

# Impact of seasonal urban greening variability on land surface temperature in Benin (West Africa).

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## Research Article

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# Abstract

Urban greening quantification and seasonal variation impact for surface temperature mitigation have been poorly understood and monitored in Benin. We have attempted to assess of season variability of normalized difference vegetation index (NDVI) using Landsat 7 and 8 data to get more accurate information about the vegetation tendency and effects on urban surface temperature mitigation. Monthly satellite derived NDVI series were collected using Google Earth Engine open-source while observed surface temperature were obtained at National Agency of Meteorology in Benin. We performed Mann-Kendall trend test to assess the significance of the trend charts. ANOVA followed by SNK test was applied to examine seasonal variation on NDVI significance. Likewise, a simple linear regression was applied to show the relationship between surface temperature and NDVI variables. No significant linear trends ( $p$ -values  $> 0.05$ ) were observed for annual vegetation index (NDVI) and surface temperature. The NDVI tendency decreased from 2000 to around 2011 and started to increase up to 2020. However significant difference ( $P$ -Value  $< 0.001$ ) between seasons was found for the mean NDVI and January-March period showed the lowest values of mean NDVI. Despite the negative relationship between annual mean land surface temperature (LST) and NDVI, the dry seasons showed a positive correlation in each city. Thereby, judicious urban management of water availability such as greenery irrigation would be more helpful for ensuring the sustainable urban surface temperature mitigation.

# Introduction

Cities are complex ecosystems that are mostly affected by ecological consequences of uncontrolled urban growth that can be briefly summarized as increasing urban heat, degradation of air, water and soil cycles and quality, trouble in land-use and cover, loss of biodiversity, change in climate as well as transport congestions and social vices (Siddiqui et al., 2018). Urban green infrastructure (urban forests, wetlands, parks, street trees, small gardens, green roofs, and walls) that is the hybrid of greenery and built systems has therefore evident impacts as strategy to improve for adapting and mitigating climate change issues as well as providing good quality of life in the world (Norton et al., 2015; Scott et al., 2016). Urban vegetation and trees have effective reduction impacts on air temperature through shading radiation and evapotranspiration functions (Wang et al., 2016). For instance, pedestrian greening within the Catania University campus reduced by  $1.00^{\circ}\text{C}$  the air temperature (Detommaso et al., 2021). Anamika & Pradeep (2016) in reviewing urban vegetation roles in air pollution found that urban trees were actor for  $\text{O}_3$ ,  $\text{PM}_{10}$ ,  $\text{NO}_2$ ,  $\text{SO}_2$ ,  $\text{CO}$  molecules pollution removal in range of 711,000 metric tons per year over 55 US cities. However, these urban greenery functions are made more efficient when vegetation phenology and structure are also agreed.

Changes in urban vegetation phenology may have significant implications for carbon cycles with feedbacks on urban ecosystem climate. Indeed, dense foliage cover is followed by simultaneous increases transpiration rates (increased canopy conductance), decreased albedo through reducing shortwave radiation absorbed and increased ground aerodynamic roughness that can favour air turbulence and the sensible heat transfer to and from the atmosphere (Piao et al., 2019). Moreover, there

is high correlation between vegetation development and season variation in tropics (Ayanlade, 2016; Ayanlade & Howard, 2019). Then, the cooling effects induced by urban vegetation should be also changed by seasons in years. The study carried out by Jeong et al. (2009) demonstrated a high decreasing correlation between the growth vegetation and the increased rate of surface temperature in spring. Likewise, a study found some asymmetric responses of vegetation development to autumn and spring warm effects (Li et al., 2018). However, these seasonal variations in the relationship vegetation-air temperature can vary significantly with region (Jeong et al., 2009).

Many cities in Benin Republic are experiencing an uncontrolled urbanization with consequences on urban heat expression that have inevitably impacted on the health and the daily lives of local residents (Lanmandjèkpogni et al., 2019; Falolou et al., 2020). Moreover, it was recognized that the development and good management of urban vegetations contribute enormously to the mitigation and adaption of this heat effect in the country (Osseni et al., 2015; Teka et al., 2017). However, the quantification and temporal variation of the contribution of these vegetations in urban warming reduction is needed for sustainable urbanisation planning. This paper examines urban vegetation trend with surface temperature, characterize the seasonal variation of urban greening and assesses the influence of the greenness in urban areas on land surface temperature reduction in Benin (West Africa).

