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Modeling the Impact of Land Cover Changes on Soil Erosion Estimation in Yewa North and Yewa South, Ogun State, Nigeria

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ABSTRACT

Soil erosion is becoming a serious problem in some communities in Yewa North and Yewa South because of rapid land use developments. This study was carried out to analyses the landcover change effects on soil erosion to determine the extent and trend of changes in the study area; estimate and characterize soil loss, and comparing the annual soil loss at different spatial scales. The extent and trend of changes in the landcover were estimated using Landsat Satellite Imageries for the year 2002 and 2017. RUSLE model was used to estimate soil loss and was characterized based on the expert description for tolerable soil loss concept. The results revealed the built-up area showed a consistent increase over time, from 349.5km^2 in 2002 to 592km^2 in 2017 of the total area. The vegetation covered about 933.4km^2 in 2002 but decreased to 509km^2 in 2017. Similarly, the area covered by bare ground increased from 243.7km^2 in 2002 to 620.4 km^2 in 2017 but the waterbody increased from 0.29km^2 in 2002 to 0.72km^2 in 2017. The areas covered by agriculture also decreased from 1088.5km^2 in 2002 to 894km^2 in 2017. The estimated soil loss values ranged from $0 - 420,276 \text{ t} \text{ ha}^{-1} \text{ yr}^{-1}$ with the mean of 231 and standard deviation 2272 in

2002 while the soil loss estimated in 2017 ranged from 0 - 186,920 t ha/ yr. with the mean of 220 and the standard deviation 1312.3. Comparatively, low erosion is observed in a total area of 45.5% in 2002 and 44.4% in 2017 while extremely severe erosion is observed in a total area of 29.8% in 2002 and 34.3% in 2017 in the study area which matches the actual bareground and agricultural land which can be attributed to change in C and P factors. The study recommended that enlightenment and awareness of erosion control should include land use habit of the people in the agricultural practice and care of vegetation

Keyword: Urbanization, Land use Land covers (LULC), Landsat Satellite Imageries, RUSLE model.

1. INTRODUCTION

Land use and land cover changes influence numerous natural procedures, for example, soil erosion, sediment creation, flood, and physical and chemical properties of the soil (Martínez-Casasnovas JA, Sánchez-Bosch 2000; Yang D, Kanae S, Oki T, Koike T, Musiake K (2003); Ochoa P, *et al.* (2016); Zare M, *et al* (2017). The land use and cover changes in urban regions is a progressive procedure however a few elements can impact this procedure, most particularly human activities. Urbanization is one of the most groundbreaking human land-use forms in history and is probably going to increment in the coming years. It is commensurate to a close perpetual adjustment of the common habitat and it wipes out the locally prevailing regular environment. Urbanization enormously affects common natural surroundings and assets, and urban development is liable for anthropogenic exercises, for example, infringement of horticultural exercises on woods zones, deforestation for business and modern purposes, and general abuse of the land. These exercises change the landuse from its unique vegetation cover to a state of lesser vegetation cover, bringing about soil disintegration (Cebecauer and Hofierka 2008). Urban development because of expanding populace and human exercises, for example, agriculture, deforestation, construction, grazing, and sand mining were distinguished as components that have upgraded the improvement of chasms and this thusly has prompted the devastation of building establishment and correspondence course and arrangement of barren wilderness in the region. Soil erosion is a typical common occurrence that can be ascribed to specific geology, soils, and relating vegetation that prevail and the broad precipitation that the locale encounters.

