

# Assessment of Climate Variability in the Niger Delta Region of Nigeria

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**Abstract.** This study investigated the variability of climate over the Niger Delta Region (NDR) focusing on the indicators of land use/land cover, land surface temperature (LST) and rainfall. Previous climate studies covered parts of the NDR; states and local areas, and these studies only quantified the amount of rainfall or air temperature separately. Hence, this study used a holistic approach. Rainfall and remotely sensed Landsat imageries (TM, ETM+ and OLI) for the period (1986-2018) were acquired and analyzed for land use/cover and obtained 97.6%, 98.8% and 99.8% accuracies landuse/cover for 1986, 2001 and 2017 respectively; 27 °C, 29 °C and 31°C was obtained as fluxes in surface temperature for 1986, 2001 and 2017 respectively. Results for rainfall variability from the interpolated maps showed a general trend for the study period. First, it was found that rainfall in terms of amount varied and increased within the study period (1985-2014) across the NDR. To further validate the variability of rainfall, the Annual Rainfall Anomaly Index (RAI) was further calculated and analyzed for frequency and intensity of the dry and rainy years in the study area and found that the area within the period of study recorded more positive anomaly (rainy years) than negative anomaly (dry years). It was established also in this study that there is a relationship between different land use/land cover categories, LST and rainfall. This was found in the correlation analysis results obtained amongst the three indicators, built up area had strong positive correlation with those of rainfall and temperature (0.78 and 0.75 %) respectively. See figure 20 and 21.

**Keywords:** Rainfall, temperature, land-use, land-cover, anomaly-index

## 1 Introduction

Variability of climate plays an important part in determining the conditions in which living organisms function in a geographical setting. The issues of variations in climate have been due to persistent increase in global temperatures and other weather conditions. Such as drastic changes in rainfall regimes, pressure belt etc., thus have been of major socio-economic and environmental concern to many nations especially those within the tropics where the impact is likely to be more (Ibitolu et al, 2014). Temperature, rainfall, wind, air pressure, cloud and relative humidity changes over a long period are used in determining the climate variability of a geographical setting, thus making their study significant. In this study changes in rainfall, land surface temperature land use land cover are the elements of concern to assess and determine climate variability.

Rainfall variability studies over time have been either areal or temporal. Areal variability is the variation of rainfall condition over space for a specific period of time, while the latter is the variation of rainfall at a given location over time. Rainfall is a critical component of water and energy cycles. It is a critical source of water for water supply, agriculture, natural ecosystems, hydroelectric power, and industry and is central to issues of drought, flood, and disease hazards. Rainfall is one of the substantial weather indicators of climate change (Stringer, 1995; Bello, 1992). On the concept of rainfall as a factor of climate change (Farmer and Wigley, 1985) stated thus; if the fluctuations in rainfall records persisted for a long time and were, furthermore, statistically significant, then we might say, there had been a climatic change based on such rainfall record. The changes in rainfall regimes have a direct relationship with the hotness and coldness of a place (temperature).

Land surface temperature (LST) is an important climatic factor in both environmental and climate studies because, Land surface temperature (LST) is an essential parameter in the physics of land surface

processes: it plays a key role in the energy and water transfers between the ground and the atmosphere. LST is controlled by solar radiation and the land-atmosphere heat exchange. Therefore, its spatial and temporal distributions reflect not only the variations of climate factors but also the land surface characteristics. An in-depth understanding of the spatial and temporal variations of LST is important to a range of research fields, including climate, vegetation, and hydrology Jin et al 1997, Manzo-Delgado et al, 2013 cited in Yongming et al, 2013.

LST represents the temperature recorded at the interface between the earth surface and the immediate atmosphere (Valiente, 2009) cited in (Ibitolu et al 2014), and is thus a measure of how hot the land surface at a particular location feels to touch. The LST is important in climatology as it has a direct impact on air temperature and it is also one of the main parameters in the underlying physics of land surface processes (Serban and Maftai, 2011; Dousset and Gourmelon, 2003) cited in (Ibitolu et al, 2014).

On the other hand, the land use practices and patterns equally affects the elements of weather that determine climate; thus any negative outcome serve as bottle-neck to the attainment of the SDG's (Sustainable Development Goals) 2030 agenda. Of note are (agenda 13, 14 and 15) which centers on climate change and its impacts, sustainable use of the ocean, seas and marine resources for sustainable development and sustainable use of terrestrial ecosystems, sustainably manage forests and biodiversity loss UN (2015).

The Niger Delta Region in recent time have been witnessing alternating changes in rainfall, land surface temperature rises and land cover changes; and these changes are partly attributed to climate change impacts resulting to extensive damages and loss of lives (Ituen, and Alonge, 2009; Ologunorisa and Tersoo, 2006; Udosen, 2012; Isikwue et al, 2013; Mathew et al, 2013; Okoro et al, 2014; Igweze et al, 2014; Ibitolu et al, 2014; and Egor et al, 2015).

Climatic Variability denotes the inherent dynamic nature of the climate on various temporal scales. Such temporal scale variations could be monthly, seasonal, annual, decadal, periodic, quasi-periodic or non-periodic (Akpodigaga and Odjugo, 2010). Variation in climate is a product of either natural (biogeographical) and human activities (anthropogenic) processes. The human factors have been proven to be responsible for the ongoing unequivocal climate change and global warming (IPCC, 2007).

Variations on climate in the Niger Delta Region has been attributed more to anthropogenic factors (gas flaring and socio-economic activities of urbanization) leading to loss of biodiversity which have left the vegetal cover bare resulting in excess of carbon in the atmosphere. This biodiversity loss has led to overheating of the atmosphere and rise in temperature; thus resulting in excessive evaporation of the sea and nearby water bodies; this results to high spatio-temporal variability of rainfall which births environmental problems (Ituen, and Alonge, 2009; Ologunorisa and Tersoo, 2006; Udosen, 2012; Isikwue et al, 2013; Mathew et al, 2013; Okoro et al, 2014; Igweze et al, 2014 and Egor et al, 2015).

