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23 **ABSTRACT**

24 **Background:** Coastal cities like Dar es Salaam face an increasing vulnerability to the adverse
25 impacts of climate variability, including urban flooding, heat stress, and altered water
26 availability. Examining the temporal evolution of key climatic variables is critical to informing
27 adaptive strategies and promoting sustainable urban development.

28 **Methods:** Decadal seasonal trends in rainfall, daytime and nighttime temperatures, and relative
29 humidity were analyzed using monthly data from January 2014 to October 2024, sourced from
30 the Tanzania Meteorological Authority. We applied the non-parametric Mann-Kendall trend
31 test and Sen's slope estimator to detect and quantify monotonic trends across five defined
32 seasons.

33 **Results:** We observed statistically significant trends across several seasons. Rainfall during the
34 long dry season (JJA) had a slight increase, with a Sen's slope of +0.39097 mm/year and a p-
35 value of 0.00457, suggesting a slight deviation from typical seasonal dryness. Daytime
36 temperatures during JJA declined (Sen's slope = -0.03675°C/year; p = 0.00121), and nighttime
37 temperatures during the short dry season (JF) also showed a significant decrease (Sen's slope
38 = -0.04000°C/year; p = 0.04445). Relative humidity demonstrated only minor, statistically
39 insignificant fluctuations across all seasons, with the highest z-value recorded in OND.

40 **Conclusion:** The findings highlight shifting climatic patterns in Dar es Salaam that diverge
41 from conventional expectations, including increased precipitation during dry periods and
42 seasonal cooling. Although the detected trends are modest in absolute magnitude, their
43 statistical significance is noteworthy because consistent directional changes, however small,
44 accumulate over time and may signal early-stage climatic shifts that warrant monitoring.

45

46 **Keywords:** Climate variability, Mann-Kendall trend test, Sen's slope, Urban resilience, Dar es
47 Salaam

48

49 **1. Introduction**

50 Climate change is increasingly recognized as one of the most critical challenges of the 21st
51 century, driving long-term shifts in global temperatures and weather patterns, primarily due to
52 the retention of solar heat in the Earth's atmosphere (Kakkar et al., 2022). It is marked by rising
53 global temperatures and an increase in the frequency and severity of extreme weather events
54 such as droughts, floods, and heatwaves. This was also broadly highlighted at the global scale
55 in the IPCC synthesis report (IPCC, 2023). Projections indicate that water availability and
56 annual average runoff could decline by 10–30% by the mid-21st century, exacerbating water
57 insecurity in many regions (Kakkar et al., 2022). These changes highlight the critical
58 importance of examining historical climate patterns to guide the development of effective
59 adaptation strategies, particularly in vulnerable regions where socio-economic systems are
60 highly susceptible to climatic stressors (Gadedjisso-tossou & Adjegan, 2021).

61 IPCC synthesis report highlights that human activities have driven rapid and widespread
62 changes in the biosphere, cryosphere, ocean, and atmosphere. These changes have caused
63 significant losses and damages, particularly affecting vulnerable communities that have
64 contributed the least to climate change (IPCC, 2023). Tanzania is particularly vulnerable to the
65 adverse impacts of climate change, with seasonal variations in recorded rainfall and
66 temperature trends observed across many regions of the country (Mbawala et al., 2024).

67 Several studies in Tanzania have used the Mann-Kendall test and Sen's slope estimator to
68 analyze climate trends (Mbawala et al., 2024; Mugabe et al., 2024; Ndabagenga et al., 2023).
69 These studies have covered various aspects, including extreme temperature changes, extreme
70 precipitation indices, agricultural impacts in specific districts and hydro-climatic trends in river
71 catchments (Sigalla et al., 2023).

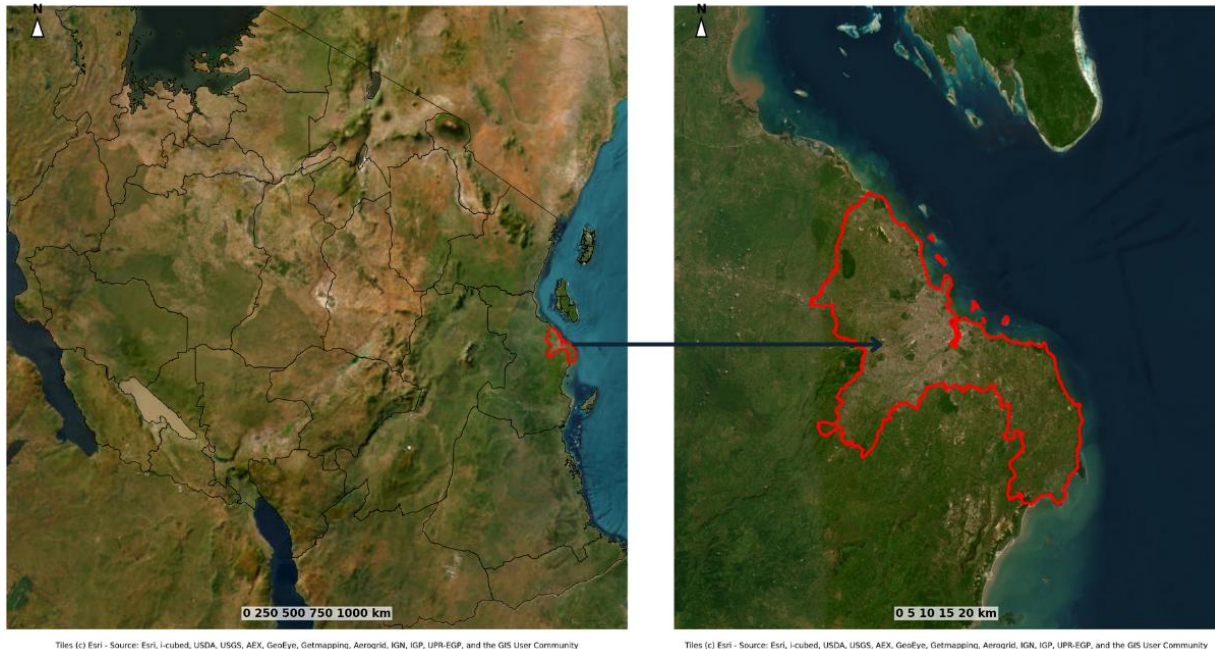
72 However, there remains a notable gap, as no recent studies have specifically examined the
73 decadal seasonal trends of key climatic variables such as temperature, rainfall, and relative
74 humidity within Dar es Salaam. Decadal trends in temperature, relative humidity, and seasonal
75 rainfall are crucial for comprehending recent shifts in climatic patterns. This study especially
76 examines possible changes in climate variability during the last ten years (2014-2024), in
77 contrast to earlier research. By doing this, it highlights how adaptation strategies based on local
78 climate data are essential for controlling climate-related risks and guaranteeing that solutions
79 are adapted to the unique requirements and vulnerabilities of communities at the local level.

80 This paper contributes to the growing discourse on urban climate resilience by offering a
81 comprehensive, decade-long analysis of climatic variability in Dar es Salaam, Tanzania’s most
82 populous coastal city. By focusing on seasonal trends in temperature, rainfall, and relative
83 humidity from 2014 to 2024, the study provides context-specific insights that are essential for
84 understanding emerging climate patterns and guiding adaptive planning.

85 2. Materials and Methods

86 2.1 Study Area

87 Dar es Salaam, Tanzania’s largest city and economic hub, is located along the Indian Ocean
88 coast between latitudes 6°36'S and 7°00'S and longitudes 39°00'E and 39°17'E. It has a humid
89 tropical climate, characterized by bimodal rainfall patterns with peaks in March–May and
90 October–December. The city experiences annual temperatures ranging from 18°C to 34°C and
91 receives an average of approximately 1,100 mm of rainfall per year (TMA, 2023). As a rapidly
92 urbanizing coastal city, Dar es Salaam is highly susceptible to climate-related challenges such
93 as flooding, heat stress, and water insecurity, making it a critical case for localized climate
94 trend analysis. Figure 1 presents a panel map illustrating the study area, with a map of Tanzania
95 on the left and an arrow pointing to a more detailed map of Dar es Salaam on the right.



96

97 *Figure 1: Map of Study Area (Author’s own contribution)*

98

99 2.2 Study design and data source

100 A retrospective, ecological time-series design was used to examine ten-year trends in climatic
101 variables across five seasons. Monthly climate data for Dar es Salaam, spanning January 2014

102 to October 2024, were sourced from the TMA. This dataset included rainfall (mm), daytime
 103 and nighttime temperatures (°C), and relative humidity (%). The TMA pre-validated the
 104 complete data, which required no imputation or correction, thus ensuring consistent and
 105 reliable trend analysis.