## Methods

### Study Area

The cities of Porto-Novo ( $6^{\circ}26'$ -  $6^{\circ}31'$  and  $2^{\circ}34'$ -  $2^{\circ}41'$ ) and Parakou ( $9^{\circ}15'$ -  $9^{\circ}27'$  and  $2^{\circ}31'$ -  $2^{\circ}45'$ ) were chosen as the case study. They are respectively the second and third largest and fastest growing cities in Benin (Lohnert, 2017). The population of Porto-Novo was 264,320 inhabitants and the population of Parakou was 255,478 inhabitants (INSAE, 2016). Porto-Novo is located in the southern region while Parakou in the central region of the country (Fig. 1).

Porto-Novo has a wet tropical climate (Subequatorial climate) with annual rainfall estimated at 1200 mm/year (Houngla et al., 2019). Two rainy seasons (usually from April to July and September to November) and two dry seasons (month of August and December to March) were observed in this area (Ahokposs, 2018). Parakou has dry tropical climate (Sudano-Guinean climate) and the average annual rainfall was between 1170 and 1200 mm/year (Neuenschwander et al., 2011; Miassi & Dossa, 2018; Lanmandjèkpogni et al., 2019) and characterized by a unimodal rainy season from June to October with a peak in August (Lanmandjèkpogni et al., 2018).

The natural vegetation is almost non-existent in urban area of Porto-Novo (Osseni et al., 2015). While tree plantations and gallery forests are mostly observed in urban areas of Parakou (Miassi & Dossa, 2018).

### Data

The urban green-stuff importance and productivity is widely characterized by the normalized difference vegetation index (NDVI) (Li et al., 2017; Malik et al., 2020). The data used in this paper included the NDVI image collections and observed land surface temperature. Monthly satellite derived NDVI series were collected on the two cities using Google Earth Engine open-source (<https://code.earthengine.google.com>). The observed monthly temperature data were obtained at the National Agency of Meteorology of Benin from 1990 to 2020. Due to the limitation in meteorological data for certain periods, we completed the missing data with remote sensing data from Landsat images with high-resolution radiometer sensors of Landsat 7 (2000 to 2013) and Landsat 8 (2014 to 2020). Land surface temperature was obtained using Google Earth Engine codes recently created by Ermida et al.(2020)

([https://code.earthengine.google.com/?accept\\_repo=users/sofiaermida/landsat\\_smw\\_lst](https://code.earthengine.google.com/?accept_repo=users/sofiaermida/landsat_smw_lst)).

## Analysis

We performed the linear-trend analyses on annual mean of NDVI and land surface temperature using a non-parametric over the time period of 2000-2020 (Li et al., 2018). Mann-Kendall test was applied to assess the significance of the trend chats (Mondal et al., 2012). The analysis of variance (ANOVA) was used after checking normality and homogeneity of data to examine the significance of year-seasons (Trimester periods) variation on vegetation index (NDVI). Student-Neuman-Keuls (SNK) test was applied to make difference in means once ANOVA test was significant. A simple linear regression was applied to show the relationship between surface temperature and NDVI. R core software version 4.0.3 (R Core Team, 2019) was employed in all statistical analyses.

## Results

# Changes in urban vegetation index (NDVI) and land surface temperature (LST) from 2000 to 2020

The test of Mann-Kendall showed no significant linear trend in annual NDVI data ( $p$ -values  $> 0.05$ ). Likewise, no significant linear trend was found in annual LST for data from the combined cities and the city of Parakou ( $p$ -values  $> 0.05$ ). However, a significant linear trend was moderately found in annual LST data for the city of Porto-Novo ( $p$ -values = 0.023). Therefore, an upward trend can be attribute to the annual LST data of Porto-Novo (Fig. 3).

We also noted a decreasing tendency of NDVI from 2000 to 2011 and an increasing up to 2020 (Fig. 2).

### Seasonal variation of vegetation index (NDVI)

The vegetation index (NDVI) assessed according to the annual quarters (sensibly season variation) from 2000 to 2020 showed significant difference ( $P$ -Value  $< 0.001$ ) for the combined cities as well as for each city (Table 1). The mean NDVI for January-March period was significantly low for the combined cities and Porto-Novo. In opposite, the period of April-June had the high mean NDVI value (Table 2). Regarding of

July-September and October-December periods, no statistically different has was found (Table 2). However, the situation of Parakou was a bit different from Porto-Novo. Although the period of January-March still kept the low mean NDVI value, no significant difference was found for the remaining periods of the year (Table 2).