The quest and strive toward industrialization, technological innovations and the drive to modernism has resulted to modification and change in land use and land cover in a geographical setting thus serving as variable conditions that intensify hazards over an area leading to changes in the natural environments of such settings (Effiong, 2011; Mmom and Nwagwu, 2013; Aderoju, 2014; Odjugo et al, 2015; Ejemeyovwi, 2015; Anieka and Ubom, 2016, Bariweni and Amukali, 2017); thus the foregoing literature indicates that previous climate studies covered parts of the Niger Delta Region; states and local area, and these studies only quantify the amount of rainfall (Akinsanola&Ogunjobi, 2014; Igweze et al , 2014), some other authors studied climate change with regard to air and or land surface temperature separately though not for the whole Niger Delta Region (Ibitolu et al, 2014).

It appears that regional assessment of climate change considering rainfall, land surface temperature and land use/cover is rarely studied. Hence, this study used a holistic approach, a combination of multiple indicators of climate change (land use /land cover changes, land surface temperature and rainfall) using remote sensing and GIS techniques to test for variations in climate over the Niger Delta Region.

## 2 Materials and Methods

### 2.1 Types and Sources of Data

This study used primary and secondary data. Thus the primary data included GPS ground truth point data to reference points of the rainfall data and also ground reference points for the land use land cover changes (supervised classification) table 1. Landsat imageries for 3 epochs (1986), (2001) and (2017) from the archives of the United States Geological Survey (USGS) were used as secondary data. From this data sets; land surface temperature and land use/land cover information were extracted after processing. Also part of the secondary data used was rainfall data (1985-2014) from 11 existing rainfall stations (Akure, Benin, Warri, Asaba, Yenagoa, Port Harcourt, Owerri, Uyo, Calabar, Ogoja and Ikom) across the NDR. The rainfall data was retrieved from the archives of the Nigerian Meteorological Agency (NIMET) and was used for rainfall variability assessment.

**Table 1.** Data types and sources

Data Type	Source	Year	Resolution	Purpose
Rainfall	Nigerian Meteorological Agency	1985-2014	-	Rainfall Variability
Landsat 4-5 (Path/Row)	www.earthexplorer.usgs.gov	1986	30×30m	LULC & LST
Landsat 7 ETM+ (Path/Row)	www.earthexplorer.usgs.gov	2001	30×30m	LULC & LST
Landsat 8 OLI (Path/Row)	www.earthexplorer.usgs.gov	2017	30×30m	LULC & LST
GPS Coordinate	Field Work	2018	-	Ground Truth Data for Change Detection

Source: Compiled by author, 2018

### 2.2 Methods

#### 2.2.1 Land use/land cover assessment

The Landsat imageries (Thematic Mapper TM, Enhanced Thematic Mapper ETM+ and Operational Land Imager OLI) were processed for change detection (from the data type used as outlined in table 3.1) within the study period under consideration for 1986, 2001 and 2017 over the Niger Delta Region NDR. The Anderson (1971) Classification Scheme Level I was adopted for this purpose. This is because the study area covers a wide area (inter-state/regional coverage), thus making this scheme suitable for this study see Table 2. Supervised algorithm (Maximum Likelihood) was used; This algorithm assumes that the estimated probabilities are equal for all classes and the histograms of the input bands have normal distributions in order to get a precise outcome. This method considers mean, variances and the variability in brightness values of each class given as a training set.

**Table 2.** Classification scheme

Land Use/Land Cover	Description
Built-Up	All residential, commercial and industrial areas (villages, settlements, transportation infrastructures)
Waterbody	River, permanent open water, lakes, ponds, canals, reservoirs.
Vegetation	Trees, forests, old vegetated lands
Bare-land	Grassland, cultivated areas, earth and sand lands, infillings and solid waste landfills.

Source: Anderson et al, (1971)

#### 2.2.2 Land surface temperature assessment

The thermal bands (band 6 in Landsat 4-5 and band 10-11 in Landsat OLI) were extracted to get the land surface temperature statistics. This was done by first converting their Digital Numbers (DNs) to Top of the Atmosphere (TOA) radiance values which were in turn converted to brightness temperature in Kelvin (K), then to degrees Celsius (°C) as final output to determine the Land Surface

Temperature(LST). For the Landsat OLI, the Land Surface Emissivity (LSE) and Normalized Difference Vegetation Index (NDVI) was further calculated and converted from degrees kelvin (0K) to degrees Celsius ( $^{\circ}\text{C}$ ) using GIS.

### 2.2.3 Rainfall data analysis

To analyze the rainfall data, the study employed basic statistical techniques such as computation of totals and mean (annual mean 1985-2014) to give meaning to the rainfall data, to have a fair look of the general pattern and characteristic of rainfall in the region.

Mean annual precipitation was then calculated for the periods 1985-1987, 2000-2002, 2012-2014 and 1985-2014 in Microsoft Excel for interpolation. The rainfall values were imported as dbf files into a GIS interface for interpolation analysis using Inverse Distance Weighting (IDW) algorithm. The IDW has the advantage that it is easy to define and therefore easy to understand the results because the interpolating point is most influenced by nearby points and less by the more distant points: the principle of spatial auto-correlation or Tobler's First Law of Geography, Burrough and McDonnell, (1998).

