ABSTRACT: For climate change scenarios, there is a high degree of complexity, with impacts on the future availability of water resources. In this context, studies related to changes in rainfall time series are essential in order to identify environmental vulnerability. The objective of this study was to analyze trends in the rainfall regime, number of rainy days and temperature for stations located at different continentality and altitude conditions in Northeast Brazil. Meteorological data of the Instituto Nacional de Meteorologia were used, being classified according to the location in relation to the continent: coastal strip (14 stations), strip of 150-300 km up to 300 m altitude (14 stations) and above 300 m (five stations), between 400 and 600 km from the coast up to 300 m (four stations) and above 300 m (eight stations). All the 45 stations used have a historical series with a data period of more than 30 years. The trend of rainfall and rainy days was obtained through the Mann-Kendall and regression analyses, at significance levels of 0.01 and 0.05, respectively. There were trends of reduction in the number of rainy days, in the coastal strip, as well as reduction in rainfall and rainy days, both in the strip of 150-300 km from the coast and in the Sertão region, with no significant effect of continentality in the strip of 400-600 km from the coast. For temperature, except for Maceió, AL, Brazil, there is a trend of increase in near future.

Key words: continentality, rainfall modeling, climate variability

Tendências da precipitação pluvial e da temperatura no Nordeste brasileiro

RESUMO: Nos cenários de mudanças climáticas verifica-se elevada complexidade, estando previstos impactos na disponibilidade futura de recursos hídricos. Nesse contexto, estudos relacionados às alterações nas séries temporais do regime pluvial são essenciais para identificar uma vulnerabilidade ambiental. Assim, objetivou-se analisar as tendências no regime pluvial, no número de dias de precipitação e temperatura em estações situadas em diferentes condições de continetalidade e altitude no Nordeste brasileiro. Foram utilizados dados meteorológicos do Instituto Nacional de Meteorologia, sendo classificados de acordo a faixa de localização em relação ao continente: faixa litorânea (14 estações), na faixa de 150-300 km até 300 m de altitude (14 estações) e acima de 300 m (cinco estações), entre 400 e 600 km do litoral até 300 m (quatro estações) e acima de 300 m (8 estações). Todas as 45 estações utilizadas possuem série histórica com período de dados superior a 30 anos. A tendência da precipitação e dias de chuva foi investigada através da análise Mann-Kendall e de regressão, para os níveis de significância de 0,01 e 0,05, respectivamente. Foram identificadas tendências de redução no número de dias de precipitação, na faixa litorânea, bem como redução na precipitação e nos dias com chuva, tanto na faixa de 150 a 300 km do litoral, quanto na região do Sertão, não havendo efeito significativo da continentalidade na faixa de 400 a 600 km do litoral. Para a temperatura, com exceção de Maceió, AL, Brasil, observa-se tendência de incremento em futuro próximo.

Palavras-chave: continentality, modelagem da precipitação, variabilidade climática
### Introduction

Severe droughts and floods are phenomena related to rainfall variability and can significantly affect agricultural production and the environment, particularly in arid and semiarid regions (Pinheiro et al., 2013; Sun & Ma, 2015).

The Intergovernmental Panel on Climate Change (IPCC, 2013) emphasizes that global warming has caused higher variability in rainfall regimes. This fact has been observed in the rainfall regimes in Brazil and worldwide, induced by the occurrence of extreme natural phenomena (Carretas, 2014).

Marcuzzo et al. (2012) analyzed data from 37 pluviometric stations with 30 years of data in the state of Goiás, Brazil, and observed that every month had a variation greater than the annual average, indicating greater dispersion. Pingale et al. (2014), investigating the trend of rainfall in 33 cities in the semiarid region of Rajasthan, India, found positive and negative trends.

Montenegro & Ragab (2010) found that the increase in temperature and the greater temporal variability in rainfall caused changes in the availability of water resources, leading to the reduction of surface runoff and recharge.

Evaluations of trends in rainfall regimes are essential in agricultural planning and preservation of ecosystems (Back et al., 2012), since rainfall is the climatic variable with highest spatial-temporal variability, conditioned to the factors of altitude, latitude, continentality and the dynamics of air masses in the atmosphere (Buriol et al., 2004).

Thus, this study aimed to analyze trends in the total annual rainfall, number of rainy days and temperature under different conditions of continentality and altitude in Northeast Brazil.

