issue was the field survey. Therefore, integration of InSAR technique, the Google Earth image and extensive field survey as well as employing SENTINEL-1 satellite imagery as the only available RADAR imagery online for free, are the significances of the current study. Finally, the main objective of this work was to detect landslide locations for a part of Cameron

Highlands, Pahang, Malaysia by using integration of interferometry synthetic aperture RADAR (InSAR) technique, the Google Earth images and extensive field investigation.

Materials and methods

Description of the study area

A part of Cameron Highlands, Pahang, Malaysia was selected for the application of landslide detection, which is located in the south western part of Cameron Highlands with an area of 81.249 km². The study area (*Fig. 1*) is experiencing rapid development of land clearing for housing, hotels and apartments, which lead to erosion and landslides (Pradhan et al., 2010). Cameron Highlands is located in the Main Range of Malaysia geological units (Makoundi et al., 2014). In the study area, there are two kind of formations, including acid intrusive and schist, phyllite, slate, limestone with minor intercalation of sandstone. The lowest and the highest area in the region is 953 and 1944 m above the sea level, respectively. Refer to Tropical Rainfall Measuring Mission (TRMM) data, the average rainfall in the study area for the year 2017 fluctuates between 3800 mm to 4200 mm annually. There are two wet seasons in Malaysia from September to December and also from February to May (Pradhan et al., 2010; Rasul et al., 2017). However, the rainfall in Cameron Highlands express itself between March and May and from November to December (Pradhan et al., 2010). During these specific times, rivers in Cameron Highlands overflow and flooding the surrounding areas, therefore landslides may occur along the river banks.

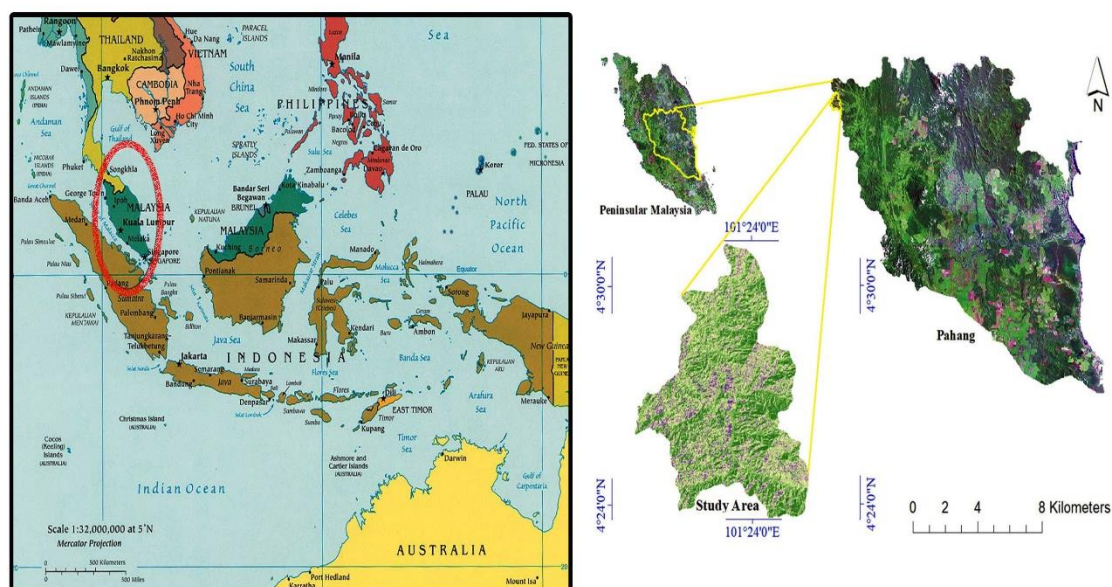


Figure 1. Shows geographical location of the study area in Asia and Malaysia

Data base and data collection

Images that are taken in separate days and positions are able to record occurred changes between two imageries. Two or more satellite imageries of the same area and the same position, which used for topographic studies at slightly different positions are generally applied to create map of surface deformation using interferometry synthetic

aperture RADAR (InSAR) technique (Massonnet and Feigl, 1998; Bürgmann et al., 2000; Hanssen, 2001). For this matter we acquired two SENTINEL-1 images of a different dates and different topographic positions for generating interferogram in this study.

SENTINEL-1 is the first Copernicus Programme, which conducted by the European Space Agency (ESA). This product consists of the two satellite platforms of SENTINEL-1A and SENTINEL-1B. SENTINEL-1 has four sensor modes of interferometric wide swath (IW), strip map (SM), wave (WV), and extra wide swath (EW), however, three product types of SENTINEL-1 are namely, ground range detected (GRD), single look complex (SLC), and ocean (OCN) (Potin et al., 2012).