106 2.3 Data Analysis

107 2.3.1 Classification of seasons

108 Analysis was conducted for the five climatological seasons recognized for Dar es Salaam: JF
 109 (Short Dry Season), MAM (Long Rainy Season), JJA (Long Dry Season), S (Transitional
 110 Period), and OND (Short Rainy Season)

111 *Table 1: Classification of seasons*

Season ID	Months within the Season	Season Name
JF	January and February	Short Dry Season
MAM	March, April, and May	Long Rainy Season
JJA	June, July and August	Long Dry Season
S	September	Transition Period
OND	October, November and December	Short Rainy Season

112

113 2.3.2 Mann Kendall test and Sen's slope estimator

114 The non-parametric Mann-Kendall (MK) trend test was employed to detect the presence of
 115 monotonic trends in the time series data without requiring the data to follow any specific
 116 distribution. This method is particularly suitable for environmental and climatological data
 117 where non-normality and missing values may be present. The corresponding magnitude and
 118 direction of the trends were estimated using Sen's slope estimator. All statistical analyses were
 119 conducted at a 5% significance level ($\alpha = 0.05$), and trends were interpreted using the computed
 120 S-statistics, variance, Z-scores, p-values, and slope estimates.

121 In the MK test, the S-statistic was calculated as follows:

$$122 \quad S = \sum_{i=1}^{n-1} \sum_{j=i+1}^n \text{sgn}(x_j - x_i)$$

123 where $\text{sgn}(x_j - x_i)$ is the sign function, returning +1, 0, or -1 depending on whether the
 124 difference is positive, zero, or negative, respectively.

125 For large sample sizes ($n > 10$), the variance of S was computed using:

126
$$\text{Var}(S) = \frac{n(n-1)(2n+5) - \sum t_p(p-1)(2p+5)}{18}$$

127 where t_p denotes the number of ties of extent p .

128 The standardized test statistic Z was then derived as:

129
$$Z = \begin{cases} \frac{S-1}{\sqrt{\text{Var}(S)}}, & S > 0 \\ 0, & S = 0 \\ \frac{S+1}{\sqrt{\text{Var}(S)}}, & S < 0 \end{cases}$$

130 The null hypothesis of no trend was rejected if the absolute value of Z exceeded the critical
131 value at the 5% significance level (i.e., $|Z| > 1.96$).

132 Sen's slope estimator was used to quantify the magnitude of the trend. For each pair of time-
133 ordered observations, the slope (Q_i) was calculated as:

134
$$Q_i = \frac{x_j - x_i}{j - i}, \text{ for all } 1 \leq i < j \leq n$$

135 Where x_i and x_j are data values at time points i and j respectively. The Sen's slope was then
136 determined as the median of all Q_i values:

137
$$\text{Sen's slope} = Q_{n/2} \text{ (if } n \text{ is odd)}$$

138
$$\text{Sen's slope} = \frac{Q_{n/2} + Q_{n/2+1}}{2} \text{ (if } n \text{ is even)}$$

139 This approach provides a robust and unbiased estimate of the linear trend over time, even in
140 the presence of outliers or non-normal data distributions (Mann, 1945; Sen, 1968).

141 2.3.3 Statistical software

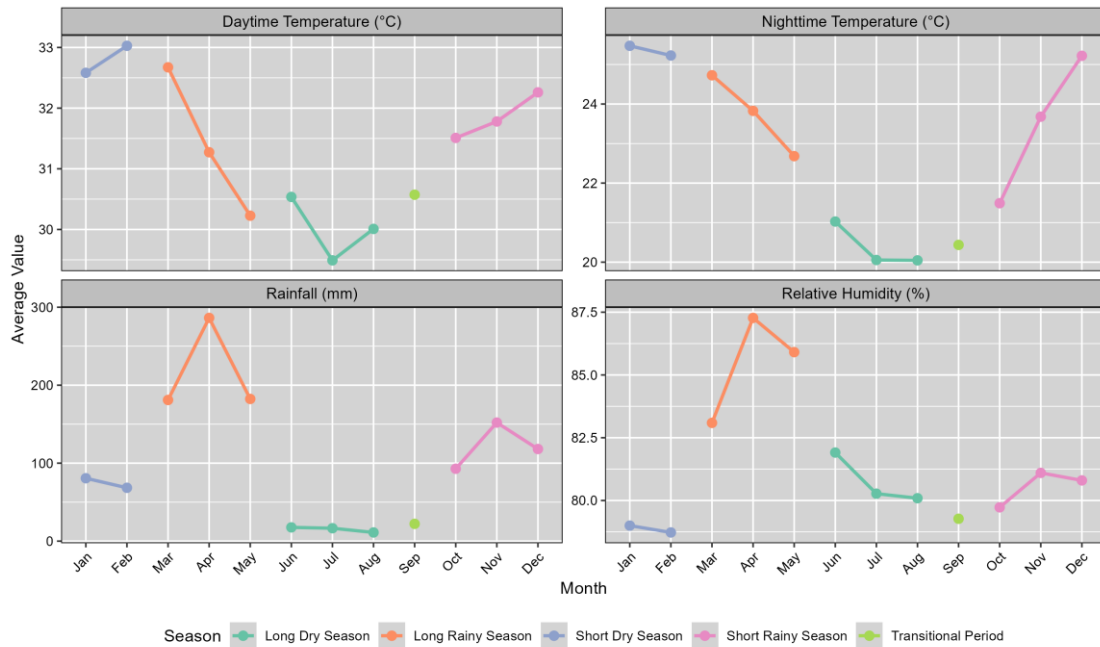
142 All analyses were performed in R version 4.5.1 (R Core Team, 2025). The Mann–Kendall test
143 was conducted using the MannKendall() function from the Kendall package, and Sen's slope
144 was estimated using the sens.slope() function from the trend package, which returns the median
145 of all pairwise slopes.

146 3. Results

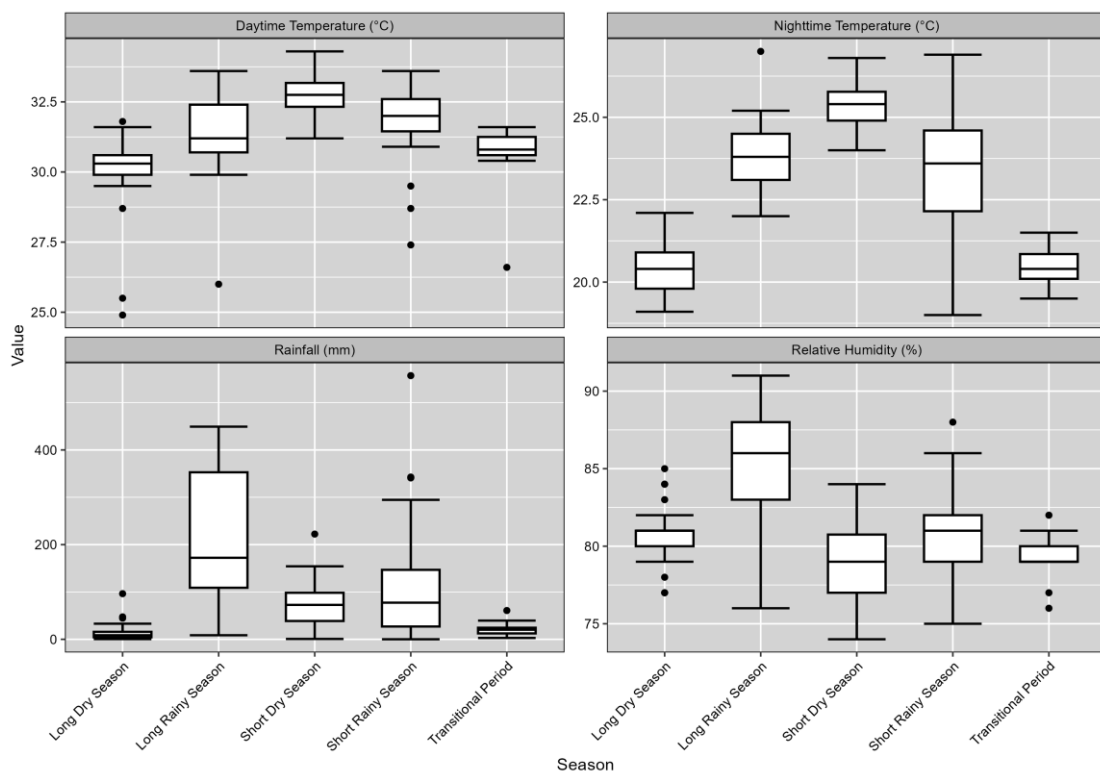
147 3.1 Description of the climatic variables

148 Monthly variations in climatic variables revealed distinct seasonal patterns and interannual
149 dynamics over the decade. Warmer months consistently occurred from January to March and
150 November to December, while cooler months with lower nighttime temperatures were
151 observed from June to September. Rainfall exhibited significant fluctuations, with higher

152 amounts predominantly in April and May. Notably, heavy rainfall was recorded in November
 153 2023, deviating from typical patterns, and a consistent rainfall shortage occurred from June to
 154 September. Relative humidity levels correlated with rainfall, being higher in April, May, and
 155 November, and lower from July to September, highlighting the interconnected influence of
 156 rainfall and humidity on the regional climate (See Figure 2-3).