Table 1  
Summary of ANOVA model on vegetation index (NDVI)

City	df	Sum sq	Mean sq	F-Value	P-Value
Both cities	3	0.701	0.233	19.46	6.71e-12 <sup>***</sup>
Porto-Novo	3	0.503	0.167	16.45	8.3e-10 <sup>***</sup>
Parakou	3	0.234	0.078	5.76	0.000 <sup>***</sup>
Signif. codes: '***'= 0.001					

Table 2  
Mean, standard deviation (Std) and difference structuration of vegetation index

Period	Both cities		Porto-Novo		Parakou	
	Mean	Std	Mean	Std	Mean	Std
Jan-March	0.147 <sup>a</sup>	0.051	0.137 <sup>a</sup>	0.054	0.162 <sup>a</sup>	0.043
April-June	0.255 <sup>b</sup>	0.133	0.260 <sup>b</sup>	0.119	0.245 <sup>b</sup>	0.146
July-Sept	0.202 <sup>c</sup>	0.151	0.200 <sup>c</sup>	0.139	0.213 <sup>b</sup>	0.159
Oct-Dec	0.213 <sup>c</sup>	0.077	0.195 <sup>c</sup>	0.063	0.233 <sup>b</sup>	0.085
Means with the same letter are not significantly different (SNK test).						

### Urban surface temperature and seasonal NDVI influence

Figure 4 showed the linear regression analysis between annual LST and NDVI for the combined cities. On this figure, each point showed the mean LST and NDVI value linked to urban area, the red smoothly line was the result of linear simulation and the function of linear regression was also exhibited at the top. It could be therefore understood that the regression analysis result showed a significant inverse correlation between mean LST and NDVI values (p-value = 0.01165, Fig. 4). Therefore, we can suggest that a seasonal time with more vegetation cover (higher NDVI) induced correspondingly greater rates of evapotranspiration and lead to the land surface and atmosphere latent and sensible heat exchange in comparison to the seasonal period with few vegetation covers at the same areas. Similar result was obtained for each individual city (Porto-Novo: p-value = 0.02 and Parakou: p-value = 0.000; Table 3). However, when considering the individual season periods, the regression analysis showed the positive

correlation between mean LST and NDVI values for January–March for Porto-Novo (Table 3 and Fig. 5) and April–June for Parakou (Table 3 and Fig. 6). This finding means that urban vegetation may not induce a moderate land surface temperature reduction at these periods respectively in Porto-Novo and Parakou.

Table 3  
Linear regression and correlation coefficients for the relationship between LST and NDVI

Periods	Regression functions	df	r	R-sq
City of Porto-Novo				
Annual	$LST = 28.300 - 1.489 \times NDVI$	248	-0.138	0.019
Jan-March	$LST = 28.798 + 1.757 \times NDVI$	61	0.112	0.012
April-June	$LST = 28.721 - 1.784 \times NDVI$	58	-0.22	0.048
July-Sept	$LST = 26.536 - 0.078 \times NDVI$	58	-0.026	0.0007
Oct-Dec	$LST = 28.300 - 0.716 \times NDVI$	58	-0.074	0.005
City of Parakou				
Annual	$LST = 28.110 - 1.398 \times NDVI$	216	-0.427	0.182
Jan-March	$LST = 29.515 - 1.315 \times NDVI$	48	-0.036	0.0013
April-June	$LST = 28.462 + 0.332 \times NDVI$	48	0.037	0.0014
July-Sept	$LST = 25.615 - 0.135 \times NDVI$	48	-0.048	0.0023
Oct-Dec	$LST = 28.547 - 4.091 \times NDVI$	48	-0.427	0.183

## Discussion

Greenery seasonal dynamics is more crucial for monitoring urban environment. We have attempted to understand the quarterly variability of NDVI using Landsat data to get more accurate information about the vegetation effects on urban land surface temperature reduction. Along this, NDVI varied from year to year in cities. A similar trend was globally observed with a decreasing trend from 2000 to 2011 and an increasing trend from 2011 to 2020. This finding can be explained by the national commitments to vegetate cities with the governmental project “10 million people – 10 million trees” instituted in 2013 (MCVDD, 2019). Therefore, this governmental project would awake spirit in good management and safeguard of the existing urban vegetation in these cities. Likewise, the project goal was that each inhabitant must plant at least one tree per year with careful monitor. Moreover, the particular changes among cities could be the response to the vegetation density and the plant species diversity changes in these cities (Kinyanjui, 2011; Zohoun et al., 2020; Sehoun et al., 2021). The land surface temperature (LST) followed similar trend for the combined cities as well as for the cities taken individually. This result

can be attributed to the changes in vegetation index. The land surface temperature was globally in decreasing (Fig. 3) with increasing NDVI (Fig. 2). The same findings were mentioned by Zhao et al. (2019) on factors affecting the decreased temperature. Likewise, a study in the horn of Africa has also found a negative relationship between air temperature and vegetation cover (Ghebregabher et al., 2020). However, although efforts were born to improve greenery in Beninese cities, a significant linear trend of LST was occurred in the city of Porto-Novo. We can justify this finding by the weakness in the management and monitoring of planted trees. It can be likely due to the density of built area (lack of areas for trees plantation), introduction of plant species without resident contentment (Osseni et al., 2014). Therefore, the NDVI trend was also somehow degraded in this city (Fig. 2).

