POST-DISASTER ASSESSMENT OF MANGROVE FOREST RECOVERY IN LAWAAAN-BALANGIGA, EASTERN SAMAR USING NDVI TIME SERIES ANALYSIS

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ABSTRACT:

In 2013, Typhoon Haiyan (Yolanda) struck the Eastern Philippines. Mangrove forests in the area were destroyed and were estimated to have at least 86% of damage. Some studies done on the typhoon-stricken mangroves had collected data such as measurements of mangrove trunk, height, roots, and seedlings to investigate the extent of damage and recovery. While these studies were proven to effectively identify mangrove gains and losses, these methods are only applicable in sites that are relatively accessible. This paper highlights the relevance of effective remote monitoring of mangrove forests that are vulnerable to typhoons including post-typhoon recovery. In this study, a Time Series Analysis using Google Earth Engine (GEE) was applied in assessing the damages and recovery of mangroves struck by Super Typhoon Haiyan in Lawaan and Balangiga, Samar (Eastern Philippines). The changes in mangrove extent followed the changes in NDVI; however, there were significant site-specific differences. Based on NDVI values, it was estimated that 83% of the mangrove area was damaged. After three years, regeneration from 2014-2017 was about 144%. Mangroves steadily developed but with a minimal change of 2.83% from 2017-2019. Most villages followed the general recovery trends in Lawaan and Balangiga. However, based on the time series analysis, some villages have minimal recovery than others. It suggests that the recovery of mangroves may be a function of the pre-typhoon mangrove extent and possibly vegetation condition. Even if there were new spaces for mangroves to colonize, some of the sites may not be conducive for plant regrowth.

1. INTRODUCTION

Mangrove forests are among the world’s most productive ecosystems. They often grow in tropical and subtropical latitudes where they thrive in regions with high salinity, high temperatures, extreme tides, and strong winds. (Bingham et al., 2001) Though exposed to such extremes, mangroves have developed several unique adaptations such as lateral roots (for anchoring in loose sediments), exposure of aerial roots (for gas exchange) (Bingham et al., 2001), and salt exclusion at the root level. (Perri et al., 2016)

The Philippines is one of the top mangrove biodiversity “hotspots” in the world with at least 35 mangrove species. (Long & Giri, 2011) But since the country is so typhoon-prone (Vergano, 2013), it continues to suffer mangrove losses. For the past decade, the country has already experienced two typhoons, one of which is Super Typhoon Haiyan (Yolanda) in 2013 (PanahonTV, 2017). It is one of the most catastrophic typhoons that struck the country and nearly wiped-out the mangrove areas in the central Visayas region.

Some post-typhoon mangrove studies were conducted in Eastern Samar and Palawan which were two provinces highly affected by Typhoon Haiyan in 2013. One study by Alura et al. (2016) focused on thriving mangrove species post-typhoon for implementation of rehabilitation technique and damage & recovery assessments conducted on Calauit Island, Palawan (Malabrigo et al., 2016) and Coron, Palawan & Balangiga-Lawaan (Buitre et al., 2019)

This research analyzed the mangrove forests of Balangiga-Lawaan and assessed the damage due to Typhoon Haiyan in 2013 and the subsequent recovery from 2014-2019 using Sentinel-2 and Landsat imageries. These were carried out by creating a Time Series Analysis and Ordinary Least Squares (OLS) Regression estimates of the slope of Normalized Difference Vegetation Index (NDVI) and Leaf Area Index (LAI).

Landsat-8 OLI/TIRS images in the Google Earth Engine (GEE) scripts were used for the initial damage and recovery assessment. Sentinel 2A and 2B derivative datasets (i.e., NDVI, LAI) were also used and assessed using Excel and SPSS (Statistical Packaging for the Social Sciences) to understand the changes in the mangrove areas. The researchers encountered some constraints during the implementation of this study due to the ongoing COVID-19 pandemic (e.g., limited access to ground data collection). The entire study was limited to datasets derived from GEE and other online resources. Small deviations in the values from the computations may be attributed to the errors brought by the calibration of the satellite images.

2. REVIEW OF RELATED LITERATURE

2.1 Mangrove Forests & Typhoon Haiyan

The typhoon Haiyan (Yolanda) has had six (6) landfalls - affecting the areas of Samar, Leyte, Cebu, Iloilo, and Palawan (Mangosing, 2013). The extent of the destruction was assessed and measured in some sites. Published studies have been recorded like Salmo’s and Gianan’s (2019) study on the mangroves of Salcedo, Samar with regards to the post-typhoon carbon stocks. Strong catastrophic typhoons like Haiyan are known to result in the uprooting of trees, breaking off branches, defoliation, and massive mortality leads to severe reductions in total carbon stock (TCS) of the area.