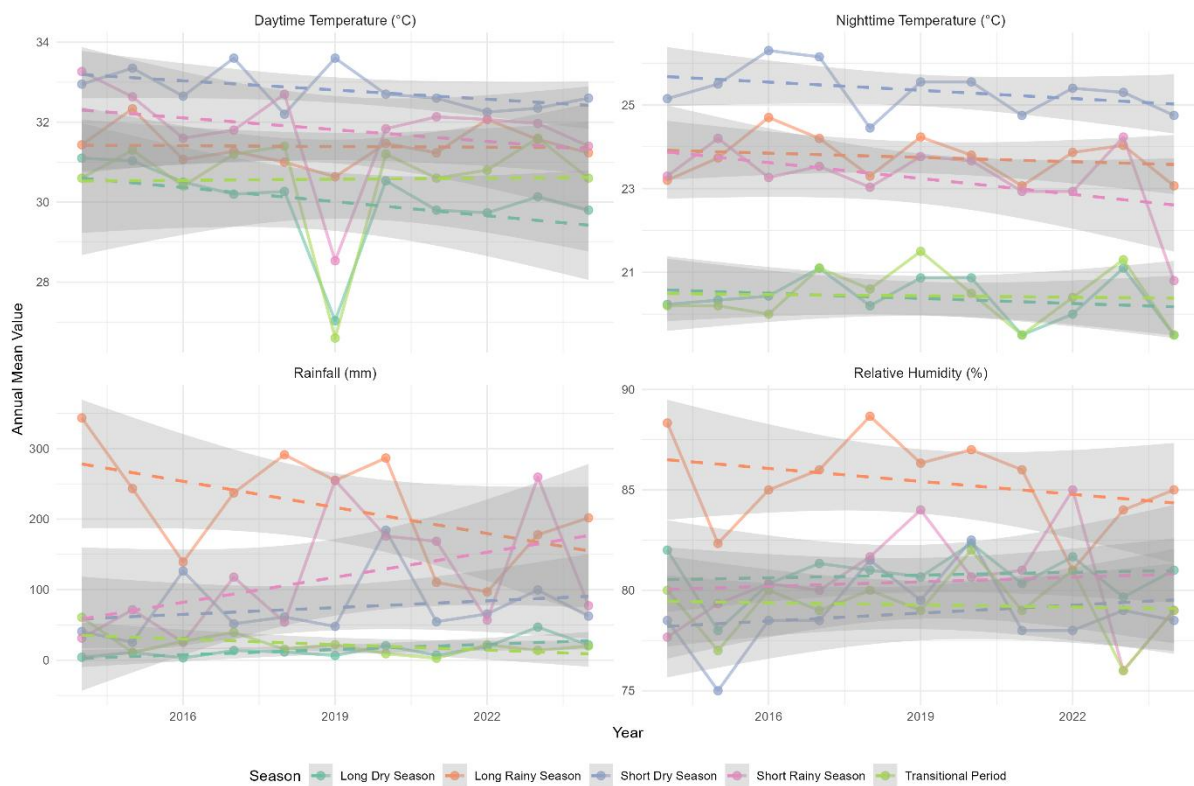


157
 158 *Figure 2: Monthly Climatology of Daytime Temperature, Nighttime Temperature, Rainfall, and*
 159 *Relative Humidity in Dar es Salaam (2014–2024)*



161 *Figure 3. Seasonal Distributions of Climatic Variables in Dar es Salaam (2014–2024) Based*
 162 *on Boxplots*

163 To complement the monthly and seasonal visualizations, annual mean values for each climatic
 164 variable were plotted across all five seasons to examine overarching directional tendencies over
 165 the decade. As shown in the figure below, the linear trend fits highlight broad shifts in
 166 temperature, rainfall, and relative humidity between 2014 and 2024, with notable differences
 167 across seasons. These descriptive patterns provide context for the formal monotonic trend tests
 168 presented in the following subsection, allowing a clearer interpretation of emerging climate
 169 signals in Dar es Salaam (See Figure 4).



170
 171 *Figure 4. Annual Trends in Climatic Variables by Season (2014–2024). Dashed lines represent*
 172 *linear trend fits with 95% confidence intervals*

173
 174 *3.2 Decadal Trend Analysis with Mann Kendall test and Sen’s slope estimator*

175 Seasonal trends in rainfall, daytime and nighttime temperatures, and relative humidity across
 176 Dar es Salaam. A statistically significant increasing trend in rainfall was observed during the
 177 Long Dry season, a period traditionally characterized by minimal precipitation. The Mann-
 178 Kendall test yielded a z-score of 2.8362 with a corresponding p-value of 0.00457, indicating a

179 robust trend. The Sen’s slope estimator quantified this increase at +0.39097 mm/year,
 180 suggesting a quite small year-to-year rise in rainfall during this typically dry period.

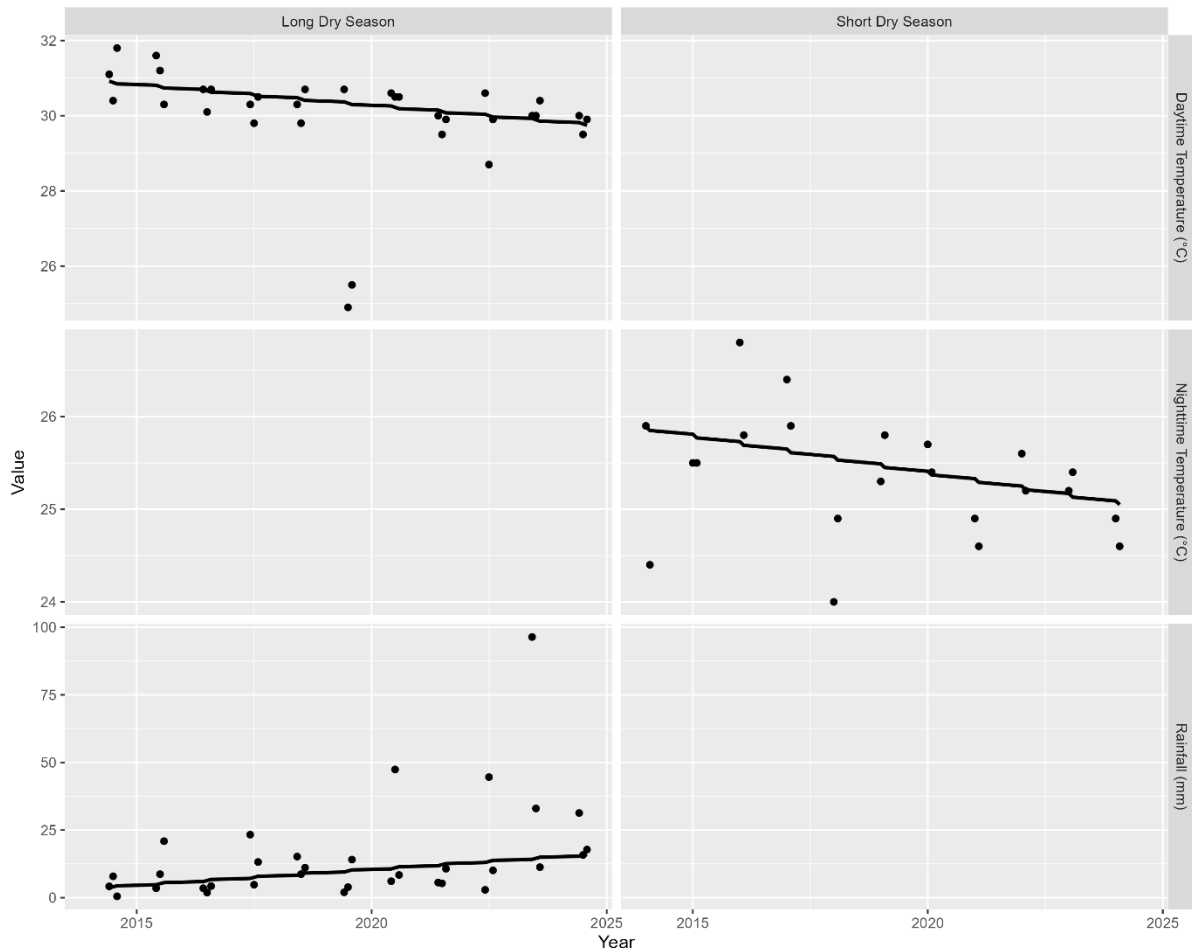
181 Conversely, a statistically significant decreasing trend in daytime temperatures was identified
 182 during the same Long Dry season. The Mann-Kendall test resulted in a z-statistic of -3.2354
 183 and a p-value of 0.00121, signifying a strong downward trend. This was further corroborated
 184 by the Sen’s slope of $-0.03675^{\circ}\text{C}/\text{year}$, suggesting a gradual cooling in daytime temperatures
 185 during this period. In contrast, nighttime temperatures exhibited a significant decrease during
 186 the Short Dry season, with a z-statistic of -2.0098 and a p-value of 0.04445. The Sen’s slope
 187 for this trend was $-0.04^{\circ}\text{C}/\text{year}$, revealing a small but statistically significant decline in
 188 nighttime temperatures during these months.