A very significant difference was shown between seasonal variations of vegetation index (NDVI). This finding can be justified by season-wise vegetation statuses in the cities. Therefore, not only urban vegetation state varied according to space but also it varied with respect to climatic period (Malik et al., 2020). For combined cities as for individual studied cities, the period of January-March had the significant low mean NDVI value. This result can be similar to Piao et al. (2019), who found a low value for the period of June-July on the city of Nanjing (China). Then, the authors explained this difference by the interference of cloud and rain occurring. Other study found that during January to April in India, the vegetation showed very low NDVI because during this period very few leaves were available to reflect NDVI signature (Malik et al., 2020). The periods from January to March and April-June respectively in south and centre Benin, refer to dry season when overall vegetation features are in water stress (Ahokpossi, 2018; Oussou et al., 2022). Moreover, a link could be also made between the vegetation index and the vegetation phenology (Su et al., 2020). Likewise, it was also demonstrated that the seasonal variability is more important to describe the variation in phenological expression of urban vegetation (Li et al., 2017; Piao et al., 2019). Thereby, Su et al. (2020) went further to make relationship between season period and canopy phenology that are very determining in vegetation index importance. Along this, what should be the impacts of both seasonal and NDVI variability on land surface temperature in the study areas?

Annual mean LST builds a moderate negative correlation with NDVI in the combined cities as well as in separate cities. The similar result was found by many researchers who demonstrated an inverse correlation relationship between LST and NDVI whatever the land use types and polygons (Yue et al., 2007; Richards et al., 2020). But although the cooling effects that greening can procure to urban areas were widely demonstrated, the vegetation physiology due to environmental conditions can have strong impacts on this service (Norton et al., 2015). Thereby, the assessing seasonal variability in relationship between LST and NDVI, revealed a moderate positive correlation between LST and NDVI in the periods of January to March and April to June respectively for Porto-Novo (Figs. 5) and Parakou (Fig. 6). This finding expressed that the increase in urban vegetation in these cities respectively for these above periods could not decrease the LST. We can justify this finding by the occurrence of dry season at these periods respectively in Porto-Novo and in Parakou. Therefore, the vegetation that would be experiencing water stress cannot have great rates of evapotranspiration, then can lead higher surface temperature (Norton et al., 2015). Porto-Novo and Parakou usually experience respectively strong dry period with high water

stress from January to March (Ahokpossi, 2018) and April to June (Ahokpossi, 2018; Lanmandjèkpogni et al., 2018). So, a wise investment in water availability such as urban greening irrigation would be more adequate for ensuring a sustainable urban surface temperature mitigation as well as other ecosystem service providing (Norton et al., 2015).

## Conclusion

No significant linear trends were observed for annual vegetation index (NDVI) and land surface temperature (LST) in the study areas. The NDVI tendency decreased from 2000 to 2011 and increased from 2011 to 2020. The city of Porto-Novo showed a moderate particularity in terms of linear significance in trend for land surface temperature (LST). The period of January to March showed the lowest values of mean NDVI. We also found a negative relationship between annual mean LST and NDVI. But, a seasonal positive correlation between LST and NDVI was shown for the periods of January-March and April-June respectively in Porto-Novo and Parakou. These periods coincide with the dry seasons and the vegetation usually express water stress. Thereby, a wise urban management for water availability such as greenery irrigation would be more helpful for ensuring sustainable urban surface temperature mitigation.

## Declarations

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### Author contribution statement

**Akakpo Bokon Alexis:** Conceptualization, Methodology, Data curation, Formal analysis, Validation, Writing - original draft, Writing - review & editing, Visualization. Professor **Okhimamhe Appollonia A.:** Principal supervisor of the work. Professor **Orekan A.O. Vincent** associate supervisor of the work.