### Material and Methods

The area of this study is the Northeast region of Brazil, where records of daily rainfall and number of rainy days (NRD) were analyzed in the pluviometric stations of the Instituto Nacional de Meteorologia (INMET, 2018). The meteorological stations were organized in relation to the distance from the coast and according to the altitude intervals.

Stations of the coastal region, in the strip of 150-300 km up to 300 m altitude (14 stations) and above 300 m (five stations), between 400 and 600 km from the coast up to 300 m (four stations) and above 300 m (8 stations) (Table 1) were selected, obeying the continuity and with a minimum period of records of 30 years of data as recommended by WMO (1989).

A descriptive statistical analysis of central tendency (mean), dispersion (coefficient of variation) and adherence to normal distribution were performed. To verify the trend of the characteristics analyzed, the non-parametric Mann–Whitney test was used, considering a significance level of 0.05. When the number of series was small (< 10), the test of runs was used to verify the trend.

### Table 1. Geographic location and descriptive statistics of the stations analyzed in the Northeast region

| ID  | Station  | Latitude (º) | Longitude (º) | Altitude (m) | Mean annual rainfall (mm year⁻¹) | CV (%) | Mean NRD (n días) | CV (%) | Mean Temperature (ºC) | CV (%) |
|-----|----------|--------------|---------------|--------------|---------------------------------|--------|------------------|--------|----------------------|--------|
| 5   | João Pessoa - PB | -7.10 | -3.86       | 7.4          | 1917               | 32.5   | 183              | 18.6   | 26.5                 | 2.1    |
| 6   | Macau - RN  | -5.15 | -3.67       | 32.0         | 573                | 42.6   | 64               | 32.8   | 27.4                 | 2.7    |
| 7   | Maciel - AL  | -9.66 | -3.70       | 64.5         | 1952               | 23.9   | 206              | 17.5   | 25.3                 | 2.6    |
| 8   | Natal - RN   | -5.91 | -3.50       | 48.6         | 1620               | 28.6   | 161              | 16.2   | 26.3                 | 1.8    |
| 9   | Paragiba - PI | -3.08 | -4.17       | 79.5         | 1229               | 47.2   | 108              | 27.4   | 27.0                 | 2.3    |
| 10  | Porto de Pedras-AL | -9.18 | -3.63       | 50.0         | 1782               | 20.7   | 190              | 14.7   | 26.1                 | 1.8    |
| 11  | Recife - PE   | -8.05 | -3.49       | 10.0         | 2291               | 21.8   | 218              | 8.7    | 25.7                 | 1.5    |
| 12  | Salvador - BA | -13.01 | -3.83     | 51.4         | 1944               | 23.1   | 201              | 13.4   | 25.5                 | 1.7    |
| 13  | São Luiz - MA  | -2.53 | -4.42       | 50.9         | 2267               | 26.8   | 140              | 20.7   | 26.6                 | 2.1    |
| 14  | Turiaçu - MA   | -1.56 | -4.36       | 44.1         | 2207               | 25.6   | 159              | 18.9   | 26.9                 | 1.8    |

150 to 300 km away from the coast (up to 300 m altitude)