189 Relative humidity did not show any statistically significant trends across all seasons, indicating
 190 relative stability in atmospheric moisture content over the analyzed period. While minor
 191 fluctuations were observed, none met the threshold for statistical significance (See Table 1)

192 *Table 2: Mann-Kendall Test for Seasonal Climatic Trends (January 2014–October 2024)*

Variable	Season	z	p-value	Sen's Slope
<i>Rainfall (mm)</i>	<i>JF (Short Dry Season)</i>	0.7895	0.42980	1.83333
	<i>MAM (Long Rainy Season)</i>	-1.1931	0.23284	-3.44500
	<i>JJA (Long Dry Season)</i>	2.8362	0.00457	0.39097
	<i>S (Transitional Period)</i>	-1.4013	0.16112	-1.84286
	<i>OND (Short Rainy Season)</i>	1.8696	0.06154	3.04706
<i>Daytime temperature (°C)</i>	<i>JF (Short Dry Season)</i>	-1.7805	0.07500	-0.03750
	<i>MAM (Long Rainy Season)</i>	-0.4035	0.68654	-0.01026
	<i>JJA (Long Dry Season)</i>	-3.2354	0.00121	-0.03675
	<i>S (Transitional Period)</i>	0.1579	0.87450	0.00000
	<i>OND (Short Rainy Season)</i>	-1.4131	0.15763	-0.02857
<i>Nighttime temperature (°C)</i>	<i>JF (Short Dry Season)</i>	-2.0098	0.04445	-0.04000
	<i>MAM (Long Rainy Season)</i>	-0.8228	0.41063	-0.01521
	<i>JJA (Long Dry Season)</i>	-0.9015	0.36735	-0.01667
	<i>S (Transitional Period)</i>	0.0000	1.00000	0.00000
	<i>OND (Short Rainy Season)</i>	-0.0170	0.98640	0.00000
<i>Relative Humidity (%)</i>	<i>JF (Short Dry Season)</i>	0.6840	0.49395	0.06250
	<i>MAM (Long Rainy Season)</i>	-1.1235	0.26123	-0.06250
	<i>JJA (Long Dry Season)</i>	0.4691	0.63901	0.00000
	<i>S (Transitional Period)</i>	-0.2428	0.80816	0.00000
	<i>OND (Short Rainy Season)</i>	1.4445	0.14860	0.09091

194 The significant seasonal trends identified by the Mann–Kendall test were further visualized to
 195 highlight the direction and magnitude of these changes over time. The plotted panels show the
 196 observed values along with their corresponding Sen’s slope lines, clearly illustrating the three
 197 variables that exhibited statistically significant monotonic trends. These visual displays
 198 reinforce the statistical results by providing an intuitive depiction of how these climatic
 199 variables have evolved across the decade, revealing subtle but meaningful shifts in both
 200 precipitation and temperature patterns in Dar es Salaam (See Figure 5).



201

202 *Figure 5. Panel plots of significant climatic trends identified by the Mann–Kendall test. Each*
 203 *panel displays observed seasonal values (points) with Sen’s slope (solid line) for the Long Dry*
 204 *Season (JJA) and Short Dry Season (JF). Only variables with statistically significant*
 205 *monotonic trends ($p < 0.05$) are shown: rainfall (JJA), daytime temperature (JJA), and*
 206 *nighttime temperature (JF).*

207 4. Discussions

208 This study identified three statistically significant monotonic trends in Dar es Salaam’s
 209 seasonal climate over the 2014–2024 period: (i) increasing rainfall during the Long Dry Season

210 (JJA), (ii) decreasing daytime temperatures during JJA, and (iii) declining nighttime
211 temperatures during the Short Dry Season (JF). Although the detected trends are modest in
212 absolute magnitude, their statistical significance is noteworthy because consistent directional
213 changes, however small, accumulate over time and may signal early-stage climatic shifts that
214 warrant monitoring.

215 The detected increase in rainfall during the long dry season suggests a shift toward more
216 irregular precipitation patterns. Similar anomalies have been reported in other parts of
217 Tanzania, where unusual dry-season rainfall has been associated with both land-use change
218 and broader climate change drivers (Kashaigili et al., 2013). Large-scale climate modes such
219 as the Indian Ocean Dipole (IOD) and the El Niño–Southern Oscillation (ENSO) increasingly
220 influence rainfall distribution across East Africa (Nicholson, 2017). Positive IOD phases,
221 characterized by warmer western Indian Ocean waters, are known to enhance rainfall even
222 during periods historically associated with suppressed convection (Chobo & Huo, 2024), and
223 climate simulations indicate that such events may become more frequent under warming
224 scenarios (Cai et al., 2014). Regional analyses further show that East Africa has experienced
225 increasingly erratic rainfall patterns due to ocean warming (Nicholson, 2017). Consistent with
226 our findings, previous research documented wetter short rains and a long-term drying of the
227 long rains, with partial recovery in recent decades (Palmer et al., 2023). These shifts have
228 implications for flooding, drought cycles, energy systems, agriculture, and vector- and
229 waterborne disease dynamics, amplifying existing vulnerabilities.

230 Comparable observations from South Sudan also highlight increasing rainfall during typically
231 dry periods, disrupting agricultural systems and heightening exposure to climate extremes
232 (Zakaria Lukwasa et al., 2022). Studies in coastal Kenya similarly report rising dry-season
233 rainfall associated with intensified moisture convergence (Ngunvava & Abiodun, 2023). Local
234 influences, including rapid urbanization and the replacement of vegetated surfaces with
235 impervious materials, may also alter convective processes and enhance localized precipitation
236 (Kimutai et al., 2023).

237 While increased dry-season rainfall may ease water scarcity, it simultaneously raises concerns
238 about flood risk, particularly in rapidly urbanizing areas with limited drainage capacity. Much
239 of Dar es Salaam’s drainage infrastructure was designed for historical rainfall patterns rather
240 than the heavier off-season precipitation now emerging (UN-Habitat, 2022). When combined
241 with urban growth, low solid waste collection coverage, and climate change, these changes
242 exacerbate flood vulnerability (Sakijege & Dakyaga, 2023). Unusual rainfall patterns also
243 affect sanitation systems in informal settlements. At the same time, increased off-season

244 rainfall may support expanded water-harvesting initiatives if adequate storage systems are
245 integrated into planning (Raimondi et al., 2023). These findings emphasize the need for
246 planners to incorporate updated rainfall baselines into urban design, particularly regarding
247 drainage and water management.

248 Temperature trends also revealed noteworthy patterns. Daytime temperatures during JJA
249 declined significantly, and nighttime temperatures during JF showed a measurable decrease.
250 Although global temperatures continue to rise, localized cooling in coastal or highly vegetated
251 areas has been linked to sea-breeze penetration, urban greening, and microclimatic regulation
252 (Esperon-Rodriguez et al., 2025). Coastal cities such as Dar es Salaam benefit from maritime
253 influences, which moderate heat extremes through enhanced ventilation. Strengthened wind
254 patterns or increased sea-breeze frequency over the study period may contribute to these
255 cooling tendencies. Similar coastal moderation effects have been documented in Xiamen,
256 where the ocean exerts cooling influence up to nearly 8 km inland (Hong et al., 2025), and in
257 Lake Ontario–adjacent urban areas, where local temperatures can be reduced by as much as
258 3°C (Jandaghian & Colombo, 2024). Vancouver’s coastal features, including Burrard Inlet and
259 False Creek, also play a key role in lowering local temperatures by approximately 2–4°C
260 (Spronken-Smith & Oke, 1998).