In Nigeria, the Soil erosion issue is a worldwide phenomenon influencing both the biotic and socio-economy of the encompassing individuals (Ofomata, 1984). One of the anthropogenic variables affecting soil erosion is landuse change. In such a manner, much work has been done to consider the effect of landuse change on soil erosion in different areas (Feng, X *et al*, 2010; Ranzi, R.; Le, T.H.; Rulli, M.C.2012; Conforti, M.; Buttafuoco, G 2017). In Yewa environs, the phenomenon of soil erosion is likewise boundless. It happened in various Communities which can be affected by quantities of variables both characteristic and anthropogenic and satisfactory records of the soil erosion are not accessible. Evaluating soil erosion is hard to execute because the

customary overview and planning procedures are costly and tedious for the estimation of land cover changes and soil erosion. Therefore, increased research intrigue is being coordinated to the checking of urban development and soil erosion utilizing GIS and far off detecting procedures (Epstein et al., 2002). Distinctive erosion models have been created in the earlier decades, utilizing different diverse scientific techniques and modeling approaches is essential to the national turn of events. (Isikwue, Abutu, and Onoja, 2012). The integration of remote sensing and geographic information system (GIS) and modeling are significant apparatuses for planning peril and hazard. In this investigation, the Revised Universal Soil Loss Equation (RUSLE) has been incorporated with remote sensing and geographic information system (GIS) to survey soil erosion, for sufficient data on soil erosion that help in legitimate control for land and water use arranging. The combined utilization of GIS and erosion models has been demonstrated to be a successful way to deal with assessing spatial dissemination of erosion

The study, therefore, seeks to analyses the landcover change effects on the soil erosion; by determining the extent and trend of landcover changes in the study area; estimate and characterize soil loss, and comparing the annual soil loss. The study aimed to provide information for effective soil erosion prevention and water conservation planning.

2. METHOD AND MATERIAL

2.1 The Study Area

The study seeks to analyses the landcover change effects on soil erosion in Yewa North and south. Yewa North and south are located in Ogun State, Southwestern Nigeria. The study area occupies a land area of 2684.3 km² and a population of 350,676. The study area is situated between Latitude $6^{\circ} 40^{\circ N}$ and $7^{\circ} 25^{\circ N}$ Longitude $2^{\circ} 40^{\circ E}$ and $3^{\circ}5^{\circ E}$ in the southwestern part of the State with the annual rainfall varying between 105cm to 128cm. The average monthly temperature ranges from 23°C in July to 32°C in February. The State has two main types of vegetation viz-tropical rain forest and Guinea savanna. The tropical rain forest is found in the coastal areas of the southern part of Yewa South LGA while the rest are Guinea savanna



Figure 1: The study area

2.1.1 Data and sources

This investigation included both primary and secondary sources of data (Table .1). The primary data includes visiting the study area to acquire the point of erosion area while the secondary include an administrative map of Nigeria, precipitation data, topographic data, soil data, and landuse data

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S/N	Source	Year	Characteristics	Relevance
1	United State Geological Survey	2002	(Path 191 and row 55)	Landuse to generate C, P
	(USGS)	2017	Resolution 30m	factor
2	Tropical Rainfall Measuring	1998-	20 years, Resolution:	To generate (R factor)
	Mission (TRIMM)	2017	resampled to 30m	
3	Shuttle Radar Topographic Mission (SRTM}	-	30m resolution	To generate LS factor
4	Nigerian Building and Road	2015	Resolution: resampled to	To generate Soil Erodibility
	Research Institute (NBRRI)		30m	Factor
5	Global Positional System (GPS)	2018	Landuse references and points of erosion area.	For training site

Table 1: Datasets, Characteristics, Sources, and Relevance

2.2 Data Processing and Analysis

2.2.1 The extent and trend of land cover changes in the study area

Multi-temporal Landsat images; 2002 and 2017 were processed with ENVI 5.1 and the image was enhanced for visualization purposes, the study area was extracted out of the full scene. The Maximum Likelihood Classifier was used to produce different periods of land cover (LC) maps, after which it was embellished using Arc Map 10.4. The landcover classes were used to generate the cover management map and support practice.