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A damage assessment on the mangroves of Calautit Island, Palawan was conducted by Malabrego, et al. (2016) from ten 10m x 10m sampling quadrats. All quadrants were evaluated and assigned to a specific damage status ranging from none to severely damaged. Study shows that the Calautit mangroves suffered damages with 20 % of the plots having 100 % mortality. Only 21% of the recorded trees were intact and presumed to be 100% recovered in shoot development.

A few species-specific studies were conducted like the study of Carlos et. al (2015) in Tacloban andOrmoc, Leyte. The research assessed three mangrove genera - Rhizophora, Sonneratia, and Avicennia. Each corresponding genus’ vegetation resistance (VR), and seedling regeneration potential (SPR) were assessed to know which of the three is the most resilient. Their results showed that both Sonneratia and Avicennia have higher natural regeneration rates due to success in seedling recruitment.

2.2 Normalized Difference Vegetation Index (NDVI)

Normalized Difference Vegetation Index (NDVI) is a vegetation index measured by getting the ratio between the difference and sum of Near Infrared (NIR) and Red channels of image.

\[ NDVI = \frac{NIR - RED}{NIR + RED} \]  

(1)

NDVI is used in agriculture, forestry, and environmental studies as a standard parameter for the identification of photosynthetically active or healthy vegetation. This ratio ranges from -1 (no vegetation) to +1 (healthy vegetation). Values close to the negative spectrum correspond to areas of water, built-up, and other man-made structures. Bare soil ranges from 0.1 to 0.2. Grasslands/sparse vegetation have NDVI values from 0.2 to 0.5, and dense vegetation has values above 0.5. (EOS.com, 2019)

In recent studies, NDVI was used to explore impacts of climate variability on mangrove ecosystems (Flores-Cárdenas et al., 2018) and even as a parameter for estimating mangrove health index (Akbar et al., 2020). One example of its reliability in mangrove mapping was a developed index called CMRI (Combined Mangrove Recognition Index) which is a combination of NDVI and Normalized Difference Water Index (NDWI) used to discriminate mangroves from non-mangroves in India. This index was tested for three (3) mangrove sites and had an average overall classification accuracy of 73%. (Gupta et al., 2018)

2.3 Leaf Area Index (LAI)

Leaf Area Index (LAI) is a biophysical parameter that refers to the number of square meters of leaves present in one square meter of ground. For most crops, LAI can range from 5 to 7 (m²/m²). Values from 3 to 4 usually refer to close canopy. (Airbus, 2020) Presence of leaves becomes an important indicator of radiation, precipitation interception, energy conversion, water balance, and a reliable parameter for plant growth. That is why most studies in precision agriculture and climate change use LAI as leaves are the most eco-physiological parts of the plant that interact with the atmosphere. (Trimble, 2019)

2.4 Mangrove Vegetation Index

A mangrove index was recently developed for rapid and accurate mangrove mapping specifically applied in the mangroves of the Philippines. Baloloy, et al. (2020) developed an equation that can delineate mangroves using NIR, SWIR, and green reflectance bands without the need for complex classification techniques which can be time-consuming and very user-skill-biased.

\[ MVI = \frac{NIR - GREEN}{SWIR - GREEN} \]  

(2)

The index is apt for the study area since it was designed by analyzing the spectral signatures and characteristics of mangroves and non-mangrove datasets from different sites in the Philippines and Japan. Accuracy assessments indicate that MVI on the average provides around 92% accuracy. Its threshold values were then computed through the use of a nationwide field-acquired mangrove inventory and drone data. According to Baloloy, et al. (2020), the optimal minimum threshold values that can discriminate mangroves from non-mangroves areas were about 4.5 and 4.6, respectively with mean MVI varying depending on the type of mangrove forest. The mean MVI threshold value was 6.7 for riverine mangrove forests (forests found in rivers) and 8.9 for fringe mangrove types (forests found in coastal areas).

3. DATA AND METHODS

3.1 Study Areas

The municipalities of Lawaan and Balangiga are situated in Eastern Samar within the bounding coordinates Latitude: 11° 19’ 0.62” N to 11° 4’ 54.95” N and Longitude: 125° 14’ 52.45” E to 125° 26’ 20.63” E (Figure 1). Both areas have villages near the Philippine Sea (East) and the Leyte Gulf (South). The mangrove areas studied were located on the coastal fringes of Bolusao, Guinoban, Beta-og, Poblacion I,II,II, Taguig, Maslog in Lawaan, and San Miguel, Poblacion V, Cansumangcay, Bacio in Balangiga.
3.2 Time series analysis using Google Earth Engine and LAI processing ESA-SNAP