261 Atmospheric aerosols may further contribute to local cooling. Increased particulate emissions
262 from traffic, industry, and biomass burning can enhance atmospheric albedo, reflecting solar
263 radiation and reducing surface heat (Onaiwu & Ayidu, 2025). Aerosols acting as cloud
264 condensation nuclei generate clouds with smaller droplets, increasing reflectivity and thereby
265 strengthening cooling effects (Akinyoola et al., 2024). Their ability to prolong cloud cover by
266 suppressing precipitation can enhance this cooling further (Irfan et al., 2024). These
267 mechanisms may help explain the localized temperature decreases observed in Dar es Salaam.
268 Nighttime cooling during JF may also be linked to reduced nocturnal cloud cover, which
269 increases outgoing longwave radiation. Previous studies in East Africa have documented
270 similar mechanisms contributing to nighttime cooling in coastal regions (Cai et al., 2014;
271 Chobo & Huo, 2024; Nicholson, 2017; Palmer et al., 2023). However, cooling trends should
272 be interpreted with caution. Localized temperature declines do not contradict broader global
273 warming trajectories. Instead, they underscore the complexity of urban microclimates and the
274 need for fine-scale spatial analyses to distinguish urban-core, peri-urban, and coastal
275 influences.

276 Relative humidity exhibited no significant trend across any season, consistent with findings
277 from other coastal environments where proximity to large water bodies stabilizes atmospheric

278 moisture (Jandaghian & Colombo, 2024). The Indian Ocean likely provides a buffering effect
279 for Dar es Salaam, maintaining moisture availability despite fluctuations in temperature or
280 rainfall. This marine influence may shield the city from humidity volatility commonly observed
281 in inland regions (Camberlin et al., 2017). Nevertheless, even modest fluctuations can
282 substantially affect human thermal comfort, as humidity intensifies heat stress under elevated
283 temperatures (Wang & Gaffen, 2001). Although humidity remained stable during the study
284 period, projected intensification of the global hydrological cycle could increase the frequency
285 of extreme humidity episodes (IPCC, 2023), making continued monitoring essential.

286 Overall, the study underscores the importance of localized, seasonally resolved climate
287 assessments for coastal urban centers. The emerging shifts documented here, particularly
288 unseasonal rainfall increases and localized cooling, carry direct implications for urban
289 planning, disaster risk reduction, public health readiness, and water resource management.
290 Strengthening observational networks, integrating remote sensing, and developing fine-scale
291 climate analyses will be critical for improving urban climate resilience in Dar es Salaam.

292

293 **5. Conclusion**

294 In coastal areas like Dar es Salaam that are fast becoming more urbanized, it is essential to
295 comprehend localized climate changes. Urban flooding, water supply, thermal comfort, and
296 public health are all impacted by climate variability. Global and regional patterns have received
297 a lot of attention, but city-specific, seasonal-scale evaluations are still rare despite being crucial
298 for climate-resilient urban development.

299 The climate of Dar es Salaam seems to be changing seasonally in subtle but noticeable ways.
300 Contrary to predictions, higher rainfall during the dry season and falling temperatures challenge
301 traditional urban calendars, water management strategies, and disaster risk reduction
302 frameworks. In order to effectively prepare for adaptation and future-proof public health
303 systems, urban infrastructure, and economic activity in the face of climate unpredictability,
304 localized and seasonal evaluations are essential. As explained above, this local-level seasonal
305 study offers vital information that can help close the gap between the realities of urban planning
306 and national climate policies. Therefore, for Dar es Salaam to become more resilient, its city
307 master plan has to undergo regular revision and incorporate updated climatic baselines.

308 **6. Limitations of the study**

309 The study has some limitations. It relied on data from a single meteorological station and a
310 relatively short temporal window (2014–2024), which may not fully capture microclimatic

311 variations across the broader Dar es Salaam area, as there is only one local station. Additionally,
312 the study did not investigate potential underlying drivers of the observed trends, such as the
313 effects of urbanization or larger regional climate dynamics. To further understand urban
314 microclimatic variability, future studies should combine lengthier datasets, remote sensing
315 data, and localized ground-truthing. Finally, while statistically significant, the magnitude of
316 detected trends was relatively small. This reflects both the short duration of the dataset and the
317 inherent subtlety of decadal climate shifts. As a result, the observed trends should be interpreted
318 as indicators of directional change rather than large, immediate impacts. Longer-term datasets
319 would allow more precise estimates of climatic significance.

320 **Declaration of generative AI and AI-assisted technologies in the writing process**

321 During the preparation of this work the authors used Grammarly in order to improve readability
322 and clarity of the paper. After using this tool/service, the authors reviewed and edited the
323 content as needed and take full responsibility for the content of the publication.

324 **Ethics statement**

325 This study did not involve human participants or the use of personal data. All data used were
326 publicly available meteorological records obtained from the Tanzania Meteorological
327 Authority. Therefore, ethical approval was not required.

328 **Author Contributions**

329 **IM:** Conceptualization, Data curation, Investigation, Methodology, Resources, Validation,
330 Writing - original draft, Writing - review & editing. **HM:** Conceptualization, Methodology,
331 Project administration, Validation, Writing - review & editing. **IHR:** Conceptualization, Data
332 curation, Methodology, Validation, Writing - review & editing. **OP:** Conceptualization, Data
333 Curation, Methodology, Validation, Writing - review & editing. **JGM:** Conceptualization,
334 Formal analysis, Methodology, Software, Validation, Visualization, Writing - original draft,
335 Writing - review & editing. All authors have read and approved the final manuscript and agree
336 to be accountable for all aspects of the work.

337 **R Codes and Data Availability Statement**

338 The dataset used in this study has been made publicly available through [DOI: 10.5281/zenodo.17720494](https://doi.org/10.5281/zenodo.17720494). All relevant materials are available to readers without restriction.
339 [10.5281/zenodo.17720494](https://doi.org/10.5281/zenodo.17720494). All relevant materials are available to readers without restriction.
340 For further inquiries, please contact the corresponding author.

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343 **Declaration of competing interests**