### Declaration of competing interest

Authors declare no conflict of interest

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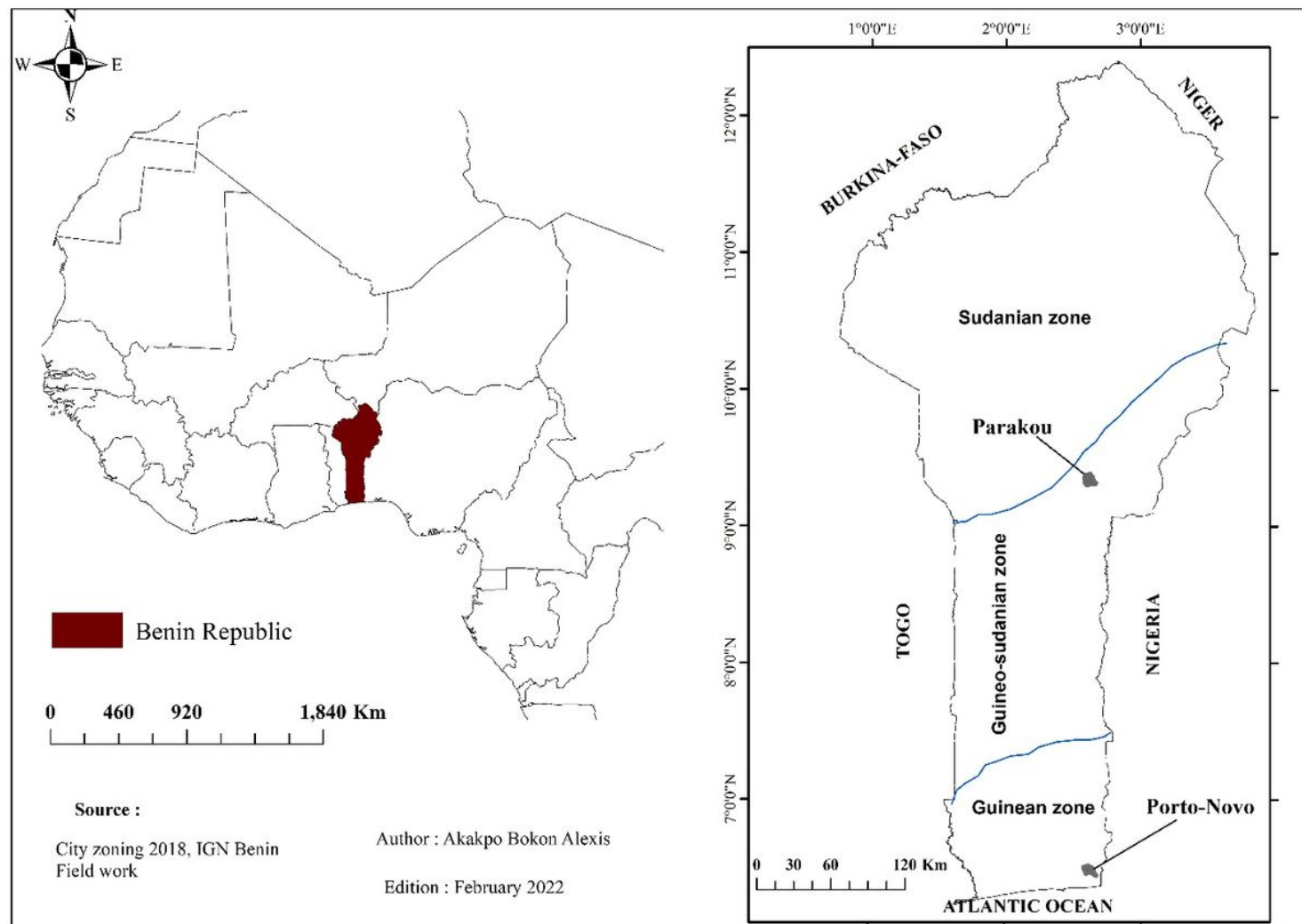
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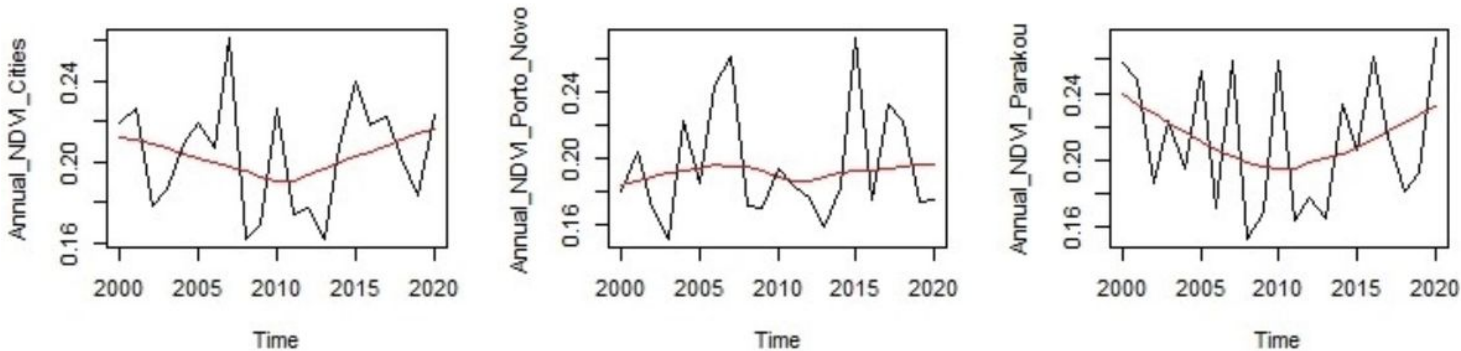
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## Figures