2.2.2 Soil Loss Estimation

Revived Universal Soil Loss Equation (RUSLE) was adopted to estimate soil loss. The model is widely used in predicting soil erosion because it can be easily parameterizing The RUSLE model can be calculated using equation 1 (Wischmeier and Smith, 1978; Renard *et al.*, 1997)

2.2.2.1 The Rainfall Erosivity (R factor)

The daily rainfall data from 1998 - 2002 and 2013 - 2017 acquired were entered into excel creating a database of average annual rainfall to generate the rainfall erosivity. The rainfall erosivity was calculated using Lee and Lee equation as shown in Table 2

Table 2: Equation for RUSLE parameter

Equation	Parameters	Source
$R = 38.5 + 0.35 \times Pr$	Rainfall Erosivity	Lee and Lee, 2006
$K = 7.594 * (0.0034 + 0.0405 * \exp[^{-0.5} * (\log (Dg) + 1.659)]$	Soil erodibility	Renard et al. 1997
$(0.7101)^2$]		
$Dg (mm) = exp^{0.01*} sum(fi \ln mc + fi \ln msi + fi \ln msa)$	DG	Renard et al. 1997
$LS = \left(\frac{FA \times CS}{22.13}\right)m \times [0.065 + 0.045(S) + 0.0065(S^2)]$	LS factor	Bizwuerk et al. (2008)

2.2.2.2 Soil erodibility (K) factor

The soil data collected were entered into excels creating a database that includes percentages of sand, clay, silt, and soil types. Each soil type based on textural classification of the United States Department of Agriculture classification (USDA) were assigned with erodibility value using geometric mean particle diameter (Dg). A new field was created in the soil type attribute table titled K factor value. The soil type map was converted from polygon to raster using the K factor value as a value field to produce the K factor map. The equation from Renard et al. 1997 was used to compote the K factor as shown in table 2

2.2.2.3 Estimating LS factor

A Digital Elevation Model (DEM) of 30m resolution from the Shuttle Radar Topographic Mission (SRTM) was downloaded. The DEM was imported into ArcGIS software to generate the different raster layer which includes; slope gradient, flow direction, and flow accumulation to generate slope length and slope steepness (LS factor)

2.3.1.3 Cover Management Factor (C) and conservation practice (P) factor

The cover management factor (C) and conservation practice (P) factor values were derived for each type of landcover classes based on the different expert description (Table 3). A new field was created in the landuse attribute table titled C factor value and P factor value. The landuse map was converted from polygon to raster using the C factor value and P factor value as the value field to produce C factor map and P factor map. The soil loss was estimated by multiplying the entire RUSLE parameter utilizing the Raster Calculator. The result generated was reclassed to improve visualization.

Table	Table 5 C and P factor values for each land cover classes						
S/N	Landcover classes	Cfactor	Pfactor	source			
1	Agricultural land	0.3	0.40	Lee & Lee 2006			
2	Bare ground	0.35	0.45	Jung (2004)			
3	Built-up land	0.1	1	Jung (2004)			
4	Vegetation	0.03	0.1	Ministry of forestry, 1985			
5	Waterbody	0	0.5	Lee & Lee 2006			

T-11-2 C and D faster values for each land seven aloges

Source: Soo (2008)

3. RESULT AND DISCUSSION

3.1 The extent and trend of landcover changes in the study area

Five classes of landcover were identified from satellite imagery classification. The result of the accuracy assessment of the image classified shows that the overall accuracy is 82% and the kappa coefficient is 0.76. The result indicates that 76% of image classification agreed with the reference data. Landis and Koch (1977) represented the value of accuracy to be (0.61 to 0.80). Based on this author's suggestion, the strength of agreement is thereby categorized as good. The producer's accuracy and user's accuracy of each class are shown in Table 4. High User's accuracy was recorded in waterbody (97%) built-up land (93%), agricultural land (71%), and forest 73% respectively while low User's accuracy was seen in the bare ground (54%). Low accuracy in the bare ground may be as a result of several problems in the interpretation of land use and confusion among other landuse types.