344 The authors declare no conflicting interests related to this study.

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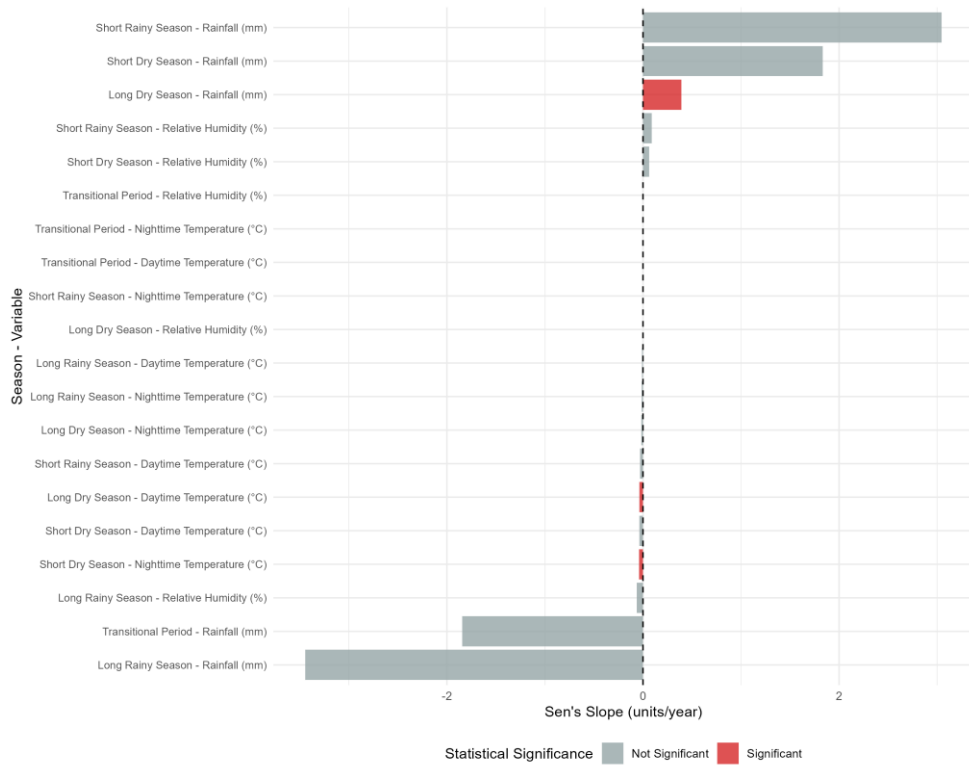
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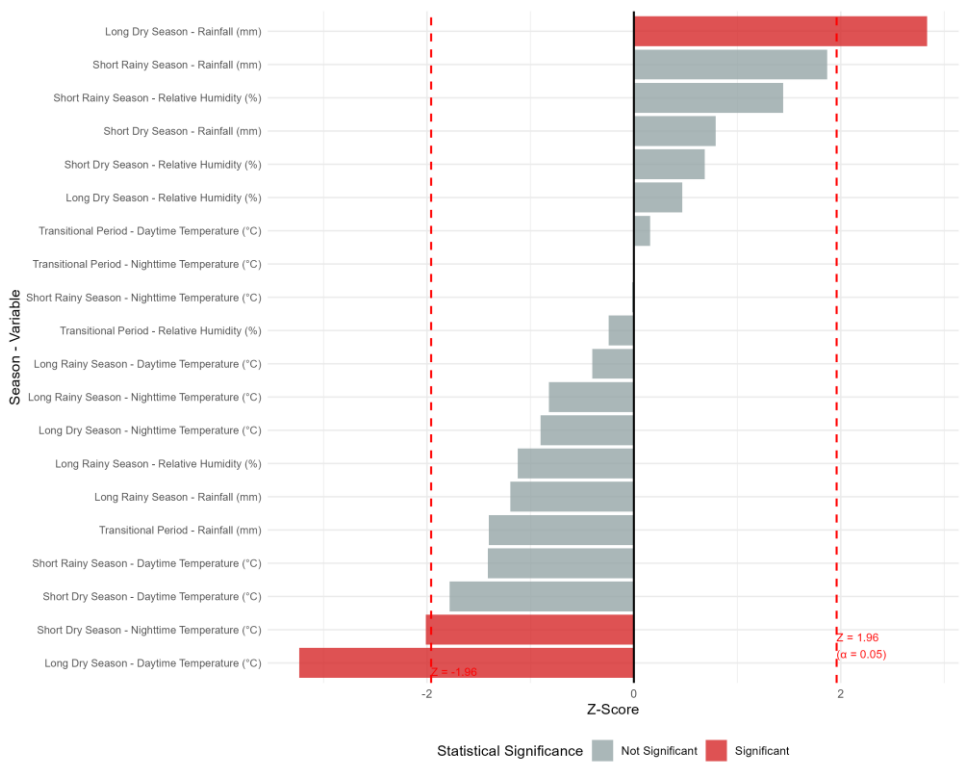
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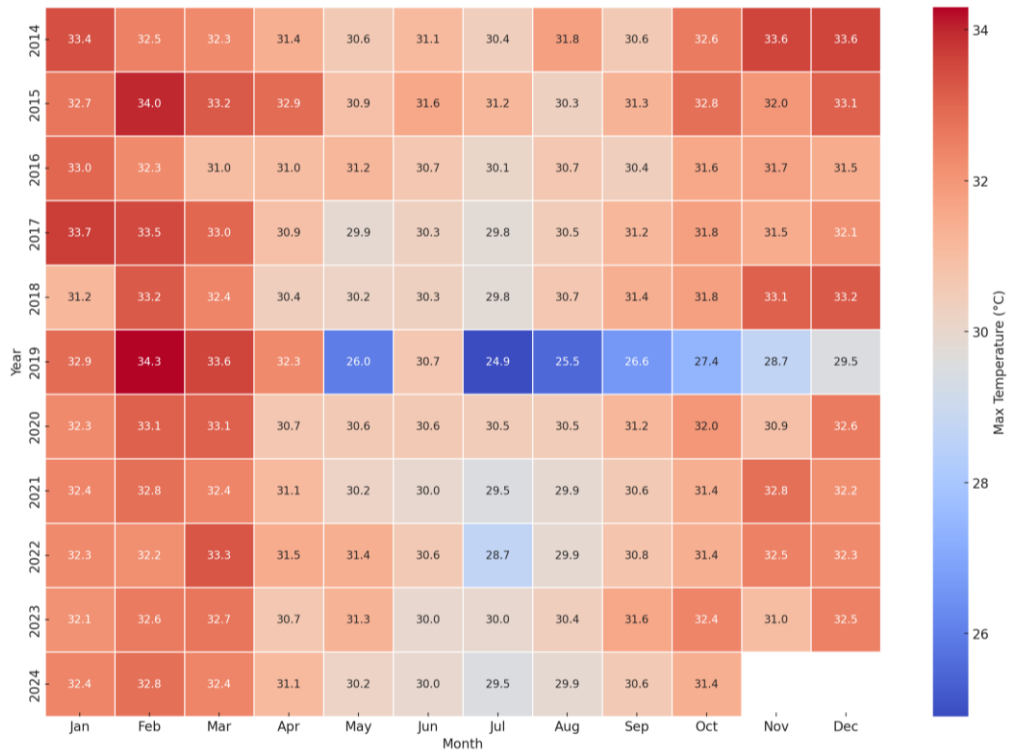
APPENDICES



Appendix I: Sen's Slope Magnitude Visualization



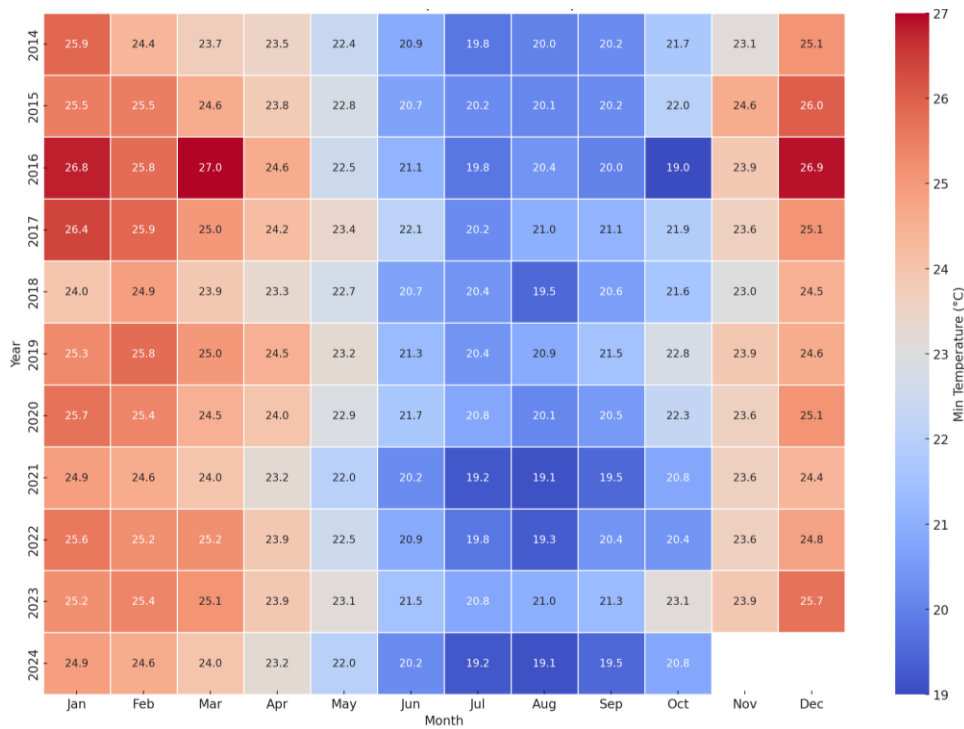
Appendix II: Mann-Kendall Z-Scores by Season and Variable (Horizontal dashed lines indicate critical values (± 1.96) for $\alpha = 0.05$)



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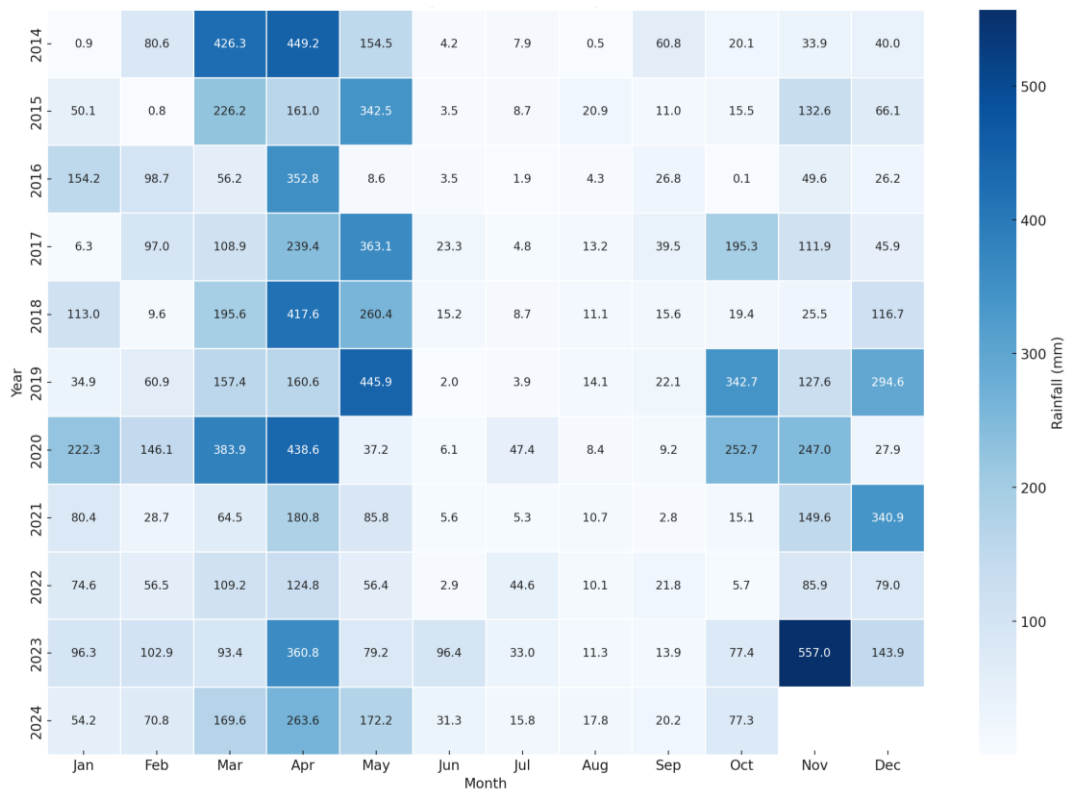
Appendix III: A heatmap of Monthly and Interannual Variations in Daytime Temperature