| ID  | Station  | Latitude (º) | Longitude (º) | Altitude (m) | Mean annual rainfall (mm year⁻¹) | CV (%) | Mean NRD (n días) | CV (%) | Mean Temperature (ºC) | CV (%) |
|-----|----------|--------------|---------------|--------------|---------------------------------|--------|------------------|--------|----------------------|--------|
| 15  | Bacabal - MA  | -4.21 | -4.74       | 25.1         | 1722               | 27.8   | 102              | 46.1   | 27.5                 | 3.6    |
| 16  | Caldeirão - PI | -4.28 | -4.18       | 160.0        | 1330               | 37.9   | 85               | 40.0   | 27.4                 | 2.8    |
| 17  | Coxias - MA   | -4.86 | -4.33       | 103.6        | 1592               | 28.7   | 97               | 29.9   | 27.4                 | 2.7    |
| 18  | Crato - CE     | -5.16 | -4.06       | 296.8        | 748                | 40.4   | 80               | 26.3   | 26.9                 | 3.7    |
| 19  | Cruzeta - RN   | -5.43 | -3.68       | 226.5        | 729                | 35.7   | 61               | 43.5   | 27.9                 | 2.2    |
| 20  | Iguatu - CE    | -6.26 | -3.29       | 217.7        | 998                | 28.6   | 80               | 31.3   | 26.9                 | 2.4    |
| 21  | Itaberaba - BA | -12.51 | -4.20     | 249.9        | 681                | 33.0   | 117              | 29.1   | 24.5                 | 2.5    |
| 22  | Patos - PB     | -7.01 | -3.26       | 249.1        | 748                | 34.5   | 66               | 37.9   | 27.5                 | 2.2    |
| 23  | Paulo Afonso - BA | -9.36 | -3.82   | 252.7        | 544                | 32.5   | 94               | 31.9   | 25.9                 | 1.9    |
| 24  | Quixeramobim - CE | -5.16 | -3.82   | 79.5         | 756                | 35.2   | 86               | 29.1   | 26.8                 | 4.4    |
| 25  | São Gonçalo - PB | -6.75 | -3.82      | 233.1       | 926                | 37.4   | 78               | 33.3   | 26.7                 | 2.7    |
| 26  | Seridó (Caiçara) - RN | -6.46 | -3.70 | 169.9       | 644                | 41.6   | 71               | 32.4   | 28.3                 | 2.3    |
| 27  | Teresina - PI  | -5.08 | -4.21       | 74.4         | 1234               | 31.8   | 101              | 25.7   | 27.1                 | 2.6    |
| 28  | Zé Doca - MA   | -3.26 | -4.65       | 45.3         | 1809               | 18.0   | 139              | 23.0   | 26.9                 | 3.1    |

150 to 300 km away from the coast (above 300 m altitude)

Continues on the next page
distribution was performed using the Kolmogorov-Smirnov test (KS) at p ≤ 0.05, for the records of total annual rainfall and average annual temperature, and for the number of rainy days (NRD) values.

The coefficient of variation (CV) was classified according to the criterion of Warrick & Nielsen (1980), which considers the degree of variability as low (CV < 12%), medium (12 ≤ CV ≤ 60%) and high (CV > 60%).

The geographic locations of the INMET pluviometric stations in which the trend of rainfall, NRD and temperature were evaluated are shown in Figure 1.

The studied period was from 1961 to 2017. Each rainfall series was evaluated for the trend of Total Annual Rainfall (TAR), Total Annual Number of Rainy Days (TANRD), average temperature (Temp) and NRD. Its temporal evolution was verified by means of the Deviations of Rainfall from the Mean (DRM) and of the Deviations of Rainy Days from the Mean (DRDM), in order to identify any changes in the climatological behavior over the years.

The non-parametric Mann-Kendall test is widely used to detect monotonic trends in hydrometeorological time series. It is based on the null hypothesis (H₀), in which the data are identically distributed (no trend), and on the alternative hypothesis (Hₐ), in which the data follow a monotonic trend in the time series. The test confirms the existence of a positive or negative trend according to the S test statistic for a given level of confidence (Pinheiro et al., 2013; Xu et al., 2018).

The test statistic (S) was applied according to the methodology of Pinheiro et al. (2013) and Xu et al. (2018), described by Eq. 1:

$$S = \sum_{i=1}^{n-1} \sum_{j=k+1}^{n} \text{sign}(x(j) - x(i))$$

where:
- $x(i)$ - time series of $i = 1, 2, 3, ..., n - 1$;
- $x(j)$ - time series of $j = i+1,..., n$, $x(j)$ is higher than $x(i)$; and,
- $n$ - length of the data set record.

Each point $x(i)$ is used as the reference point of $x(j)$, and the results are recorded as sign(x) (1, θ > 0; 0, θ = 0; -1, θ < 0).

For a long time series, the statistical value S can be transformed into Z, according to the following conditions: $Z = S - 1/\sqrt{\text{Var}(S)}$, $S > 0$; $S = 0$; $S+1/\sqrt{\text{Var}(S)}$ = $0$. When $-1.96 \leq Z < 1.96$, the null hypothesis (H₀) is accepted, which indicates that there is no trend in the time series. The trend is significant at 0.95 confidence level if $|Z| > 1.96$, and at 0.99 confidence level if $|Z| > 2.58$. A positive Z value indicates that the sequence