It is in C-band with 5.7 cm wavelength, which able to get data day and night in all weather conditions, however, the short wavelength of C-band SAR sensors is an obstacle in highly vegetated areas, because cannot penetrates through thick trunk and branches of trees (Geudtner et al., 2014). In this study a pair of SENTINEL-1 image were obtained from scihub.copernicus.eu website for free. *Table 1* and *Figure 2* illustrate the technical characteristics of SENTINEL-1 data and geographical position of the study area on satellite imagery, respectively.

Table 1. Technical attributes of SENTINEL-1 data used in the study area

Platform	Sensor mode	Product type	Path	Date
S1A	Interferometry Wide swath (IW)	Single Look Complex (SLC)	Ascending	20/02/2017 04/03/2016

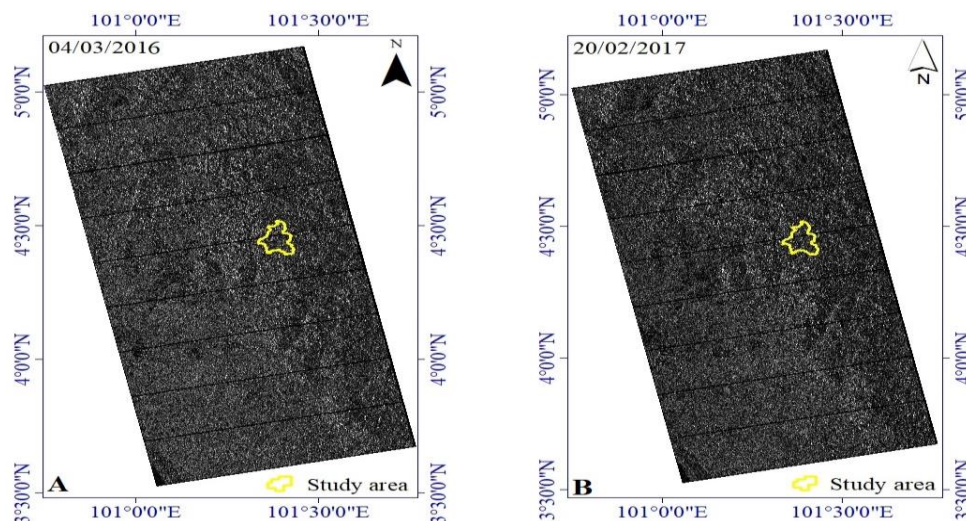


Figure 2. SENTINEL-1 data used in the study area

Data pre-processing and processing

The comparison of the phase values in two different radar images, known as InSAR technique (Navarro-Sanchez et al., 2010; Jebur et al., 2015). A pair of synthetic aperture RADAR (SAR) images are required to generate an interferogram. However, prior to InSAR processing the raw data must be pre-processed. In this study, in order to pre-

process and process the imageries, sentinel application platform (SNAP) software was employed. *Figure 3* clearly shows the step by step methodology of the current study.

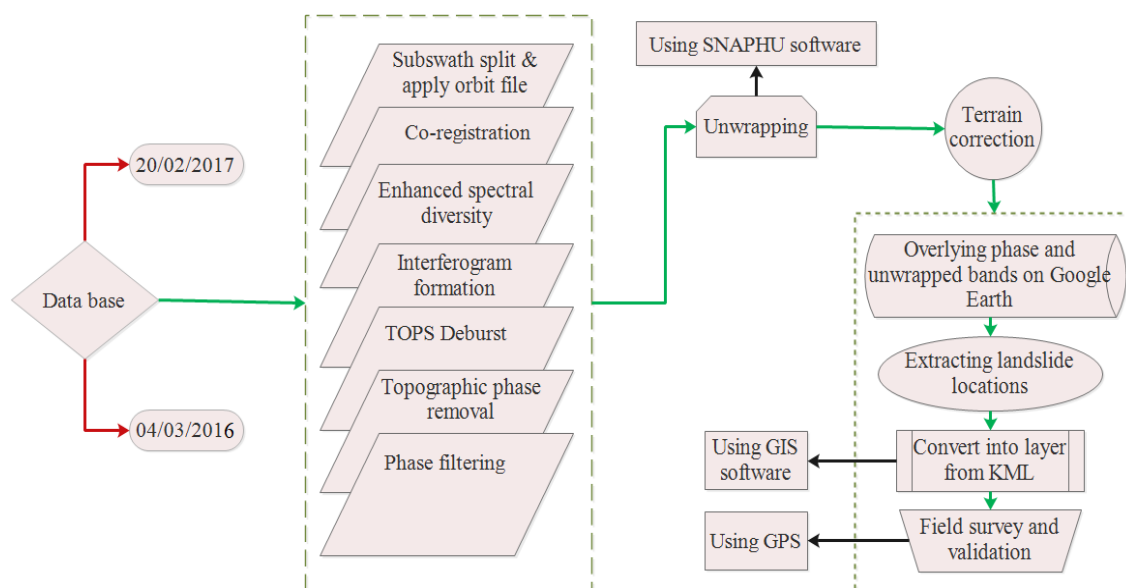


Figure 3. Methodology of the study area

TOPS split and apply orbit file

SENTINEL-1 products have three sub-swaths, which must be split and based on the study the appropriate one should be selected (Luis, 2015). After sub-swaths step, the orbit file has to be updated. 3 sub-swaths of IW1, IW2 and IW3 can be seen in SENTINEL-1 imageries, which each of them belongs to an adjacent acquisition (*Fig. 4*).