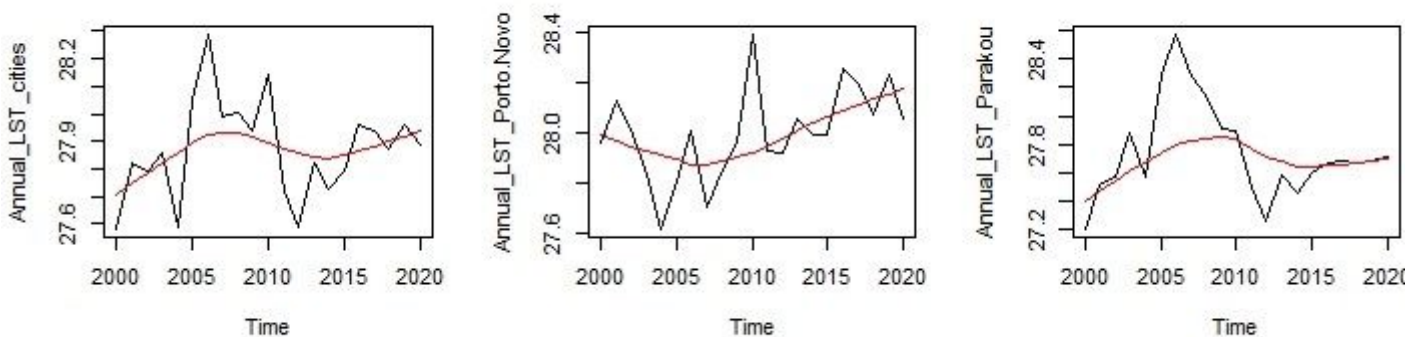
**Figure 1**

Location map of the case study cities



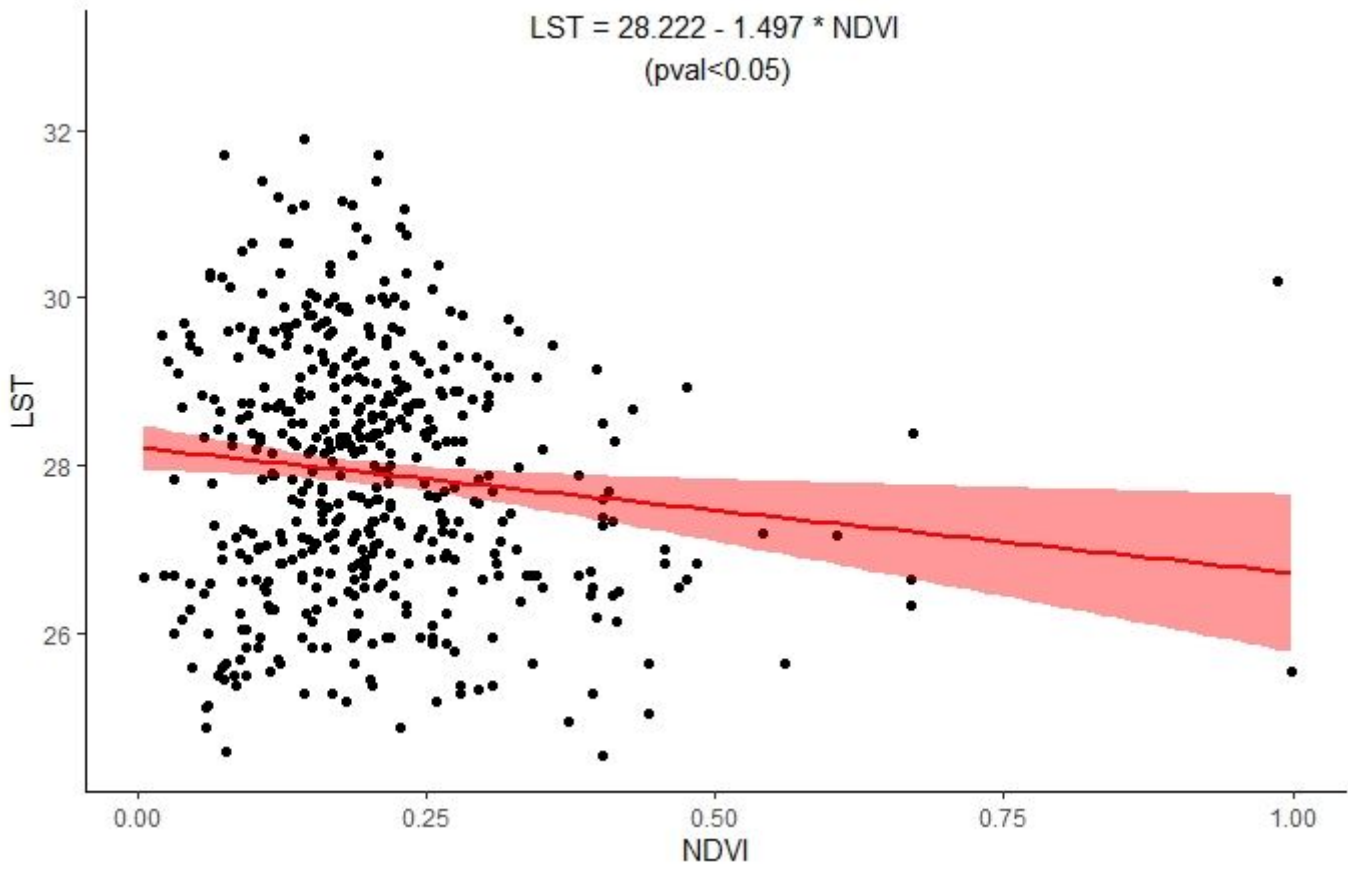
**Figure 2**

Trend of annual normalized difference vegetation index (NDVI) from the years 2000 to 2020