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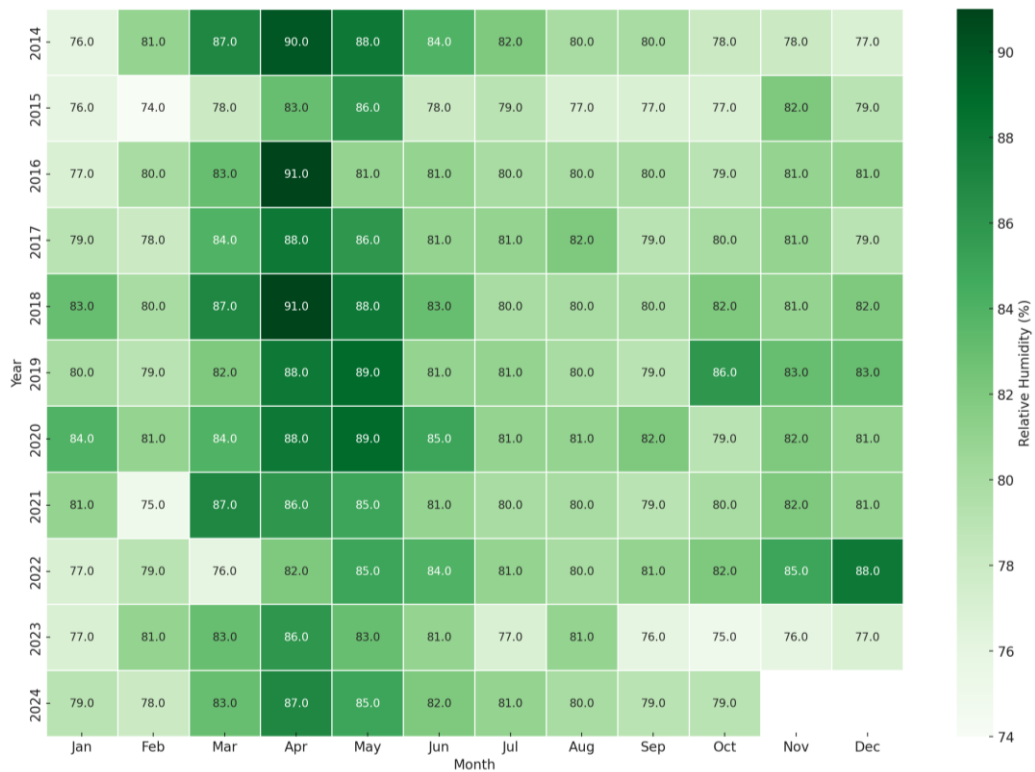
Appendix IV: A heatmap of Monthly and Interannual Variations in Nighttime Temperature



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Appendix V: Heatmap of monthly rainfall in Dar es Salaam from Jan 2014 - Oct 2024



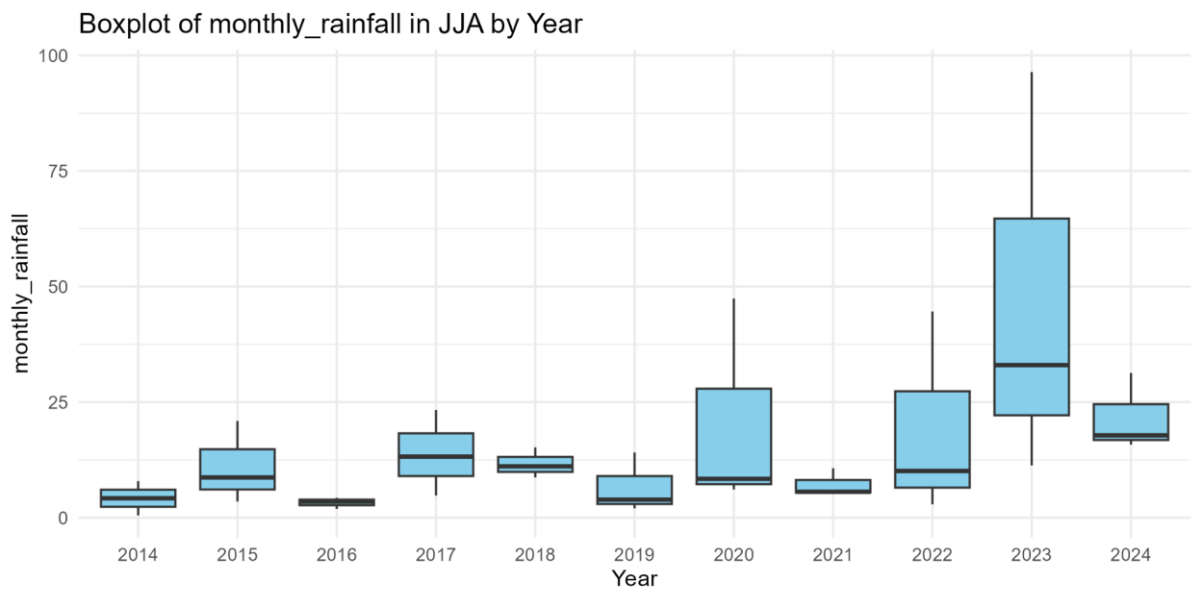
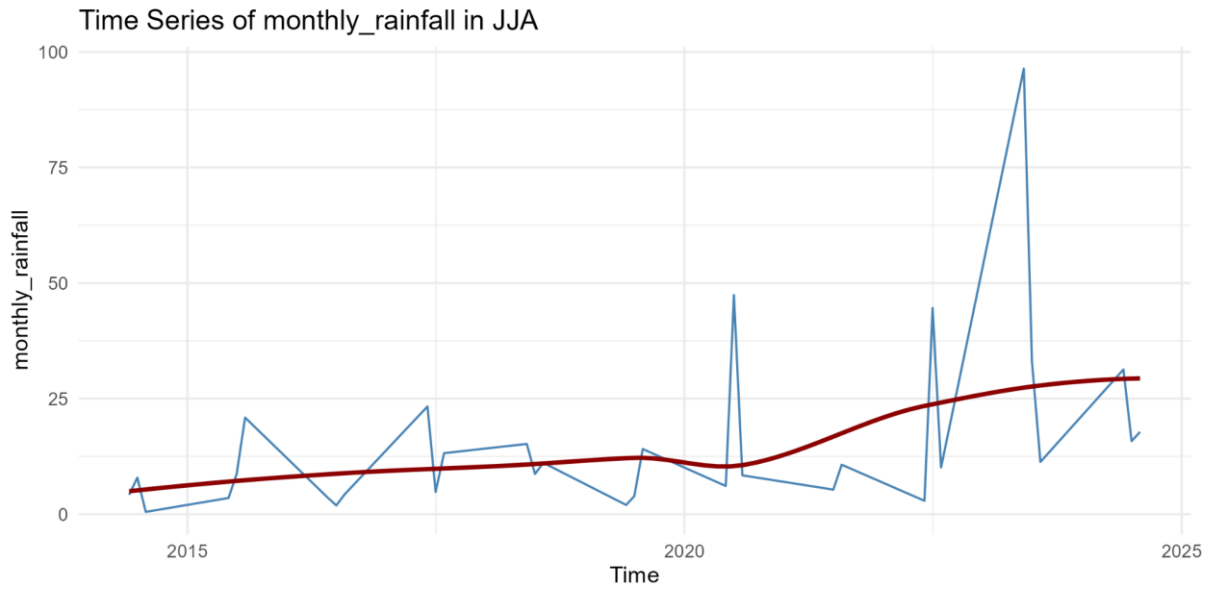
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Appendix VI: Heatmap of Monthly Relative Humidity in Dar es Salaam from Jan 2014 - Oct 2024



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Appendix VII: Time series plots of rainfall in the long dry season.