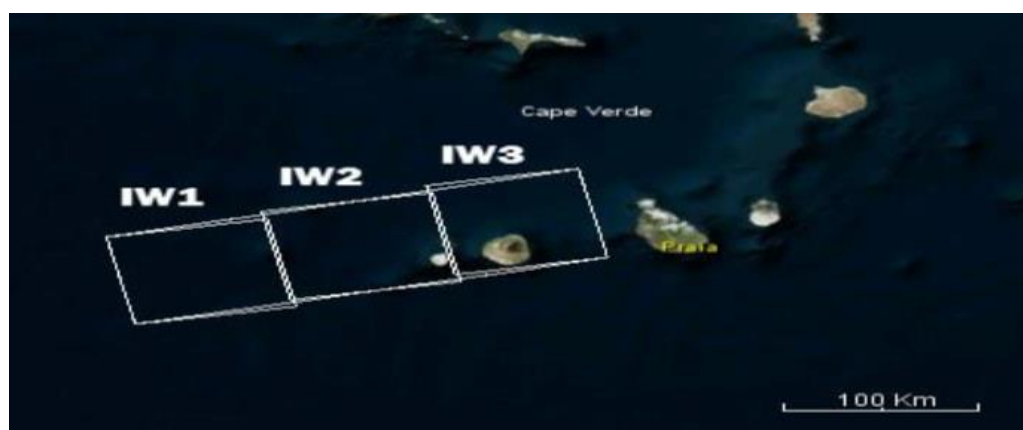


Figure 4. Sub-swaths IW1, IW2, IW3 of the study area (Luis, 2015)

Co-registration and enhanced spectral diversity

One of the main basic steps of InSAR processing is accurate co-registration of the slave and master images, by which the slave image is then resampled to match the

geometry of the master image. Co-registration process is done by using the coherence between the two satellite imageries (Kervyn, 2001). Once the imageries have been successfully stacked, it is then the co-registered image should be spectrally enhanced.

Interferogram formation

The interferogram is formed by using cross multiplication of each pixel in the two images (Kervyn, 2001; Nobile et al., 2018). An interferogram will be measured by the calculation of phase differences ($\Delta\Phi_{int}$) between the two co-registered SAR imageries (Schlögel et al., 2015; Eq. 1):

$$\Delta\Phi_{int} = \Phi(t_0) - \Phi(t_1) = \Delta\Phi_{def} + \Delta\Phi_{orb} + \Delta\Phi_{topo} + \Delta\Phi_{atm} + \Delta\Phi_{noise} \quad (\text{Eq.1})$$

where $\Delta\Phi_{int}$ is interferometric phase, $\Phi(t_0)$ and $\Phi(t_1)$ are the phase values at day t_0 and day t_1 , respectively. $\Delta\Phi_{def}$ is deformation between two acquisition times on ground, $\Delta\Phi_{topo}$ is topographic contribution, $\Delta\Phi_{orb}$ is orbital contribution, $\Delta\Phi_{atm}$ is atmospheric contribution and $\Delta\Phi_{noise}$ is noise contribution. In this step, the phase and the coherence bands (Figs. 5A and 5B, respectively), were generated.

Deburst

TOPS data is acquired in bursts, which is separated by demarcation zones. However, any data within these zones can be considered invalid and must be debursted (Fig. 5C).

Topographic phase removal and phase filter

The created interferogram should be flattened using removing the topographic phase, by which the interferometric phase due to the curvature of the Earth is removed, however, to simulate the contribution of the topography to the interferometric phase a digital elevation model can be employed in conjunction with the baseline image, the topographic phase can then be removed from the interferogram (Kervyn, 2001). Because of the noisy nature of the created interferogram, a proper filter must be applied. There are a few common filters, which refer to Jebur et al. (2015) the best noise reduction is the Goldstein filter.

Unwrapping

The phase unwrapping associated with the integration of phase difference by adding the integer number of cycles, which minimises the phase difference (Kervyn, 2001). The range of the phase is represented in 2π (Jebur et al., 2015). Once the time in phase reaches to 2π , the cycle will be repeated (Navarro-Sanchez et al., 2010). However, to address the 2π ambiguity, phase unwrapping was employed using SNAPHU Software (Fig. 5D).

Terrain correction

Generally, terrain correction is the final step, by which the created InSAR will be geo-referenced and geometrically corrected. Terrain correction converts an image from either slant range or ground range geometry into a coordinate system (Kervyn, 2001).