3.2.1 Mangrove Forest Pixels

Normalized Difference Vegetation Index (NDVI) and a newly developed mangrove vegetation index (MVI) (Baloloy, et al. 2020) were added to the list of spectral bands present on Landsat 8 images of the study area. NDVI and MVI were calculated using the red, green, SWIR, and NIR bands. The equations are as follows:

Landsat 8: \( \text{NDVI} = \frac{\text{B}_5 - \text{B}_4}{\text{B}_5 + \text{B}_4} \) (3)

Landsat 8: \( \text{MVI} = \frac{\text{B}_5 - \text{B}_3}{\text{B}_6 - \text{B}_3} \) (4)

A mangrove mask function was created to separate the mangrove from the non-mangrove area. Using MVI bands per image, the range of values was set to 4.5, the optimal minimum threshold (Baloloy, et al., 2020) up to a maximum of 20. Pixels within this range are mangrove areas.

3.2.2 Time Series Analysis of the mangrove area

Time series of the NDVI values of the mangroves in Lawaan and Balangiga, Eastern Samar was made. The time series shows the general trend of the mangrove growth in the area. The decline of NDVI indicates the loss and the increase indicates the gain or recovery. For each month from January 2013 to December 2019, the mean NDVI of the mangroves was transferred to a new image collection to create the general time-series graph. Time series analysis was also performed in each village to compare the trends of mangrove recovery.

3.2.3 Ordinary Least Squares Regression Analysis

Google Earth Engine’s Ordinary Least Squares Regression (OLS) was used to estimate the slopes of the growth trend of mangroves each year. The slope indicates the rate of change of mangrove growth or loss in the study area. The output shapefile containing the mangrove extent of Lawaan and Balangiga in the year 2013 was used as the area of interest (AOI) for the regression analysis. The output of the OLS Regression script was used to confirm the rate at which the mangrove areas change over time.
3.2.4 Sentinel Data Collection and Processing

Sentinel 2A and 2B satellite images were downloaded from EarthExplorer (https://earthexplorer.usgs.gov/) and prepared using the ESA-SNAP. The images were adjusted and fine-tuned using the Sen2Cor plugin of ESA-SNAP. This calibrates the satellite images and creates cloud masks. Sentinel 2A and 2B images have different spatial resolution per bands, therefore the images were resampled to 10 meters and then were subset to the study area.

3.2.5 Thematic Land Processing

Thematic Land Processing was done to compute the NDVI values of each image per month. The outputs of this processor were biophysical parameters used to assess the mangrove growth further. Leaf area index (LAI) was mainly used in this study. The researchers used LAI values greater than 3.0 to indicate mangroves that have reached closed canopy.

4. RESULTS AND DISCUSSIONS

4.1 NDVI time series analysis in Lawaan-Balangiga mangrove area

![Figure 4. NDVI Time series chart of Lawaan-Balangiga mangroves](image)

Based on the resulting time series chart (Figure 4), there has been a significant drop in NDVI immediately after Typhoon Haiyan. Values gradually increased starting in 2014 until they plateaued by 2017 indicating post-typhoon vegetation recovery (Figure 5). The general NDVI time series charts of the mangrove areas in Lawaan-Balangiga were divided into three different sections, namely (a) Damage Period, (b) Recovering Period, and (c) Stabilizing Period.

![Figure 5. 2013 to Early 2014 NDVI Time Series Chart (Damaged Period)](image)

The Damaged Period is from 2013-2014 in which the mean NDVI values dropped from October (pre-typhoon) and November 2013. The slope of the red trend line in the figure has a value of -17.48. The absolute value of the slope describes the sudden drop of the NDVI value. The NDVI value drop is indicative of the significant mangrove loss in the area, estimated to be around 83.07%. Around January 2014, the NDVI values started to increase. (Figure 5)

![Figure 6. 2014 to Early 2017 NDVI Time Series Chart (Recovering Period)](image)

The trend of the Recovering Period from 2014 to early 2017 has a value of 1.34. Mangrove area change was observed to be 114.30%. Compared to the absolute slope of the Damaged Period, this has a smaller value indicating the rapid and steady growth of mangroves. (Figure 6)

![Figure 7. 2017-2019 NDVI Time Series Chart (Stabilizing Period)](image)

The Stabilizing Period from 2017-2019 has a slope value of 0.093. During this period, the mangroves steadily grew with a mangrove change of 2.83%. The mean NDVI values of the mangroves during this period stabilized and stayed around 0.8. This shows that mangroves steadily recovered and most of it were retained up to 2019 (Figure 7).