This study equally tested for frequency and intensity of dry and rainy years that is; increases and reductions in rainfall. The Van Rooy's Rainfall Anomaly Index was employed to check for trend (Van Rooy, 1965) table 3. RAI, developed and firstly used by Rooy (1965) and adapted by Freitas (2005), constitutes the following equations:

$$\text{RAI} = 3\left(\frac{N-\hat{N}}{\hat{M}-\hat{N}}\right) \text{ For positive anomalies} \quad (1)$$

$$\text{RAI} = -3\left(\frac{N-\hat{N}}{\hat{X}-\hat{N}}\right) \text{ For negative anomalies} \quad (2)$$

where:

$N$  = current monthly/yearly rainfall, in order words, of the month/year when RAI will be generated in (mm);

$\hat{N}$  = monthly/yearly average rainfall of the historical series (mm)

$\hat{M}$  = average of the ten highest monthly/yearly precipitations of the historical series (mm);

$\hat{X}$  = average of the ten lowest monthly/ yearly precipitations of the historical

**Table 3.** Classification of rainfall anomaly index intensity

	RAI Range	Classification
Rainfall Anomaly Index (RAI)	Above 4	Extremely Humid
	2 to 4	Very Humid
	0 to 2	Humid
	-2 to 0	Dry
	-4 to -2	Very Dry
	Below -4	Extremely Dry

Source: Freitas (2005) adapted by Araújo et al. (2009)

### 2.2.4 The link between the different land use/land cover categories, land surface temperature and rainfall

To establish relationship between two land use/land cover categories; (vegetation and built-up land, land surface temperature and rainfall) were used to test for correlation. The correlation and regression analysis in Microsoft Excel (data analysis tool) was used to quantify the measure of strength of association between variables (e.g., between an independent and a dependent variable or between two independent variables). The sample correlation coefficient, denoted  $r$ , ranges between -1 and +1 and quantifies the direction and strength of the linear association between the two variables. Thus, the following regression equation applies;

$$r = \frac{\Sigma(x-x')(y-y')}{\sqrt{\Sigma(x-x')^2 \Sigma(y-y')^2}} \quad (3)$$

where

$n$  = Quantity of information

$\Sigma x$  = Total of the first variable value

$\Sigma y$  = Total of the second variable value

$\Sigma xy$  = Sum of the product of first & second value

$\Sigma x^2$  = Sum of the squares of the first value

$\Sigma y^2$  = Sum of the squares of the second value

On the other hand, regression is used to explain the relationship between one continuous dependent variable and an independent variable (s). The independent variables can be continuous or categorical.

Thus the simple linear regression equation as shown below;

$$Y_i = \alpha + \beta \times X_i + \epsilon_i \quad (4)$$

where

Y is the dependent variable

X is the independent (explanatory variable)

$\epsilon$  is the random error term.

For this analysis, the dependent variables (Y) are vegetation and built-up, and the independent (X) variables include Land Surface Temperature and Rainfall. So each of the dependent variables (vegetation and built-up land) will be paired separately with those of the independent variables (LST and rainfall).

### 3 Result and Discussion

#### 3.1 Land Use and Land Cover of the Study Area between 1986 and 2017

To achieve the objective of the land use land cover change over the study area, an object-oriented image analysis was carried out to generate the LULC maps. To justify quality, the generated images were further validated and quantified for classification accuracy assessment. The overall accuracies for the maps of 1986, 2001 and 2017 were 90.80%, 96.06% and 95.40% respectively. This is considered satisfactory for being up to the minimum accuracy threshold requirement of 85% (Eastman, 2000). The categories applicable to this study under the Anderson's level I classification scheme include; built-up land, Forest land, water-body, wet-land and bare-land. Classified land use is presented in figures 1, 2 and 3. The results detailed forest land as the dominant land cover occupying between 88% and 43%, with water body having less than 3% as the least category (Table 4).

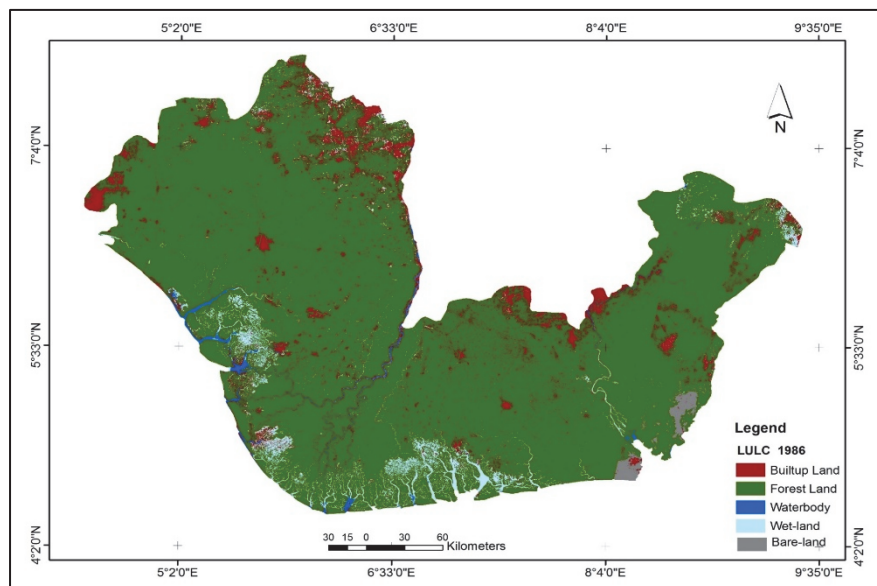
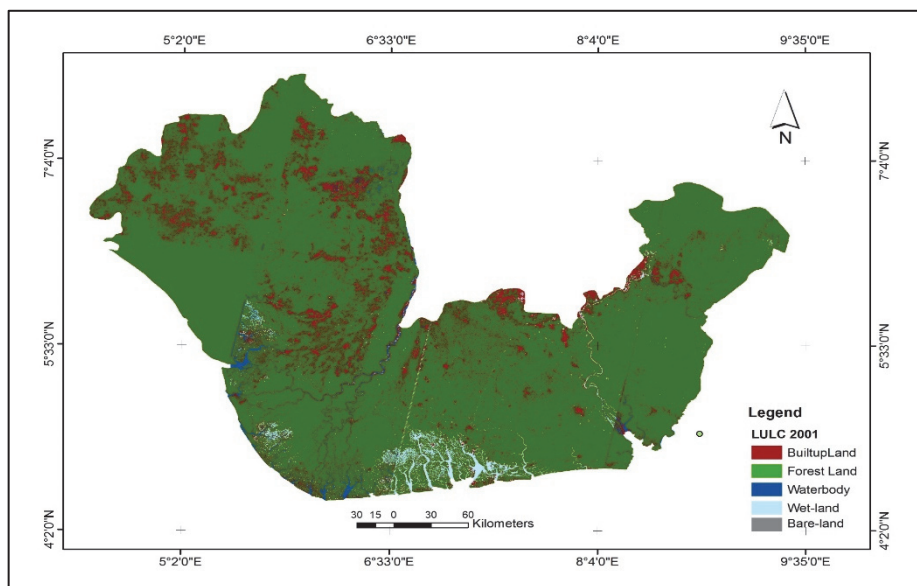
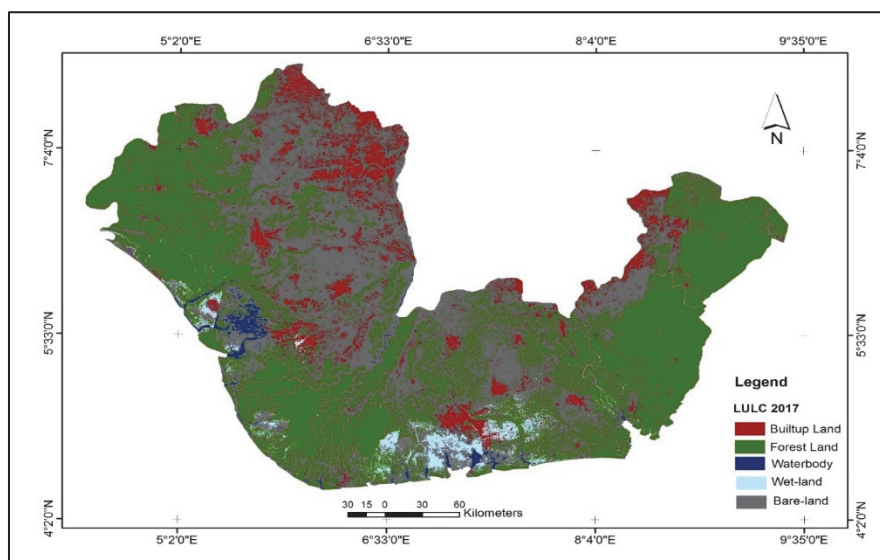