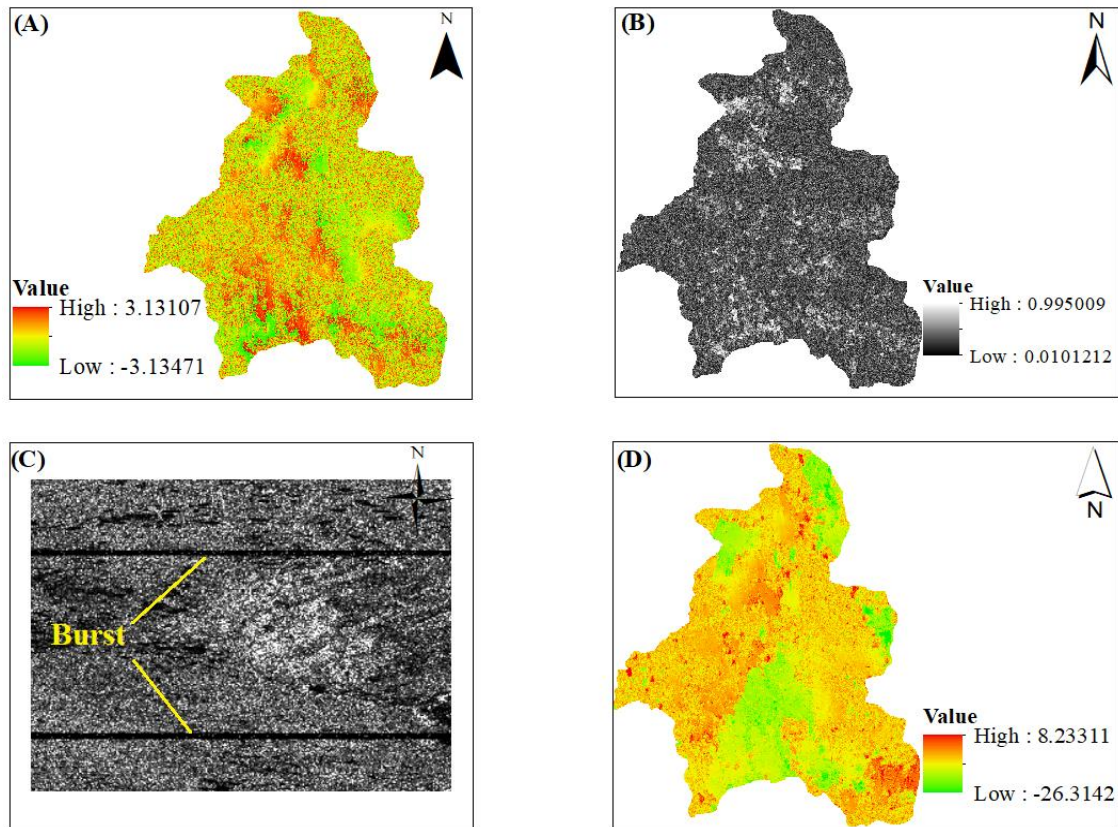


Figure 5. Interferogram formation process of applied data; (A) phase band, (B) coherence band, (C) bursts in SENTINEL-1 imagery and (D) unwrapped band

The Google Earth and overlying the phase, coherence and unwrapped bands on it to detect landslide locations

The Google Earth (GE) was originally developed in 2001 and its imageries displayed on a digital sphere retrieved from satellites or aircraft, which display the surface of the planet by a single composited image (Potere, 2008). However, after zooming, the imageries transitioned into finer details, which differ in time and date from one area to the next. Before Landsat-8 products, the imageries of Landsat-7 were used, which faced a hardware malfunction that resulted in spatial gaps in images (Markham et al., 2004). Now Landsat-8 is used to provide imageries in a higher quality and frequency. However, resolution of the Google Earth imageries ranges from 15 m to 15 cm (Potere, 2008). The Google Earth is a valuable and free application that enables users to overlay GIS data, satellite images and other maps from global to regional scales (Sönmez et al., 2011). The GE can be utilized to enhance cross discipline unification to support the national geography standards and education (Patterson, 2007).

In the current study, the phase, coherence and unwrapped bands of the generated InSAR map with the assist of the GE images were used to digitize landslide locations. Where, created phase, coherence and unwrapped bands were overlaid on the Google Earth. However, by zooming in and scrolling around in those areas in the abovementioned bands that there were a sign of land surface deformation the landslide locations were identified and digitized.

Validation

Accuracy assessment points out the quality of obtained information (Alqurashi and Kumar, 2014). In this study, Geographical Positioning System (GPS) was used as a valuable tool for validating the extracted landslides. GPS is defined as one of the best techniques in deformation monitoring (Xiao and He, 2013). Because of its automated module and applicability in all-weather conditions, GPS by far is more helpful than the traditional geodetic survey (Leick et al., 2015). The spatial resolution of GPS depends highly on density of the antenna (Xiao and He, 2013). In the current study, 20% of the detected landslides (30 locations out of 152 locations), were transferred into GPS to be validated in the region.

