General methods for visualization of Hydro-climatic data

Hydro-Climatic data

Hydro-climatic data typically includes various hydrology (water) and climate variables. Here's a list of common variables found in hydro-climatic datasets:

1. Precipitation	10. Relative Humidity
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- 3. Humidity 12. Solar Radiation
- 4. Evaporation 13. Atmospheric Pressure
- 5. Streamflow 14. Cloud Cover
- 6. Snow Water Equivalent (SWE) 15. Potential Evapotranspiration (PET)
- 7. Soil Moisture 16. Water Temperature
- 8. Groundwater Level 17. Soil Temperature
- 9. River Discharge 18. Drought Indices
 - 19. Flood Indices

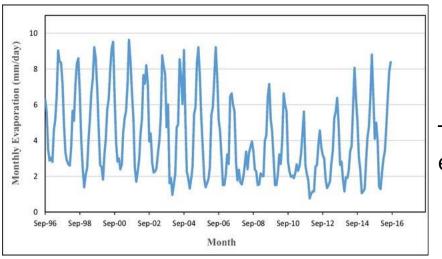
Visualization techniques used for hydro-climatic data

- 1.Time Series Plots
- 2. Scatter Plots
- 3.Heat maps
- 4. Contour Plots

- 5. GIS Maps
- 6. Animated Visualizations
- 7. Box Plots and Violin Plots
- 8. Radial Plots

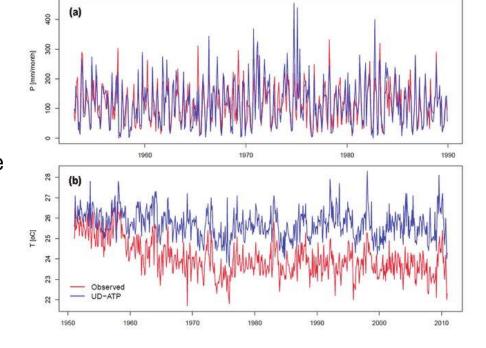
Time Series Plots:

- Time series plots are widely used to visualize the temporal variation of hydro-climatic variables over time.
- The x-axis represents time (e.g., days, months, years), while the y-axis represents the variable of interest (e.g., precipitation, temperature).
- Multiple time series can be plotted on the same graph to compare different variables or locations.

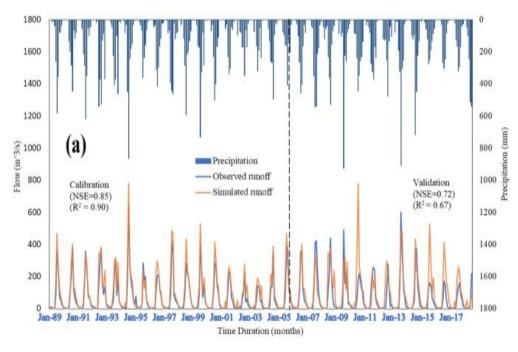


Time series of observed evaporation

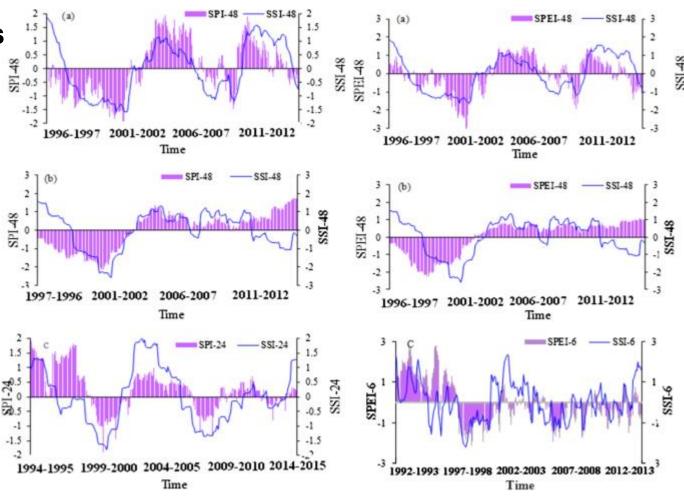
(a) Precipitation and(b) temperature timeseries plots



Time Series Plots for other variables



- Monthly time series plot showing the variation of observed and simulated runoff at the bottom axis and precipitation at the top axis.
- Best example for multiple time series plot where two variables are plotted on the same graph to compare.