Figure 1. Land use land cover 1986 (Source: Author, 2018)



**Figure 2.** Land use land cover 2001 (Source: Author, 2018)



**Figure 3.** Land use land cover 2017 (Source: Author, 2018)

**Table 4.** Land class statistics 1986-2017

Category	1986(Sq. k)	Percentage	2001 (Sq. k)	Percentage	2017(Sq. k)	Percentage
Builtup Land	7019.36	6%	8096.29	7%	9083.81	8%
Forest Land	95690.20	88%	90863.74	87%	53144.74	49%
Waterbody	1015.07	1%	1688.12	2%	3321.25	3%
Wetland	3856.82	4%	1688.12	2%	3546.01	3%
Bareland	914.53	1%	66.95	0%	39639.77	36%
<b>Total</b>	<b>108496</b>	<b>100%</b>	<b>108736</b>	<b>100%</b>	<b>108736</b>	<b>100%</b>

Source: Author, 2018

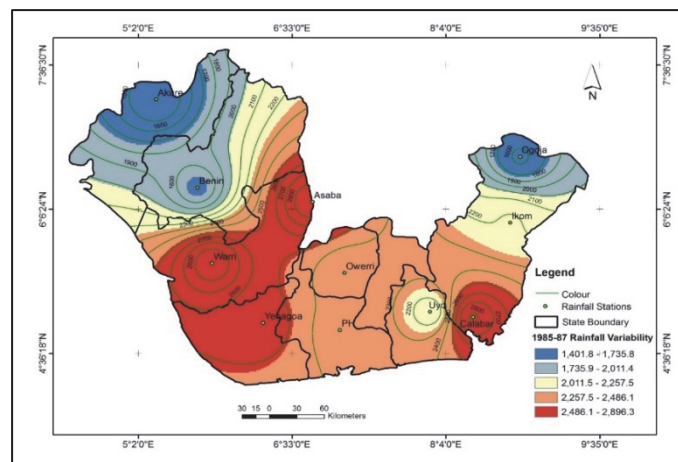
### 3.2 Land Surface Temperature Assessment

The land surface temperature was assessed, the results showed persistent increases from 1986 through 2001 to 2017. The result showed temperature fluxes between 18 °C to 27 °C for 1986 as minimum and maximum average temperatures, 2001 recorded 18 °C to 29 °C minimum and maximum average temperatures. For 2017, higher temperature values of 19 °C to 31°C as minimum and maximum average temperature values were found table 5.

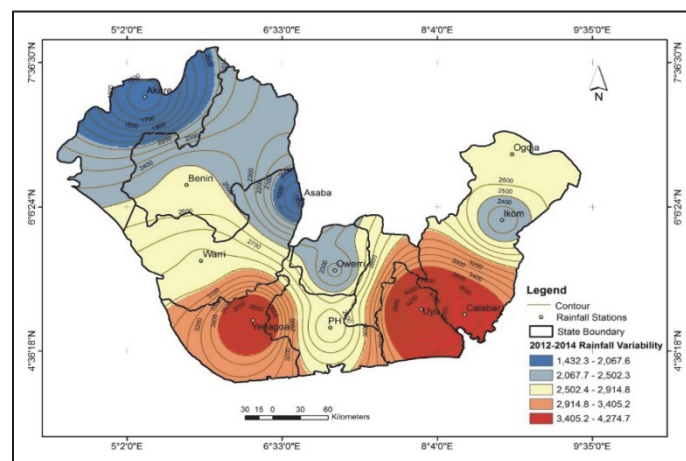
**Table 5.** Land surface temperature averages