Results and discussion

Landslide inventory

Landslides are abrupt geomorphic events, which constitute the rapid downward motion of materials (Makoundi et al., 2014; Marchesini et al., 2018). The triggering mechanism for landslide in the study area is excessive precipitation, which downwards the natural stability of the slope, resulting in falling and sliding (Pradhan et al., 2010). Because of their stereo viewing capability and even high spatial resolution, RADAR satellite imageries have been extensively used to produce landslide inventory maps (Pradhan et al., 2016; Nobile et al., 2018).

Remote sensing technique is among helpful methods for identification of landslides (Singhroy et al., 1998; Guzzetti et al., 2012). Historical landslides for the study area were first recognized by the InSAR technique and then by overlying the phase, coherence and unwrapped bands on the Google Earth, the landslide locations were identified, digitized in either polygon or point layers. Using Geographical Information System (GIS), these layers were converted into shape-file format.

Figure 6 clearly shows the deformed surfaces in the phase, coherence and unwrapped bands. These bands were generated from SENTINEL-1 satellite imagery for two different dates (04/03/2016 and 20/02/2017), by using interferometry synthetic aperture RADAR (InSAR) technique and the SNAP software. In the phase and unwrapped bands (*Figs. 6A* and *C*), the deformed surfaces are shown in the red colour (The highest amount) and the green colour (The lowest amount). Furthermore, the coherence band is shown the deformed surfaces better than them, because white areas (High value), are clearly differentiated from stable places (Dark value) (*Fig. 6B*).

The phase, coherence and unwrapped bands of the generated InSAR map were converted into KML format and then were overlaid on the Google Earth. Wherever there was a sign of deformation by zooming in and scrolling around the landslide locations were detected and digitized. After zooming in the GE images, the imageries transitioned into the finer details. Which, deformed surfaces can be easily recognized, especially in vegetated areas, where landslide left bare lands. However, the study area of the current study is highly vegetated and completely appropriate for this kind of method (*Fig. 7*).

Table 2 and *Figure 8* show the number of landslide inventory and their geographical position on the slope, aspect, and elevation maps, respectively. However, a total number of 129 landslides out of 152 landslides were occurred on the slope 11 to 40%. Also based on the aspect map, the lowest and the highest number of the landslides were

occurred on the flat and southeast classes of 0 and 32 landslides, respectively. Moreover, no landslide was found for elevation 1659 to 2067 m above the sea level, while 47 landslides were detected on the elevation class of 1,168 to 1,253 m above the sea level.

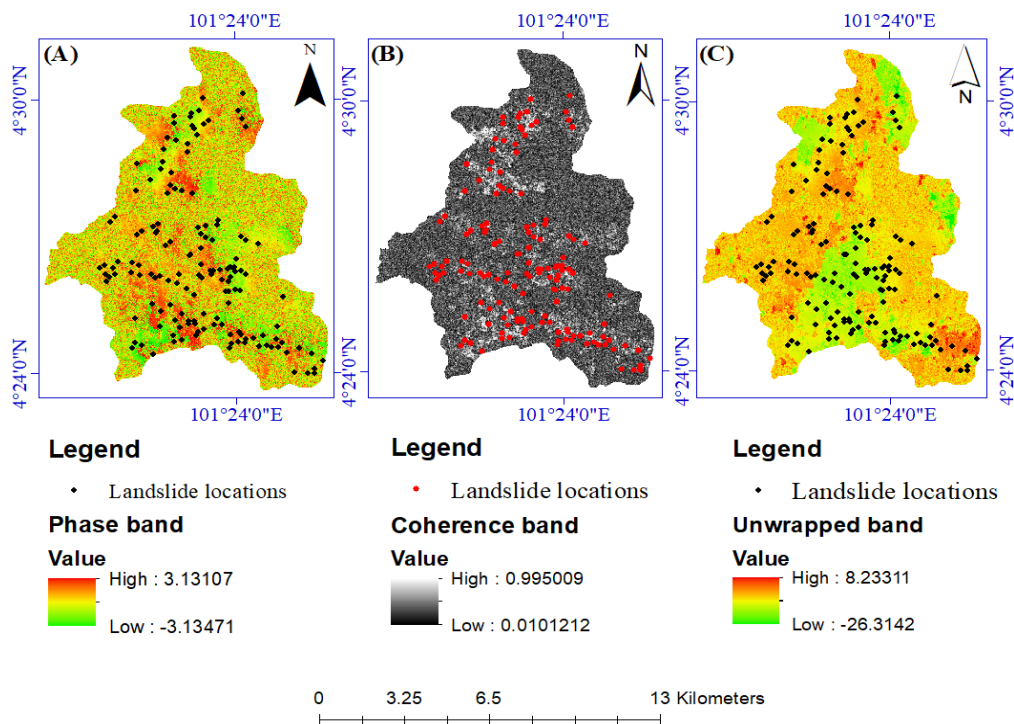


Figure 6. The generated InSAR maps; (A) phase band, (B) coherence band, and (C) unwrapped band