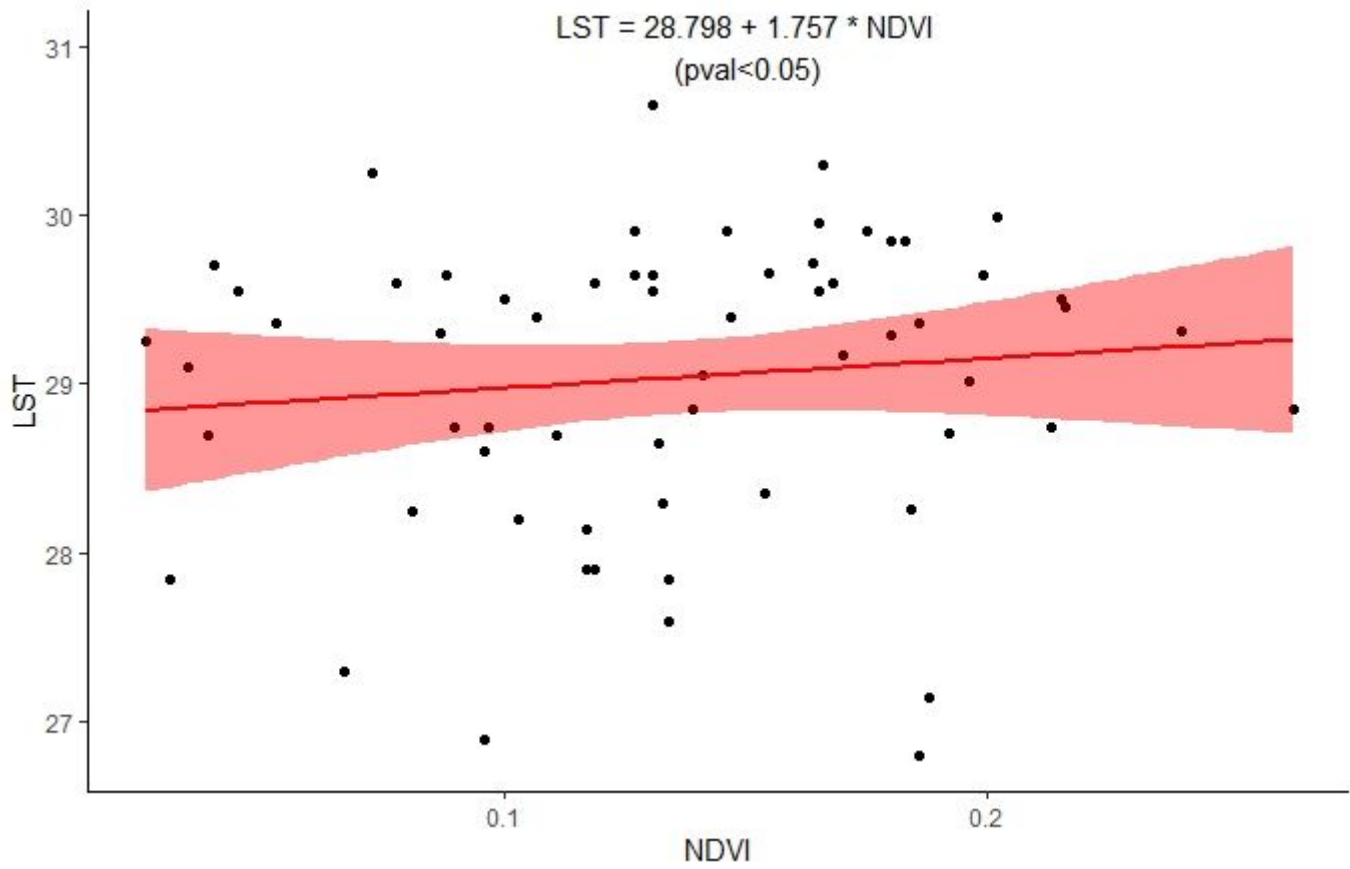
**Figure 3**

Trend of annual land surface temperature (LST) from the years 2000 to 2020



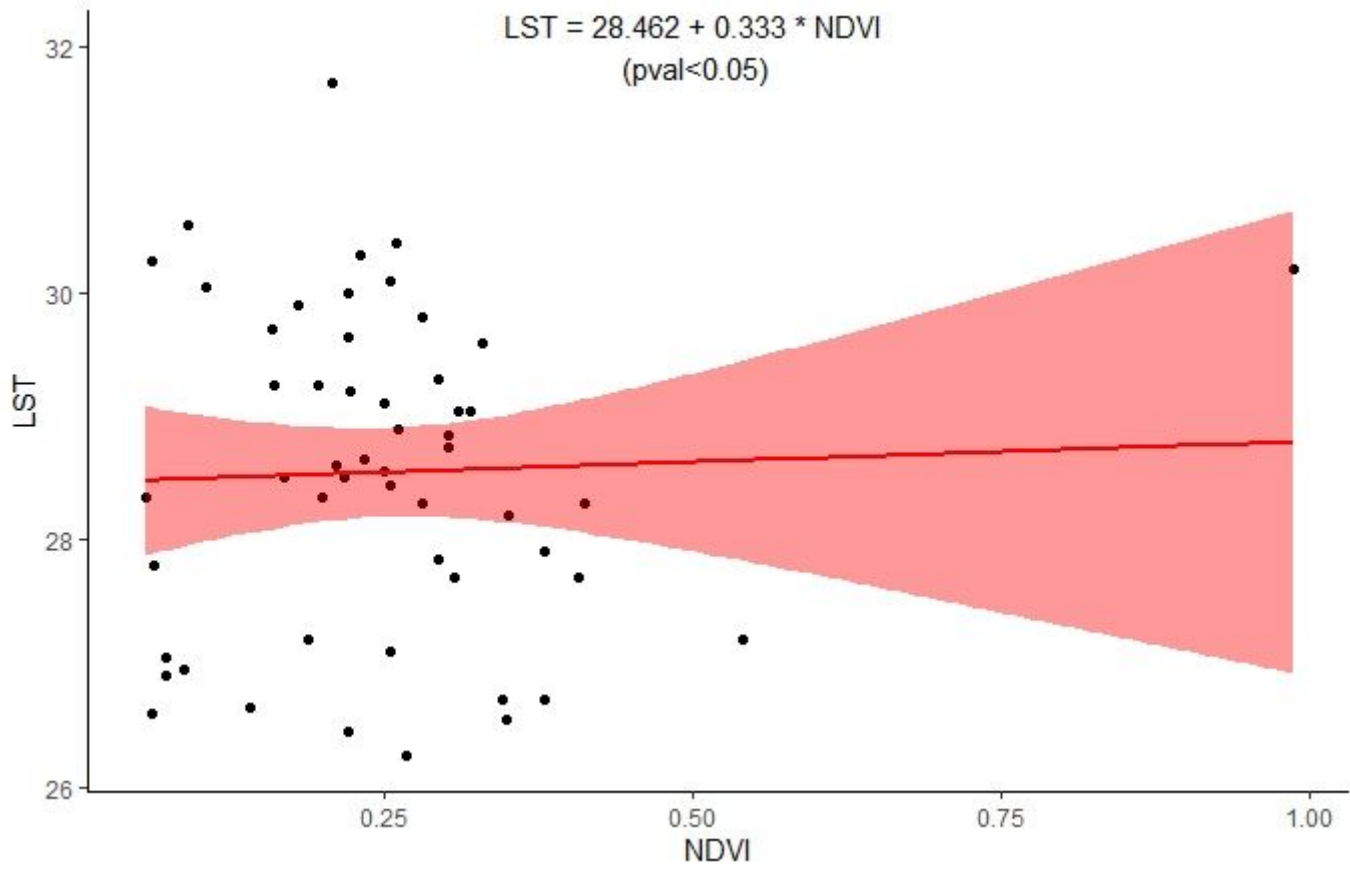
**Figure 4**

Scatter plot linear regression of mean LST and NDVI for both cities



**Figure 5**

Scatter plot linear regression of mean LST and NDVI for January-March period in Porto-Novo



**Figure 6**

Scatter plot linear regression of mean LST and NDVI for April-June period in Parakou