Period	Minimum Average	Maximum Average
1986	18 °C	27 °C
2001	18 °C	29 °C
2017	19 °C	31 °C

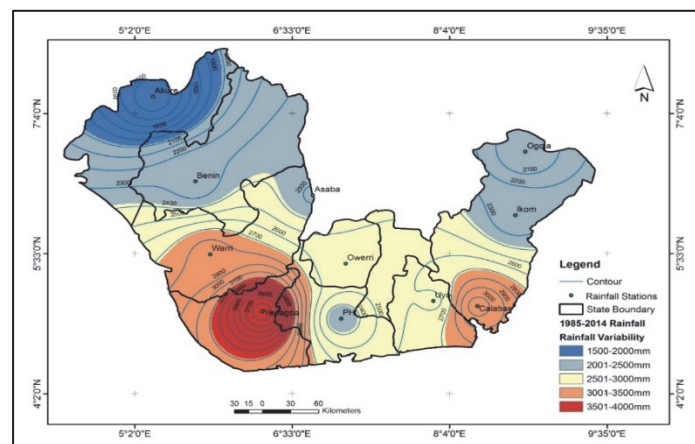
Source: Author 2018



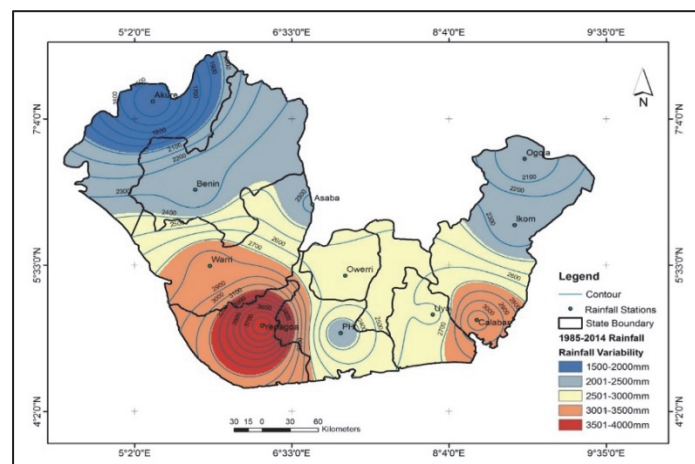
**Figure 4.** Rainfall between 1985-1987 (Source: Author, 2018)



**Figure 5.** Rainfall between 2012-2014 (Source: Author, 2018)



**Figure 6.** Rainfall between 1985-2014 (Source: Author, 2018)



**Figure 7.** Rainfall between 1985-2014 (Source: Author, 2018)

### 3.3 Rainfall Variability over the Study Area

Rainfall for this study was considered in terms of annual mean amount and variability; both total amount and variability has been considered and assessed on annual and periodic basis (3 epochs; 1985-1987, 2000-2002 and 2012-2014). The average rainfall amount for each year within the study period (1985-2014) was also computed for in the NDR and mapped (Fig 4-8). From the interpolated maps, the pattern of rainfall showed a general trend for the study period. First, it was found that rainfall in terms of amount increased within the study period (1985-2014). Second, the latter years in the southern part of the region recorded consistent higher rainfall amount (above 3000mm/annum) throughout the period of study. From here (the south parts) rainfall decreases to within 2000mm and below in the south east and South Western parts as areas of moderately consistent rainfall. The north east and extreme north west portion were generally found to be areas with consistent low rainfall values (Fig. 4-7).

### 3.4 Frequency Intensity of Dry and Rainy Years (1985 to 2014)

To further validate the variability of the rainfall, the Annual Rainfall Anomaly Index (RAI) was calculated to analyze the frequency and intensity of the dry and rainy years in the study area. In addition, the monthly RAI was computed and calculated for the historical series (1985-2014) aiming to analyze the distribution of rainfall in the years of greatest anomaly and the obtained results were interpreted against the Freitas Classification scheme as shown in table 3.



From the temporal and spatial distribution of rainfall in the NDR, the rainy season begins in the month of March and lasts until the month of September-October. Within this range, the precipitation is higher than the annual average. The month of September presented the highest precipitation value, over 3000 mm in most of the stations assessed within the study period, this month is more efficient in terms of precipitation in the Southern parts of the region, this is because it has a greater influence of the Intertropical Convergence Zone (ITCZ), which is important for the recharge of the river system. The drought years and the rainy years, during the period from 1985 to 2014, can be visualized by means of RAI (Figure 8-17), enabling to identify periods where these events were more intense and/or lasting.