### Table 1

| ID  | Station         | Latitude | Longitude | Altitude (m) | Annual rainfall (mm year⁻¹) | CV (%) | NRD (n days) | CV (%) | Temperature (°C) | CV (%) |
|-----|-----------------|----------|-----------|--------------|------------------------------|--------|--------------|--------|-----------------|--------|
| 34  | Balsas - MA     | -7.53    | -46.03    | 259.4        | 1265                         | 20.2   | 91           | 25.3   | 26.5            | 3.4    |
| 35  | Floriano - PI   | -7.67    | -43.01    | 123.3        | 971                          | 32.2   | 96           | 20.8   | 27.7            | 3.8    |
| 36  | Imperatriz - MA | -5.53    | -47.48    | 123.3        | 1473                         | 21.9   | 101          | 30.7   | 26.9            | 3.2    |
| 37  | Picos - PI      | -7.03    | -41.48    | 207.9        | 736                          | 31.4   | 74           | 28.4   | 27.8            | 4.1    |
| 38  | Barra - BA      | -11.08   | -43.16    | 401.6        | 661                          | 30.9   | 56           | 21.4   | 26.4            | 2.9    |
| 39  | Bom Jesus da Lapa - BA | -13.26 | -43.41    | 439.9        | 816                          | 24.1   | 75           | 18.7   | 25.8            | 3.5    |
| 40  | Carinhanha - BA | -14.28   | -43.76    | 450.2        | 802                          | 25.1   | 72           | 20.8   | 25.6            | 2.6    |
| 41  | Correntina - BA | -13.33   | -44.61    | 549.5        | 995                          | 25.0   | 80           | 16.3   | 24.1            | 3.2    |
| 42  | Ouricuri - PE   | -7.90    | -40.04    | 459.3        | 614                          | 41.0   | 75           | 30.7   | 25.8            | 4.5    |
| 43  | Paulistana - PI | -8.13    | -41.13    | 374.2        | 571                          | 40.8   | 57           | 31.6   | 26.9            | 3.6    |
| 44  | Petrolina - PE  | -9.38    | -40.48    | 370.5        | 472                          | 38.8   | 49           | 30.6   | 26.9            | 2.6    |
| 45  | Remanso - BA    | -9.63    | -42.10    | 400.5        | 634                          | 31.2   | 48           | 25.0   | 26.7            | 2.4    |

NRD – Number of rainy days; CV – Coefficient of variation

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**Figure 1.** Location of the stations in the Brazilian Northeastern states
has an upward trend, whereas a negative Z reflects a downward trend (Pinheiro et al., 2013; Xu et al., 2018).

After the trends are identified, their magnitude is analyzed using the Sen curvature test (Sen, 1968), according to Eq. 3:

$$\beta = \left( \frac{x_j - x_k}{j - k} \right) , \text{ for } i = 1, 2, 3, ..., n$$

where:

- $\beta$ - Sen's slope estimator. When the values are positive, the trend is positive, and when the values are negative, the trend is negative; and,

- $x_j$ and $x_k$ - are the values given at the times $j$ and $k$ ($j > k$), respectively.

The behavior of the annual time series of meteorological variables was evaluated by adopting the calculation of the moving averages, employing the order five for the data, aiming to avoid fluctuations and smoothing the data (Ferreira et al., 2015).

Future projection for rainfall and NRD was then carried out for the next 30 years based on the Mann-Kendall trend analysis and the $\beta$ (Sen’s slope) coefficient, up to 2047, assuming that no significant change will occur on the climate variability patterns in near future.

For geostatistical analysis, the classical statistical estimator of semivariance was adopted. After obtaining the semivariances, the Gaussian, spherical and exponential models were tested, and the one which best fitted to the experimental values according to the cross-validation and which produced standardized errors close to zero was chosen, according to Jackknifing criterion (Montenegro & Montenegro, 2006).

From the fit of the semivariograms, the spatial distribution analysis was performed through the Kriging method in order to map the climatic variables correlated in space (Montenegro & Montenegro, 2006), for the variables rainfall, number of rainy days and temperature.

The degree of spatial dependence was based on Cambardella et al. (1994), according to the following classification: strong (less than 25%), moderate (between 25 and 75%) or weak (above 75%) spatial dependence.

### Results and Discussion

The trends of Total Annual Rainfall (TAR), Total Annual Number of Rainy Days (TANRD), Mean temperature (Temp), Deviations of Rainfall from the Mean (DRM), Deviations of Rainy Days from the Mean (DRDM) and the respective Z values of the coastal strip of Northeast Brazil are presented in Figure 2.

The station of Parnaíba, PI, was the only one which showed a trend of reduction in TAR, with an estimated value of 118 mm for 30 years. The others had a trend of reduction for NRD in Caravelas, BA (23 days), Salvador, BA (17 days), Fortaleza, CE (19 days), Maceió, AL (28 days) and Canavieiras, BA (29 days).