The results reveals that in the periods 2002, 2017, the spatial extent of built-up area showed a consistent increase over time, from 349.5km² (13.35%) in 2002 to 592km² (22.6%) in 2017 of the total area. The vegetation decreased from 933.4km² (35.7%) in 2002 to 509km^2 (19.45%) in 2017. Similarly, the area covered by water body increased from 0.29 km² (0.01%) in 2002 to 0.72 km² (0.02%) in 2017; while the areas covered with bare ground increased from 244km² (9.32%) to 620.4 km² (23.72%) in 2017 and agricultural land decreased from 1088.5km² (41.62%) in 2002 to 894km² (34.2%) in 2017

	2002		2017		
Class Name	Area km ²	Area%	Area km ²	Area%	Trend
Builtup Area	349.5	13.35	592	22.6	9.25
Agricultural land	1089	41.62	894	34.2	-7.42
Vegetation	933.4	35.7	509	19.45	-16.25
Bare ground	244	9.32	620.4	23.72	14.4
Waterbody	0.29	0.01	0.72	0.03	0.02
	2616.19	100	2616.12	100	

Table 4: Distribution and spatial extent of landcover in 2002 and 2017



Figure 2: Landcover classes for 2002 and 2017

3.2 RUSLE Factors

3.2.1 The rainfall erosivity (R factor)

Table 5 shows the estimated average annual rainfall value ranges from 2951.98 -3161.36 in 2002 while the average annual rainfall value for 2017 ranges from 2937.98 – 3200.44. The R factor values estimated for 2002 range from 1071.69 – 1140.79 $MJ/mm\&ha^1 hr.^1$ /year with the mean of 1098.6 while R factor value estimated ranges from 1066.79 to 1151.07 MJ/mm&ha1 hr.1/year with the mean of 1098.6 as shown in Figure 3.

Table 5: Average a	nnual rainfall
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	0	
point	2002	2017
	Ave.annual Rainfall (mm)	Ave.annual Rainfall (mm)
801	3161.36	3200.44
802	3075.5	3148.36
848	3128.08	3030.24
849	3147.84	3015.62
895	3002.78	2980.44
896	2951.98	2937.98





3.2.2 Soil erodibility (K Factor)

The USDA textural classification of soil in the study area comprises loamy fine sand, fine sand, sandy clay loam, clay loam, clay, and loam as in Figure 6. The values obtained for K factor for each sample are shown in Figure 7 where K value ranges from 0.02 to 0.19 t ha-¹ h-¹ MJ mm-¹ with a total mean of 0.10 and the standard deviation was 0.03. The result also reveals that higher erodibility values is associated with soils having high silt content and lower value of K factor is associated with the soils having low permeability, low antecedent moisture content.



Figure 4: Soil types and Soil erodibility (K factor)

3.2.3 Estimating LS factor

The LS factor provides better identification of problematic areas that will affect soil erosion. Figure 5 shows the different raster layers for slope length and steepness (LS) factor. The LS factor value ranges from 0 to 157.8 with a mean and standard deviation of 0.29 and 1.3 respectively. The estimated LS result shows that 91.7% of the total study area has a value greater than 15% while 8.3% of the total area falls under the LS values that is less than 15% as shown in Table 7 and Figure 6. The highest values corresponded with the greater slope while the lowest value corresponded to the study area lowlands. The average LS value is 0.29 which implies that study area land belongs to a low-risk class (0 - 2 unit) based on the grading standards of Manrique (1988).