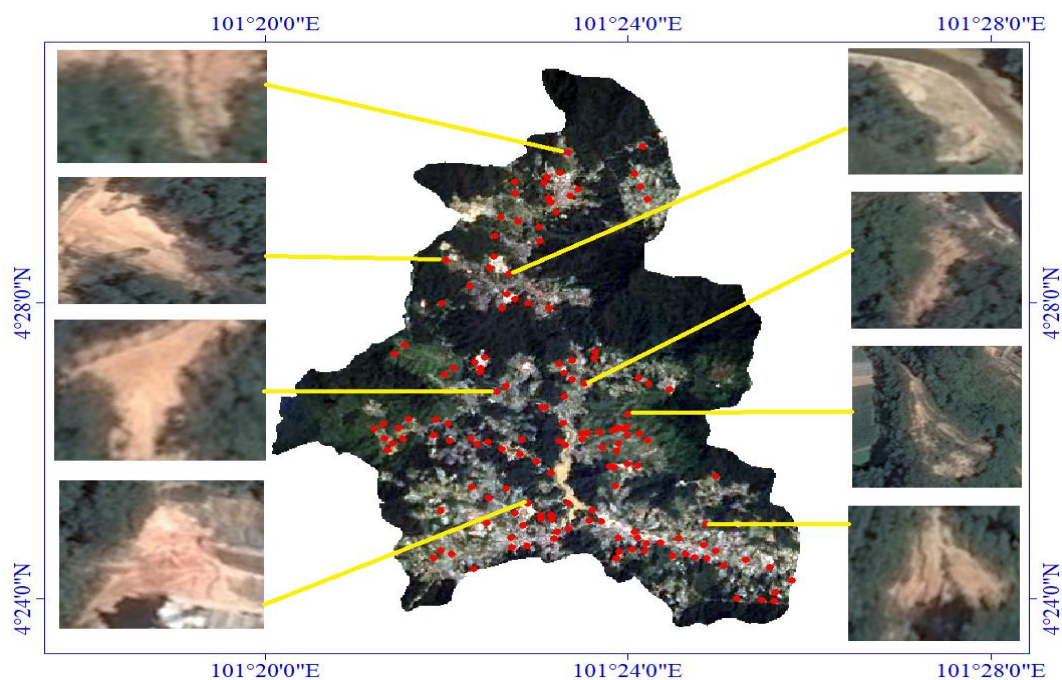


Figure 7. The importance of the Google Earth imagery in the produced landslide inventory

Table 2. Number of occurred landslides on the slope, aspect, and elevation classes

Slope classes (%)	No. occurred landslide	Aspect classes	No. occurred landslide	Elevation classes (M)	No. occurred landslide
0 - 10	11	Flat	0	920.1 - 1,077	21
11 - 20	37	North	22	1,078 - 1,167	27
21 - 30	57	Northeast	21	1,168 - 1,253	47
31 - 40	35	East	23	1,254 - 1,338	22
41 - 50	10	Southeast	32	1,339 - 1,424	5
51 - 60	1	South	16	1,425 - 1,500	11
61 - 70	0	Southwest	18	1,501 - 1,572	16
71 - 84	1	West	5	1,573 - 1,658	3
-----	-	Northwest	15	1,659 - 1,770	0
-----	-	-----	-	1,771 - 2,067	0