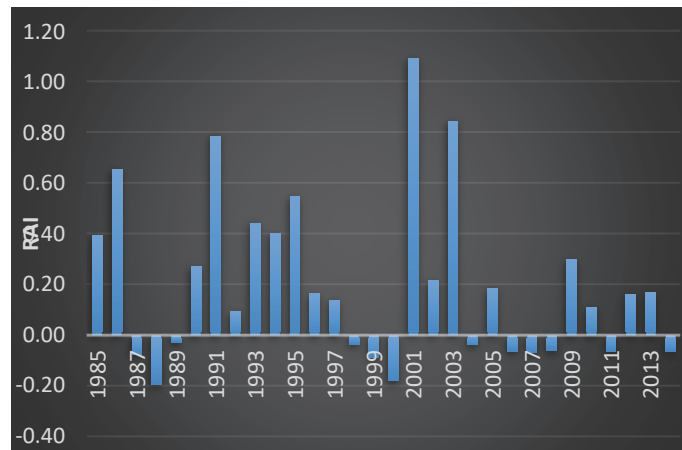


Figure 8. Akure RAI index

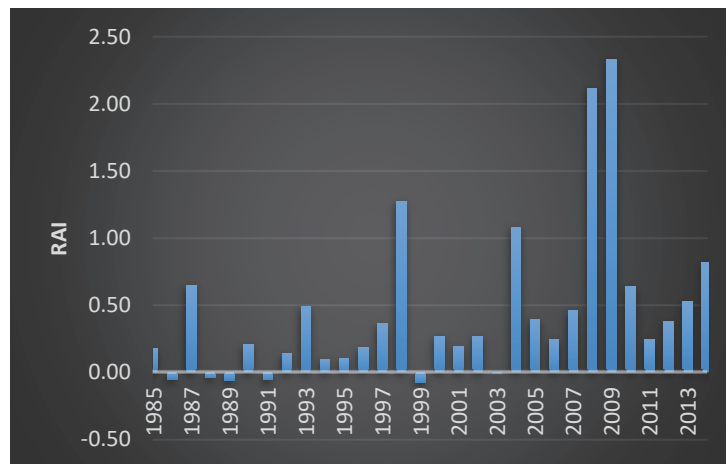


Figure 9. Owerri RAI index

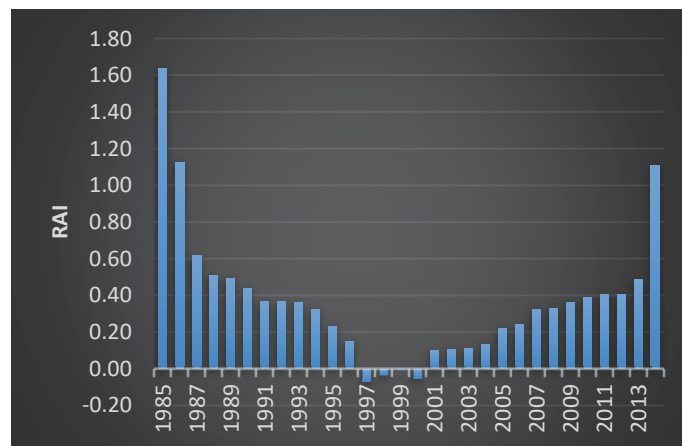


Figure 10. Benin RAI index

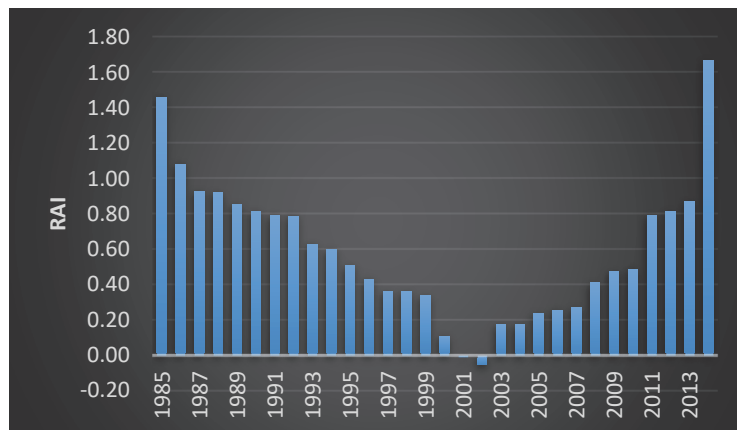


Figure 11. Ogoja RAI index

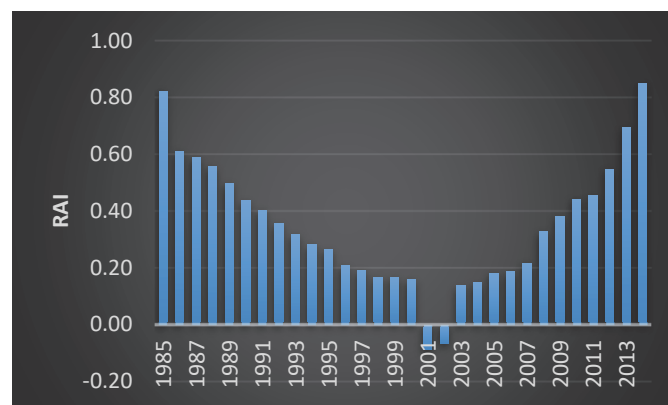


Figure 12. Calabar RAI index

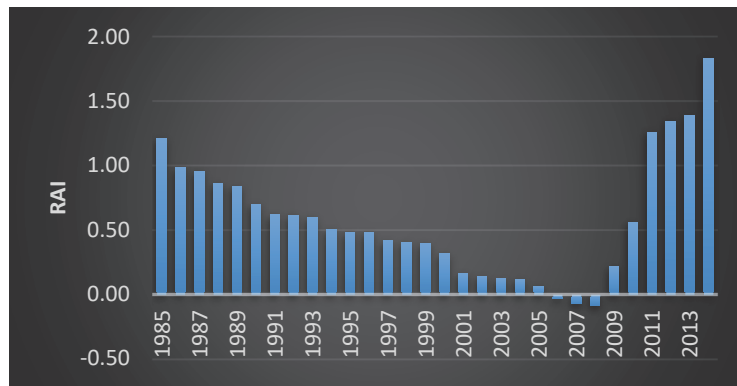


Figure 13. Uyo RAI index

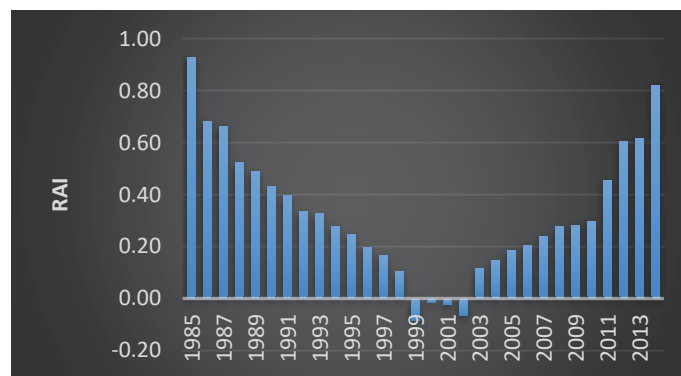


Figure 14. Port Harcourt RAI index

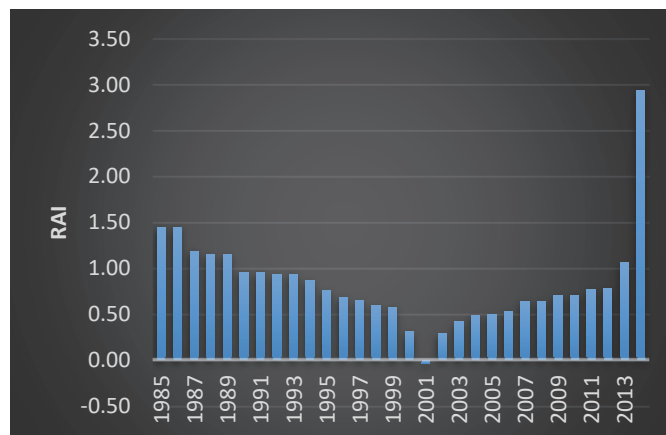


Figure 15. Asaba RAI index

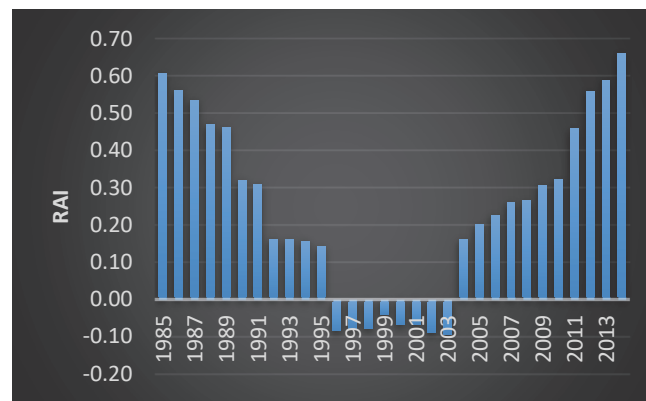


Figure 16. Warri RAI index

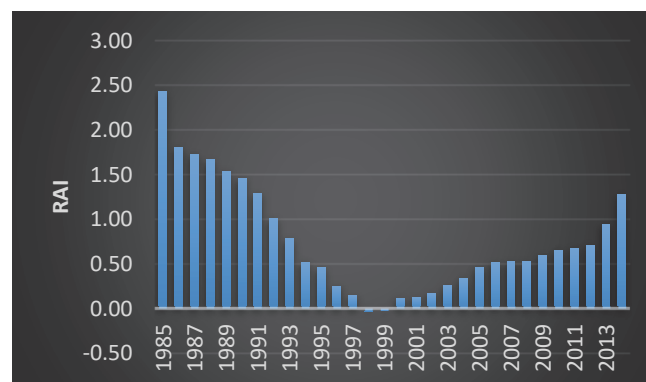


Figure 17. Yenagoa RAI index

The positive values observed in the figures above (figure 9-17) represent rainy or wet years and the negative values represent the dry years, with different degrees of intensity. Within the 30 years' period each across the 11 rainfall stations in the NDR (making a total of 330 years'), the study found occurrence of positive RAI values in 276 years, varying from very humid to humid, and 54 years with a negative RAI, varying between very dry and dry. In other words, there were more years of rainy than dry years within the period of study ((1985-2014). The periods that remained the longest with droughts were (Akure 1987 to 1989, 1998 to 2000, 2005 to 2008, 2011 and 2014); (Ikome 1995 to 2004); (Warri 1986 to 2003); (Owerri 1986, 1988 to 1989, 1991, 1999, 2003); these locations in the study area had years with the highest negative value, with a RAI of 0.0 values, classified as dry (table 3). The years of greatest positive value were (Asaba, 2014; Yenagoa 1985; Benin 1985; Owerri 2009 and Akure 2001), with an average RAI of 2.0 to 2.8, being classified as very humid (table 3). Thus, the revelations above suggest one thing and that one thing is that climate change indicators have varied over time and these fluxes calls for concern.