Figure 5: Different raster layer for slope length and steepness (LS) factor



Figure 6: LS factor map

	1 0	1 0	· · ·	
Slop	e% LS	value	Areakm2	Area%
0 - 5	6 0 -	0.25	1458	55.9
5 – 1	5 0.2	.5 - 1.2	935.13	35.8
15 -	35 1.2	2 - 4.25	188.11	7.2
35 -	45 4.2	.5 - 7.5	15	0.6
45-5	0 7.5	- 12	6.4	0.3
>50	>1	2	5.5	0.2
			2608.14	100

Table 6: Slope gradient and Slope Length Index (LS)

3.2.4 Cover Management Factor (C)

The cover management factor (C) values range from 0 - 0.35 with a mean of 2.38 and 2.98, and a standard deviation of 1.29 and 1.33. Table 7 and Figure 8 show the Spatial distribution of Cfactor values. Higher values of C factor imply no cover effect and soil loss similar to that from a tilled bare fallow, while lower C means a very strong cover effect indicating no erosion effect (Erencin, 2000). The result reveals that about 35.7% of the total area has a strong cover effect indicating no erosion. The study area is mostly covered by agriculture activities (41.62% and 34.2% respectively) which are the direct relation between the C factor and soil loss

Table 7: Distribution and spatial extent of C factor

Class Name	2002	2017			Cfactor
	Area km ²	Area%	Area km ²	Area%	
Builtup Area	349.5	13.35	592	22.6	0.1
Agricultural land	1089	41.62	894	34.2	0.3
Vegetation	933.4	35.7	509	19.45	0.03
Bare ground	244	9.32	620.4	23.72	0.35
Waterbody	0.29	0.01	0.72	0.03	0
	2616.19	100	2616.12	100	

Source: Authors Analysis, (2019)



Figure 8. Cover Management Factor (C)

3.2.5 Support practice factor (P)

The support practice factor was estimated based on the different Land cover as shown in Table 8. The P factor values range from 0.1 - 1 with a mean of 2.13 in 2002 and 2.98 in 2017, and a standard deviation of 1.29 in 2002 and 1.33 in 2017 as shown in Figure 9. The lower the p-value, the more the effect the conservation.

Table 8: Distribution and spatial extent of P factor						
Class Name	Name 2002 2017			Pfactor		
	Area km ²	Area%	Area km ²	Area%		
Builtup Area	349.5	13.35	592	22.6	1	
Agricultural land	1089	41.62	894	34.2	0.4	
Vegetation	933.4	35.7	509	19.45	0.1	
Bare ground	244	9.32	620.4	23.72	0.65	
Waterbody	0.29	0.01	0.72	0.03	0.5	



Figure 9: Support practice factor (P)

3.3 Estimation of Soil Loss and extent of soil loss

Figure 10 shows the estimation of soil loss using Revived universal soil loss equation (RUSLE) and the values range from 0 - 186,920 t ha/ yr. with the mean of 220 and the standard deviation 1312.3 in the year 2017 while the soil loss estimated in 2002 ranges from 0 - 420,276 t ha-¹ yr-¹ with the mean of 231 and standard deviation 2272. Based on the results of the RUSLE model, the study area was divided into 5 classes in terms of soil erosion in two study years and it indicates soil sensitivity to erosion. The soil loss from 0-25 (t ha-¹ yr.-¹) erosion risk is characterized "Low", for 25 - 50 (t ha-¹ yr.-¹) "moderate", for 50 -100 (t ha-¹ yr.-¹)" severe", for 100 - 200 (t ha-¹ yr.-¹)" very severe", for soil loss >200 (t ha-¹ yr.-¹)" extremely severe " erosion risk (Table 9). It was observed that in the year 2002, 45.5% of the total area falls under low soil loss, followed by 29.8% of the total area that comes under extremely severe soil loss. The soil loss classes of moderate, severe and very severe covers 8.2%, 8.6% and 7.9% of the total area falls under the low soil loss, followed by 34.3% of the total area that comes under extremely severe soil loss and soil loss classes of moderate, severe and very severe covers 5.2%, 6.4% and 9.7% of the total area respectively.