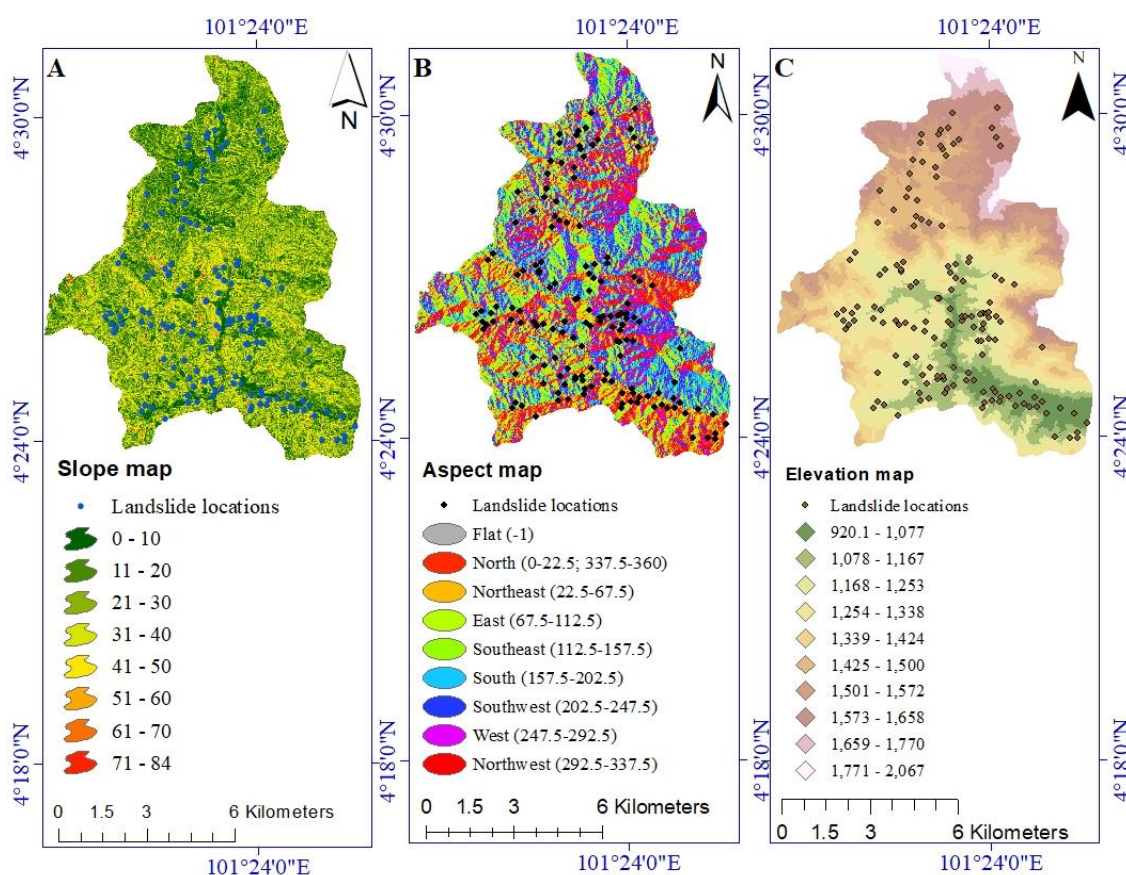


Figure 8. Occurred landslide locations; slope (A), aspect (B), and elevation (C) maps

The approach used in this study differs from previous inventory works in a number of ways:

- Convergence of satellite imagery, the Google Earth and extensive field survey

- Applying field survey to check 20% (30 landslide locations) of detected landslides despite limited field access
- Using SENTINEL-1 satellite imagery as the only RADAR imagery online for free

Validation

The Landslide locations were validated using GPS in the study area. However, 30 landslide locations (20%) out of the total number of 152 locations were transmitted into GPS and validated in the study area. The field checks imply that the landslides were correctly mapped, therefore integration method used in the current research is applicable and valuable to detect landslide in a vegetated and cloudy region. *Figure 9* indicates the geographical position of the validated landslides. However, for better understanding we have attached a few photos of the validated landslides in the figure. From the total number of 152 landslide locations, 64 landslides have an area larger than 500 m² and the remaining 88 landslides have an area smaller than 500 m². However, the smallest and the largest landslides have areas of 40 m² and 5735 m², respectively.