### 3.5 Relationship between Land Use/Land Cover, Land Surface Temperature and Rainfall

To establish relationship between the two land use/land cover categories (Vegetation, Built-up), land surface temperature and rainfall. The correlation and regression analysis was used. Vegetation and built-up (dependent variables) was correlated and regressed against LST and rainfall (independent variables). The results reveal two kinds of relationship, first the category with strong correlation, Figure 20 (temperature and built-up land) and Figure 21 (rainfall and built-up land). The second category is the category with weak correlation Figure 19, (between temperature and vegetation). The study also found a strong negative correlation between vegetation and rainfall; as rainfall increases, vegetated land decreases over time (Figure 19).

The study found a strong degree of association between the indicators of both rainfall and temperature, built-up land and land surface temperature within the period of study. That is as built up increases the climate indicators of rainfall and temperature increases likewise, amounting to a strong positive correlation (0.78 and 0.75 respectively) as shown in (Figure 20 and 21). The second category of association is the category with weak correlation, under which we have (vegetation and temperature). In this category, there is relationship but a weak relationship Figure 19 and 18.

This study in a nutshell found variation in land use/land cover, rainfall, land surface temperature as having persistent increase over the period of study and these variations have micro climatic implications. The Intergovernmental Panel on Climate Change predicts that the level of threats to forests and vegetation will increase in the 21st century; this is the case in the Niger Delta Region, the region has witnessed persistent decrease in vegetation corroborating with the findings of Okali and Eleri (2004). Rising temperatures, heavy rains, and humidity will render forests and vegetation more prone to many threats, including pests and diseases. Forests are vitally important as they soak up carbon dioxide, the main greenhouse gas responsible for global warming, and help regulate the world’s climate. They are also home to countless plant and animal species.