**Figure 2.** Trend of rainfall (TAR), total annual number of rainy days (TANRD), temperature (Temp), deviations of rainfall from the mean (DRM) and Z values for the coastal strip of Northeast Brazil between 1961 and 2017.
Except for Maceió, AL, a trend of increase in the temperature was observed. These changes in the temperature pattern, according to Fall et al. (2011), may be associated with human activities, particularly related to excessive emissions of greenhouse gases, from agriculture, livestock farming and industry, regarded as the main cause of global warming (Mahlstein & Knutti, 2010).

The reductions in TANRD represent a higher concentration of rainfall, consequently resulting in the increase of its intensity and, therefore, in major damage risk to urban and rural environments, possibly increasing of floods events and soil degradation.

According to Silva et al. (2012), intense rainfall can aggravate erosion processes in the semiarid region in areas with shallow soils and with restricted natural drainage. In addition, knowledge on rainfall trend is important for the analysis of environmental impacts and dimensioning of hydraulic structures for flood control and urban drainage (Mello & Viola, 2013).

The stations within the strip of 150-300 km from the coast showed a trend only at altitudes up to 300 m, both for rainfall and total annual number of rainy days (TANRD), with increase of rainfall in Bacabal, MA, and reduction in Itaberaba, BA, Paulo Afonso, BA, and Quixeramobim, CE (Figure 3). The temperature tended to increase in these stations.

The trends for the stations of Quixeramobim, CE, Itaberaba, BA and Paulo Afonso, BA, until the year 2017 are of reduction in rainfall (121, 148 and 140 mm) and in rainy days (3, 21 and 24 days), as well as increase of temperature (0.9, 0.69 and 0.6 ºC), respectively.

Asfaw et al. (2018) found a reduction of up to 101 mm in the average annual rainfall in the Woleka basin in Ethiopia. Xavier et al. (2014) point out that the changes in rainfall...
regimes may threaten the biodiversity of Brazilian biomes, particularly the Caatinga, rich in fauna and flora, which has undergone significant impacts, partly due to the reduction in native vegetation, increasing the Areas Susceptible to Desertification (ASD).

According to Souza et al. (2015), there are more than 1,338 km² in Northeast Brazil classified as ASD. In these areas, desertification should be regarded as a complex environmental problem, which impacts the support capacity of ecosystems, and where studies analyzing rainfall trends and the number of rainy days are of great relevance to support sustainable environmental management.

The trend of increase and reduction in rainfall was observed for four stations, and the trend for NRD showed the same behavior. A trend of increase in the consecutive dry days was observed for the Agreste Pernambucano and Sertão Pernambucano mesoregions (Nóbrega et al., 2015). Great variability of rainfall and number of rainy days in Northeast Brazil was observed by Silva et al. (2011), who warn about the possible environmental and socioeconomic impact related to rainfed agriculture.

The coefficients of determination ($R^2$) of the equations for rainfall and rainy days were low and ranged from 0.02 to 0.48, but were significant for the trend analysis. These low values are due to high uncertainties in studies on the rainfall regime. Ferreira et al. (2015), Asfaw et al. (2018) and Xu et al. (2018) also found $R^2$ values always below 0.5 for the trend equations.

In the strip of 450-600 km from the coast for altitudes of up to 300 m, only the station of Balsas, MA, showed a trend of increase in NRD and temperature. For the altitude from 300 to 600 m, the station of Correntina, BA, Brazil, showed a trend of reduction in rainfall and increase of temperature (Figure 4).

Rainfall distribution in the Northeast region has an impact on the occurrence and location of extreme events (drought and floods). The more distant from the coast, the higher the rainfall variability and the lower the rainfall values.

The trend of change in the TAR, TANRD, DRM, DRDM from the mean and NRD was not clearly identified in Northeast Brazil, for the studied period, since the stations located on the coast showed a trend only for NRD isolated from rainfall. In the stations of the strip from 150 to 300 km, the trends were verified for number of rainy days, while in the strip from 450 to 600 km the stations showed isolated trends.

Xu et al. (2018) also report a lack of clarity in the rainfall trends in different magnitudes of topographic elevation. In addition, changes in the trend magnitudes in sites of low elevation may be associated to the topographic conditions of alluvial valleys and to orographic effects.