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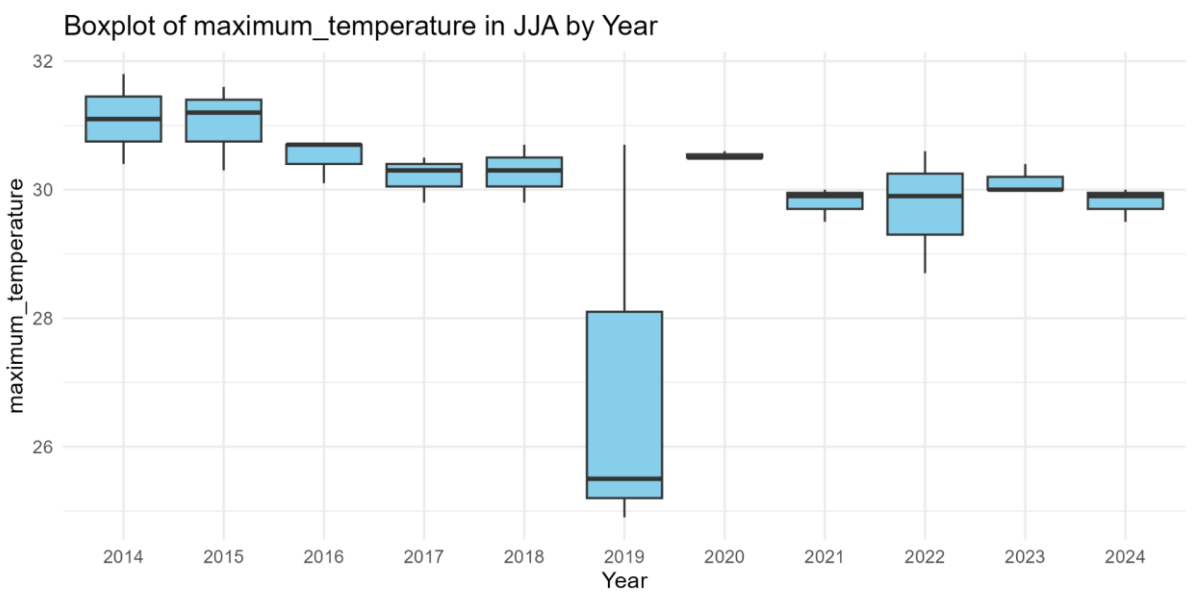
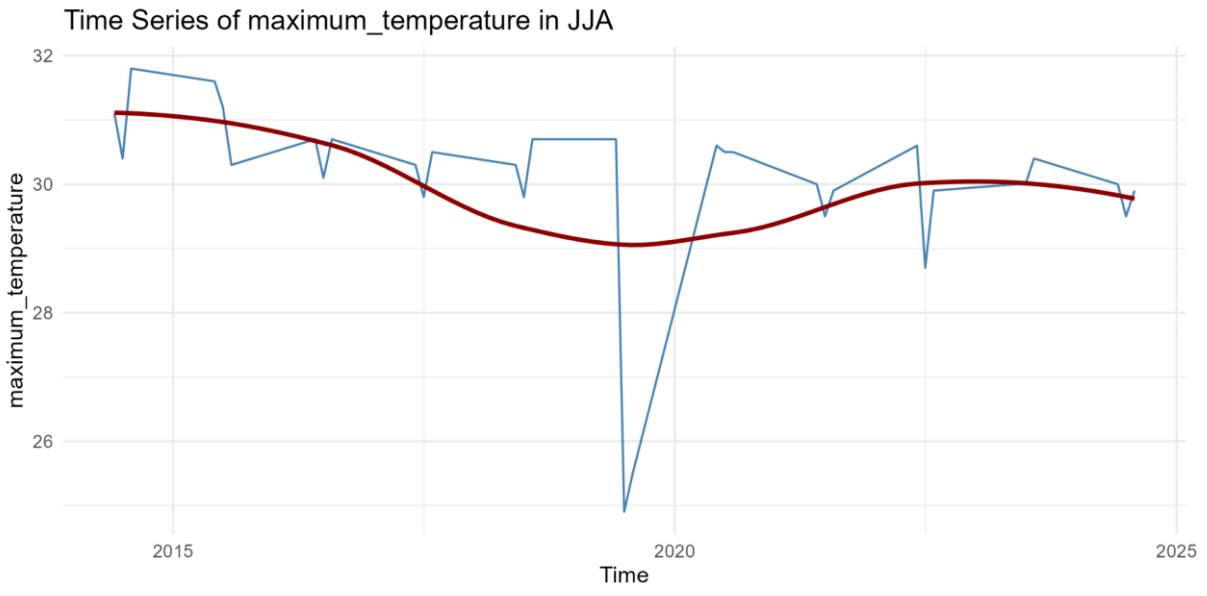
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Appendix VIII: Time series plot of daytime temperature in the long dry season

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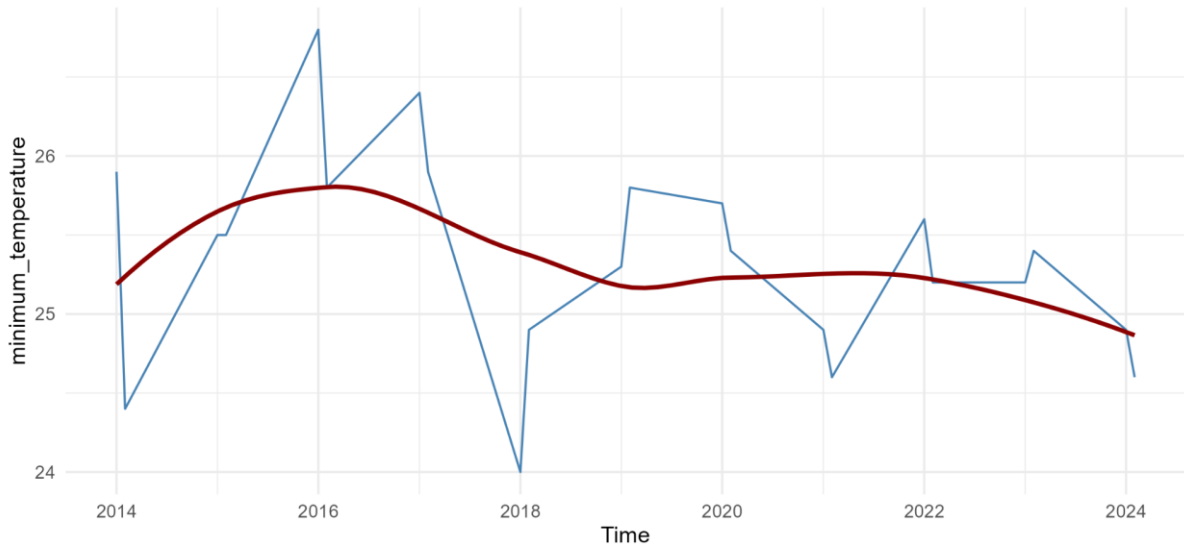
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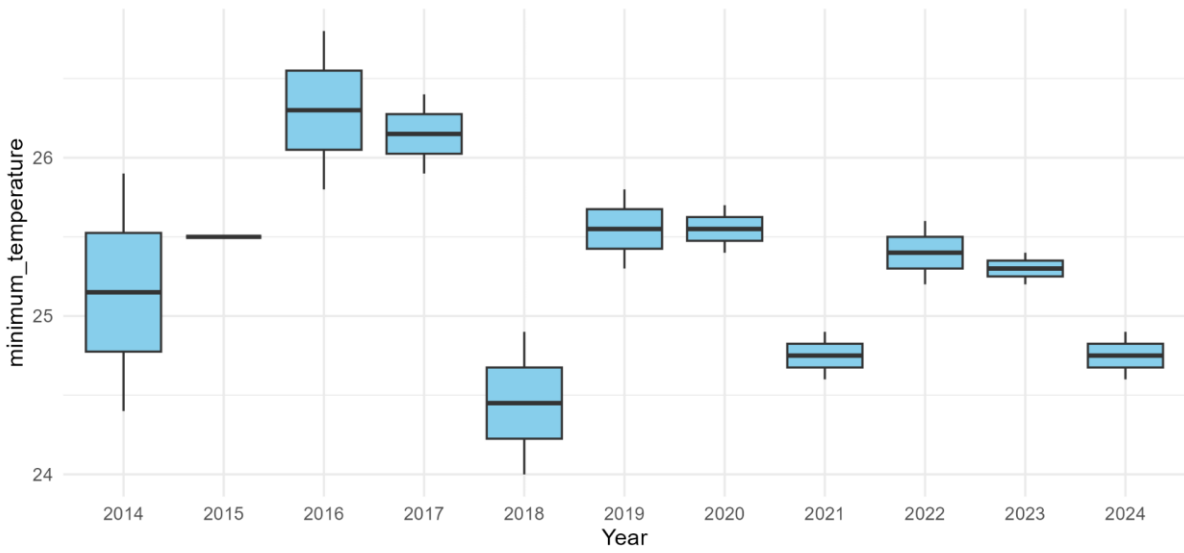
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Time Series of minimum_temperature in JF



Boxplot of minimum_temperature in JF by Year



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Appendix IX: Time series plot of nighttime temperature in the short dry season