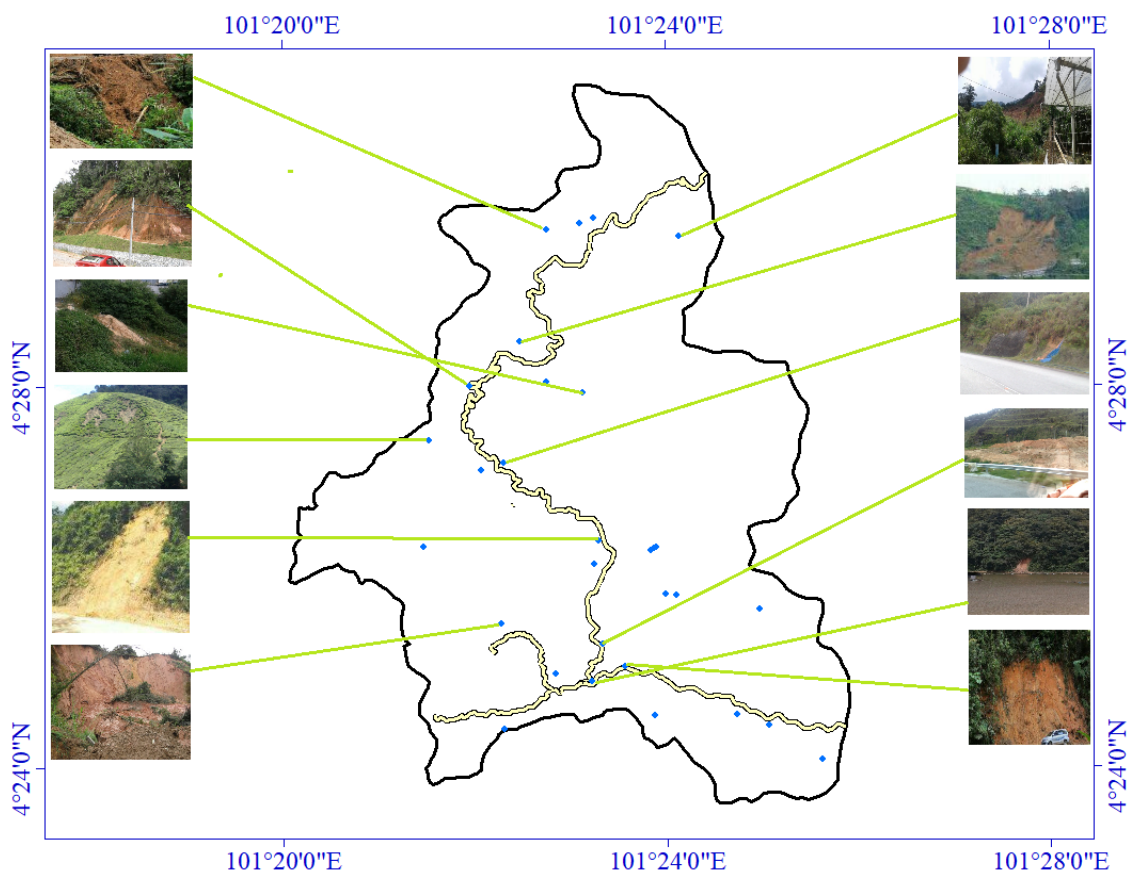


Figure 9. Shows the validated landslide locations using GPS

Table 3 lists the area and attributes of landslide locations, which were used for validation purposes in this study. These locations were randomly selected for field checks, however, the biggest and the smallest landslides have the areas of 2625 m² and 78 m², respectively.

Table 3. Lists the validated landslide locations in the study area using GPS

No.	Area_m ²	X coordinate	Y coordinates
1	332	761669.703	491002.116
2	232	763152.498	489538.965
3	365	763213.396	487495.887
4	135	761762.287	493068.394
5	930	763183.593	492653.848
6	601	762771.802	492493.392
7	1278	766107.429	487775.765
8	369	767715.562	487797.997
9	2625	769429.086	486872.411
10	163	768311.418	487589.843
11	450	764924.011	497368.124
12	101	764018.098	497145.497
13	78	766583.302	497033.463
14	270	764710.241	494005.554
15	1157	764023.307	494201.886
16	343	764662.954	497256.227
17	323	763499.397	494990.104
18	506	764943.555	490694.459
19	229	766132.101	491018.098
20	287	766095.916	490985.837
21	454	762522.67	494101.393
22	135	766032.607	490970.753
23	1012	764210.016	488554.45
24	626	764897.98	488425.154
25	240	765555.957	488728.318
26	490	765022.494	491144.831
27	1744	766337.317	490121.71
28	1044	766529.743	490088.773
29	523	768156.935	489822.915
30	630	765132.229	489079.609

Conclusion and recommendation

Cameron Highlands are among the most landslide prone areas across Peninsular Malaysia. This work represents the integration of interferometry synthetic aperture RADAR (InSAR) technique with assist of the Google Earth image to identify the surface deformation of large and small landslides in a part of Cameron Highlands, Pahang, Malaysia. A pair of C-band SENTINEL-1 imagery were acquired, pre-processed, and then phase, coherence and unwrapped bands were generated to recognize the surface deformation. In order to address some limitations of interferometry synthetic aperture RADAR processing in mountainous terrains, different contributions to the interferometric phase are required. The topography phase must be removed, and then spectral filtering should be applied on the created interferogram. Generated phase band is wrapped, however using the SNAPHU software was unwrapped.

By overlying the phase, coherence and unwrapped bands on the Google Earth the landslide locations were identified and digitized in both polygon and point layers. A total number of 152 landslides were identified. GPS was used to validate 20% of the total extracted landslide locations in the study area. SENTINEL-1 imagery with the ability of penetrating through cloud coverage and the applicability of getting data day and night in the cloudy regions like Cameron Highlands with assist of the Google Earth imagery supported by extensive field survey is a robust way for landslide inventory in the vegetated areas. Hence, the information provided by landslide inventory in the current study can be used for planners and engineers to make better decisions about landslide prevention, mitigation and avoidance through suitable prophylactic assessments and minimization procedures. We recommend more works for future researches to produce more detailed inventory map of occurred landslides in tropical environment like Cameron Highlands using L-band satellite imageries such as ALOS-PALSAR-2 with 24 cm wavelength, which is able to penetrate further into vegetation coverage rather than C-band imageries like SENTINEL-1 satellite image, in order to more accurate detection of occurred landslides in the same study area situation.

Acknowledgements. We express our thanks to Editor-in-Chief of applied ecology and environmental research journal and our anonymous reviewers. With their comments and suggestions, we were able to significantly improve the quality of our paper. Also, authors wish to express their sincere thanks to Universiti Teknologi Malaysia (UTM) based on Research University Grant (Q.J130000.2527.17H84) for their financial supports.

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