	()	· //			
Soil erosion potential	2002		2017		
(t ha-1 year-1)	Area (km ²)	Area (%)	Area (km ²)	Area (%)	Changes rate (%)
Low (0-25)	1187.6	45.5	1104.4	44.4	-1.1
Moderate (25-50)	212.6	8.2	130.2	5.2	-3
Severe (50-100)	225.2	8.6	159.2	6.4	-2.2
Very severe (100-200	207.1	7.9	240.6	9.7	1.8
Extremely severe (>200)	776.1	29.8	851.1	34.3	4.5
Extremely severe (>200)	776.1	29.8	851.1	34.3	4.5

Table 9: Erosion Risk level based on (FAO) in (1979),

Source; Authors Analysis, (2019)





3.4 Discussion

From 2002 to 2017 periods witnessed significant urban expansion, this indicates a remarkable increase in the urban built-up areas, up to twice what was obtained at the beginning of the period. This rapid rate of urban expansion has been found to have encroached greatly into other urban landcover. For instance, as the population grew in the study area, agricultural land, and vegetation began to decrease to support the increasing need for land to support residential, commercial, and institutional development. This result corresponded with Ramachandran, 2001; Mary and Raghavaswamy, 2000 and Madhavi *et al.*, 2004 investigation which demonstrated that the city has been seeing fast development in the urban populace somewhere in the range of 1981 and 1999. The result from soil erodibility that range from 0.1 to 0.19 agrees with Renard *et al.*, (1991) report where K values can range from about 0.10 to 0.45-ton acre⁻¹ U.S customary units (0.01 to 0.5-ton ha h ha⁻¹ MJ⁻¹ mm⁻¹ SI metric units

The study also shows that there is an increase in the rate of soil loss in the study area. The mean rate of soil loss in 2002 is 231.88 t ha-¹ yr.-¹. and 219.79 t ha-¹ yr.-¹. in 2017. Comparatively, low erosion is observed in a total area of 45.5% in 2002 and 44.4% in 2017 while extremely severe erosion is observed in a total area of 29.8% in 2002 and 34.3% in 2017 in the study area. The extremely severe soil loss was found to be distributed mainly within the areas of moderate to high slope gradient while the estimated soil loss was relatively much lower on plain sites compared to the high slope lands.

Among the five RUSLE factors, the LS and K factor remains the same. Except for the R factor was found to be less during 2017 when compared to 2002. The land cover in the study area comprises of agricultural land, vegetation, bare ground, builtup land, and waterbody. Among the landcover classes, the agricultural lands, and bare ground experiencing high erosion along the extreme slope and vegetation areas were least eroded or had low to moderate erosion. The soil loss value was found to be more during 2002 because the R factor was higher but it is significantly different in prediction between 2002 and 2017 result concerning landscape, higher erosion risk level rate was estimated concerning landscape 34.3% in 2017 against the 29.8% in 2002 which can be attributed to change in C and P factor. These factors are directly linked to land cover change which is the ultimate result of urbanization. Determining soil loss rates helps in understanding the efforts needed to save the physical quality of land and ultimately holds valuable information for developing necessary conservation strategies.

4.0 CONCLUSION AND RECOMMENDATION

The general objective of the study is to examine the effects of urban growth on soil loss. In achieving the objective, this study applied Remote Sensing and GIS techniques with the RUSLE model. The Revised Universal Soil Loss Equation (RUSLE) model was categorized into four classes of erosion risk ranges from a low, medium, high, and very high soil loss.

Based on the result and findings of this research, there is significantly different in predicted result between 2002 and 2017 concerning landscape, higher erosion risk level rate was estimated concerning landscape and it is obvious that soil erosion is becoming a serious ecological problem in some places within the study area. To curb the menace of soil erosion and improve sustainable management in this area, it is recommended that

- i. Farming practices should be improved to reduce the soil erosion processes to the barest minimum through proper enlighten on the danger of poor farming techniques.
- ii. The traditional method of erosion control which is found to be cheap and effective (i.e. planting of plantain and banana on the floodplains) should be adopted
- iii. Adequate awareness of the effects of human activities that promote soil erosion on both floodplain and river channels should be encouraged in these areas.

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