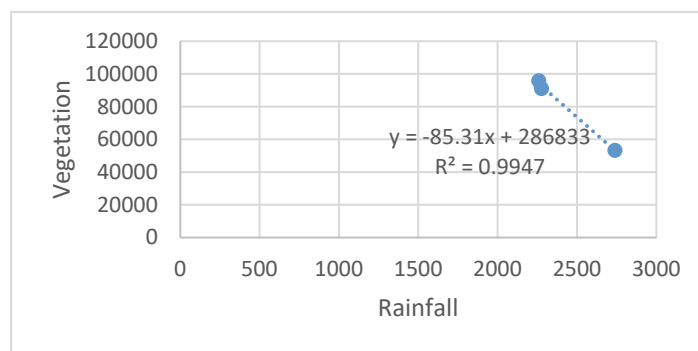


Figure 18. Rainfall & vegetation strength/trendline

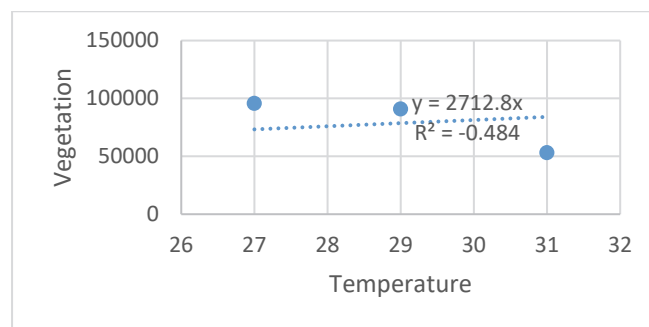
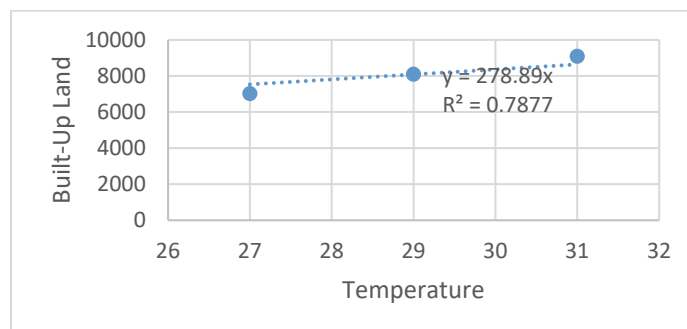
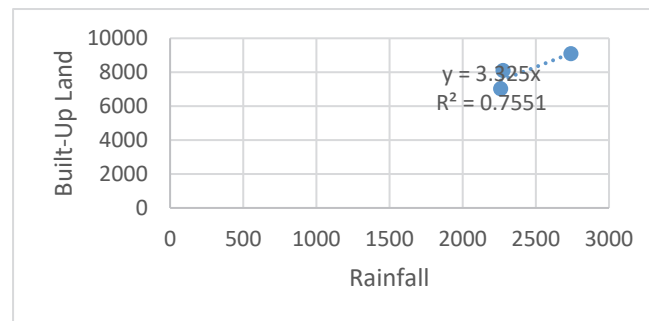


Figure 19. Temperature & vegetation strength/trendline



**Figure 20.** Temperature & strength/trendline**Figure 21.** Rainfall & builtup strength/trendline

In tropical forests such as the Niger Delta Region where there is abundant biodiversity, even modest levels of climate change can cause high levels of extinction. When large areas of forest are destroyed it is disastrous for the local species and communities that rely on them. There are variable conditions that intensify environmental hazards and degradation as a result of human occupation that are also evident in the Niger Delta Region that have led to vegetal cover changes in the region due to developmental activities which have led to rapid growth that also reduce surface intensity; for example, residential land use, roads, (Twumasi and Merem, 2006; Abbas, 2012; Onwuteaka, 2014 and Ejemeyovwi, 2015). State creation and the development of the oil industry have resulted in the emergence of new towns and subsequently increase in the spatial extent of built-up areas and development of infrastructure; all these have direct effect on CO<sub>2</sub> emissions.

Dying trees emit their stores of carbon dioxide and methane, adding to atmospheric greenhouse gases and setting us on a course for runaway global warming leading to increases in coastal rainfall & eventual rise in sea level. This persistent decrease in vegetation cover is also directly related to increases in the built-up land for this region as shown from our correlation analysis. changes in the land use/land cover affects global systems: atmosphere, climate, forests, sea levels and also a significant effect on localized places were such changes occur because of their global warming, loss of biodiversity and impact on human life. Thus, built-up land was found to be of increase over the past decades. This was also reported by the previous studies of Twumasi and Merem (2006), Onwuteaka (2014) and Ejemeyovwi (2015); these studies found increase in built-up land in the Niger Delta Region. The study also found extreme rainfall conditions (persistent increases) over the Niger Delta Region that makes it vulnerable to both flooding and draught. This was further confirmed from the results obtained from the Rainfall Anomaly Index (RAI) analysis conducted in this study, which reveal that the region has recorded more years of positive rainfall anomaly than negative. Previous studies of Mathew et al (2013), Okoro et al (2014), Akinsanola and Ogunjobi (2014) and Egor et al (2015) also found steady increase in rainfall over the past decades in the Niger Delta Region.

## 4 Conclusion

Previous climate studies covered parts of the Niger Delta Region; states and or local areas, thus studies only quantify the amount of rainfall and some other authors studied climate variability with regard to air and or land surface temperature separately though not for the whole Niger Delta Region. Thus warranted this holistic approach a combination of multiple indicators of climate change (land use /land cover changes, land surface temperature and rainfall variability) using remote sensing and GIS techniques to test for variations in these climate indicators over the last 30 years in the Niger Delta Region; (the knowledge gap in this study). The results showed a joint link between rainfall variability, land use changes and land surface temperature changes over the Niger Delta Region within the period of study. This was found in the correlation analysis results obtained amongst the three indicators, built up area had strong positive correlation with those of rainfall and temperature (0.78 and 0.75 %) respectively. See figure 20 and 21.

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