The high spatial and temporal variability of rainfall can be observed in the relative difference for rainfall and NRD in all localities, particularly in the strip from 150 to 300 km (Figure 3).

The different patterns of rainfall and NRD observed, despite having regional trends of increase and decrease in rainfall and rainy days, promoted changes in rainfall intensity for different localities in the context of continentality, and it is necessary to reevaluate water security in the region. Sayemuzzaman & Jha (2014) and Steinke & Barros (2015) highlight the importance of understanding the climate, in the aspect of rainfall trend, in order to support agricultural planning and thus mitigate the losses of biodiversity and crop yields and, consequently, alleviate the socioeconomic impacts.

Given the greater complexity in rainfall and NRD patterns, with trends of both increase and decrease and greater variability, observed through the coefficient of variation, classified as low to medium with minimum values (21.3 and 8.7%) and maximum values (52.5 and 46.1%), respectively, as well as a lower temperature variability, with CV values ranging from 1.5 to 4.5% (classified as low) (Table 1), the complementary spatial geostatistical analysis of the future scenarios by kriging was performed only for rainfall and NPD.

The experimental and theoretical semivariograms, fitted and the spatial distribution of rainfall for the current scenario (Rainfall, P – Current and NRD, D – Current) and estimated scenarios for 30 years in the future (Rainfall, P – 30 years and NRD, D – 30 years), are shown in Figures 5A and B, respectively. All experimental semivariograms were adequately fitted to the exponential model, with the respective fitting parameters of the semivariogram (nugget effect ($C_0$), sill ($C_0$), range (Ao) and coefficients of determination ($R^2$)), showed in Figure 5.

The two semivariograms of rainfall and NRD showed strong spatial dependence. The validation was performed according to the standard deviation of the residual statistics (0.83, 0.75, 0.78 and 0.83) and for mean errors (-0.034, -0.047, -0.038 and -0.047), for rainfall and days until 2017, rainfall and days estimated for 30 years, respectively.

The ranges (Ao) and $R^2$ values obtained for rainfall in the current scenario (1338 km and $R^2 = 0.81$) were higher than those corresponding to the future scenario (854 km and $R^2 = 0.69$). The same behavior was observed for NRD with ranges for the

![Figure 4. Trend of rainfall (TAR), total annual number of rainy days (TANRD), temperature (Temp), deviations of rainfall from the mean (DRM) and Z values for the strip of 450-600 km from the coast of Northeast Brazil between 1961 and 2017](image-url)
Trends of rainfall and temperature in Northeast Brazil

Figure 5. Semivariograms of rainfall (A) and number of rainy days (NRD) (B) for the Brazilian northeast region for the current scenario (1961–2017) and future scenario (1961–2047) and isoline maps for rainfall (mm) in the current (C) and future (E) scenarios and NRD in the current (D) and future (F) scenarios for the Brazilian northeast region, obtained by kriging.

The pattern of rainfall and temperature trends in Northeast Brazil is crucial for understanding the region’s climate variability. Silva et al. (2012) observed high variability in rainfall and number of rainy days in the dry and rainy seasons, influencing the agricultural production in the microregions located in the semiarid areas.

Figure 5E, projecting the trend for 2047, shows a reduction in rainfall mainly in the central portion of the Northeast region, in the area encompassing the Sertão in relation to the current scenario (Figure 5C).

The trend of reduction in rainfall accompanies the reduction of rainy days in the current scenario (Figure 5D) for the Northeastern Sertão in relation to the future scenario (Figure 5F). These results show changes in rainfall patterns more concentrated in the Northeastern Sertão area, which may cause intensification of water deficit.

Assessing the behavior of rainfall, Nóbrega et al. (2016) found a trend of reduction in rainfall in Pernambuco State. The authors also emphasize that the irregular spatial-temporal distribution of rainfall in the Northeastern Sertão is associated to prolonged periods of water restriction and negatively influences the water level of rivers, turning this region more vulnerable to extreme events.

Conclusions

1. The central portion of the Northeast region has higher trend of variation in rainfall, number of rainy days and temperature.
2. There will be greater spatial variability of rainfall and number of days without rain for Northeast Brazil, especially for the central region encompassing the Sertão.
3. The trend analysis shows reduction in the number of rainy days on the Northeastern coast, while in the central region there is reduction in both rainfall and number of rainy days, and trend of increase in temperature for regions more distant from the coast.

4. Greater increases in temperature and reductions in total annual rainfall and number of rainy days are observed in Northeast Brazil.

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