The trends of NDVI per village followed the general growth patterns (Figures 8-9). This means that the general trend of mangrove growth shown in Figure 4 represents the mangrove growth in most villages. The villages of Taguite, Maslog, and San Miguel, not only have the largest mangrove cover in these two municipalities, but they also have closely similar trends during mid-2014. The NDVI started to exhibit signs of positive change in mangrove growth, although they had the greatest mangrove losses.
In contrast, smaller villages like Poblacion V and VI don’t follow the trends of larger villages. Poblacion V NDVI values stayed around 0.75 while Poblacion VI deviated from 0.2 to 0.4 level. Some mangroves in Poblacion VI were not able to adapt and recover based on its NDVI time-series graph. The growth may be a function of the area. Even though there were new sites for mangrove to colonize, some mangroves were not able to recover.

4.2 Ordinary Least Square (OLS) Slope Estimates for Growth Trend

The OLS slope estimates were also able to quantify the rate of area change in the mangrove forests of Lawaan and Balangiga. Figures 10-11 shows that in the municipalities of Lawaan-Balangiga, the rates of change of the NDVI values varied depending on their location (if the barangay is more eastward) and the extent of the mangrove area (how large/small the villages are). The villages of San Miguel, Taguite, and Maslog were the three areas with the highest decline in the growth trend slope, which means that these areas had greater damage compared to other villages (Figures 8-9). It is evident that these areas had more mangrove losses mainly because they have vast mangrove forest than other villages.

Figure 8. NDVI time series charts showing the general trend of mangrove growth for each coastal barangay of Lawaan (a) Bolusao, (b) Guinob-an, (c) Beta-og, (d) Poblacion I, (e) Poblacion II, (f) Taguite, (g) Maslog

Figure 9. NDVI time series charts for each coastal barangay of Balangiga (a) San Miguel, (b) Poblacion V, (c) Poblacion VI, (d) Cansumangcay, (e) Bacjao

Figure 10. Bar graph showing the comparison of slopes per barangay in Lawaan

Figure 11. Bar graph showing the comparison of slopes per barangay in Balangiga
### 4.3 Mangrove Area Change Maps Using MVI and Sentinel 2-derived NDVI images

Annual and 2013-2019 mangrove area change maps (see Figure 12) were created using MVI and NDVI in QGIS. NDVI values and MVI-detected mangrove pixels were used as inputs for the maps. These maps show the immensity of the mangrove loss vs. the changes that happened (i.e., recovery, retention) post-Haiyan. “Lost” areas (red) show mangrove forests decimated by Typhoon Haiyan (approximately 494.85 has) and additional mangrove areas lost thereafter until 2019 (around 54.86 has). “Retained” areas (yellow; estimated at 217.93 has) are mangroves which remained after Haiyan until 2019. As shown in Figure 12, there are minimal “gained” areas (green; ca. 23.35 hectares).

Between 2013 and 2019, the initial mangrove areas from the start of the analysis have grown back based on the areas tagged as gained (green) and retained (yellow) in the change map (Figure 13). Though there has been a recovery period from 2014-2017 which then stabilized until 2019, loss in the mangrove area in the study region is still evident. Not only in the 2013 vs. 2019 change map was this observed but also, mangrove areas that seemed to survive the typhoon as seen on the 2013-2014 change maps, eventually died in the succeeding years. Furthermore, continuous degradation in mangrove plants in some regions of the study area are still visible which can be seen in Figure 13 even though rehabilitation efforts were conducted.

*Figure 12. 2013-2019 Mangrove Area Change Map*

*Figure 13. Damaged Period (2013-2014), Recovery Period (2014-2017) & Stabilizing Period (2017-2019) Mangrove Change Maps*
4.4 Mangrove Canopy Maps

Mangrove canopy area maps were produced. These show mangrove areas with NDVI values 0.6-1.0 and LAI values 3.0-4.0, referring to healthy mangroves that have reached closed canopy. These areas generally increased from 606.2 ha in April 2017 (Figure 14a) to 1047.3 ha in June 2019 (Figure 14b). This positive change in area based on NDVI and LAI confirms that the mangroves in Lawaan and Balangiga continued to grow from 2017 to 2019.

![Figure 14. Mangrove canopy area maps (a) APRIL 2017 (b) JUNE 2019](image)

5. CONCLUSIONS

This research was intended to investigate the (a) mangrove damage brought by Typhoon Haiyan to Lawaan-Balangiga, Eastern Samar in November 2013, and the (b) mangrove growth and recovery six years post-disaster from 2014 to 2019. The study documented general rapid mangrove recovery but stabilized 4-5 years post-typhoon Haiyan. However, the recurrence of mangrove losses from the initial recoveries may indicate a possible lag effect that could be related to slow vegetation recovery and possibly still inhospitable sediment conditions. Moreover, the recovery varied with sites that were probably related to the extent of pre-typhoon mangroves. Although the NDVI and LAI values are correlated to post-typhoon recovery, its fluctuations at different sites and at different post-typhoon periods may indicate the role of localized site-specific factors (e.g., mangrove extent and possibly species composition) in post-typhoon recovery.

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