Time series of drought indices with high cross-correlation coefficient in the **a** Navroud, **b** Lighvan, **c** Seqez

Source: https://doi.org/10.1007/s13201-020-01345-6

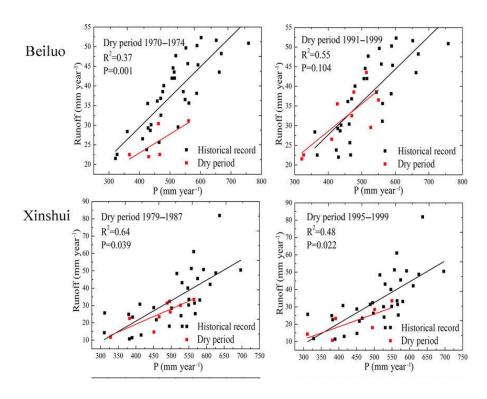
Scatter Plots:

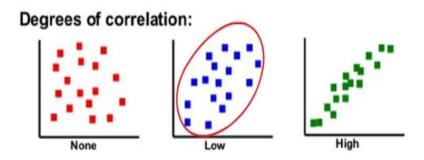
- Scatter plots are useful for exploring relationships between two continuous variables, such as precipitation and stream flow.
- Each data point represents a combination of values for the two variables, with one variable plotted on the x-axis and the other on the y-axis.
- Trend lines or regression lines can be added to identify patterns or correlations in the data.

The annual precipitation-runoff scatter plot for each watershed.

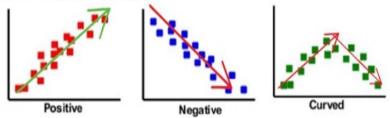
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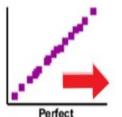
https://doi.org/10.5194/hess-22-1749-2018



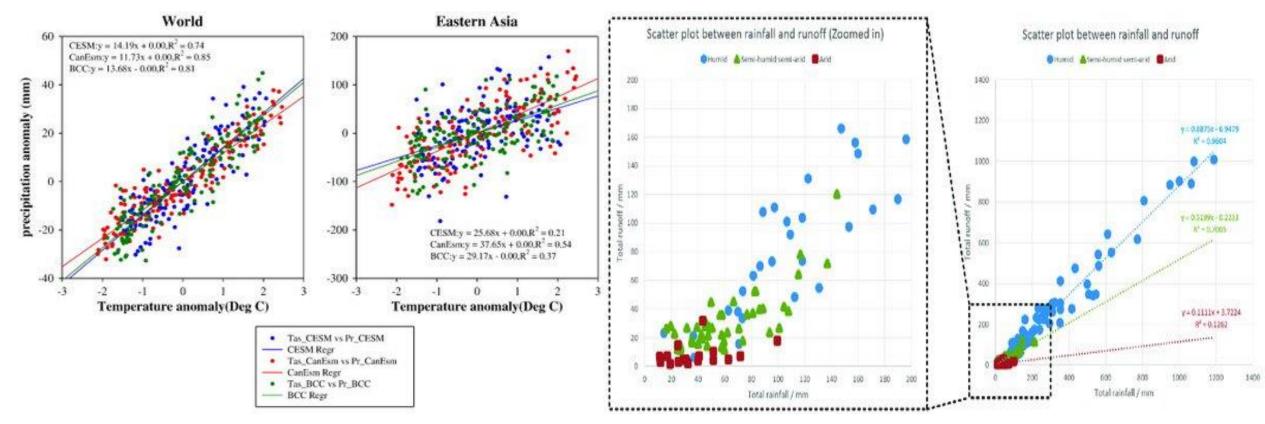


Types of correlation:





More on Scatter Plots



Scatter plots of temperature anomaly and precipitation anomaly over the world and Eastern Asia during the period of 1850–1989. Solid lines indicate linear relationships between temperature anomaly precipitation anomaly.

Source:

Scatter plot between rainfall and runoff.

Source: https://doi.org/10.3390/w9100719

https://doi.org/10.1016/j.gloplacha.2013.11.007

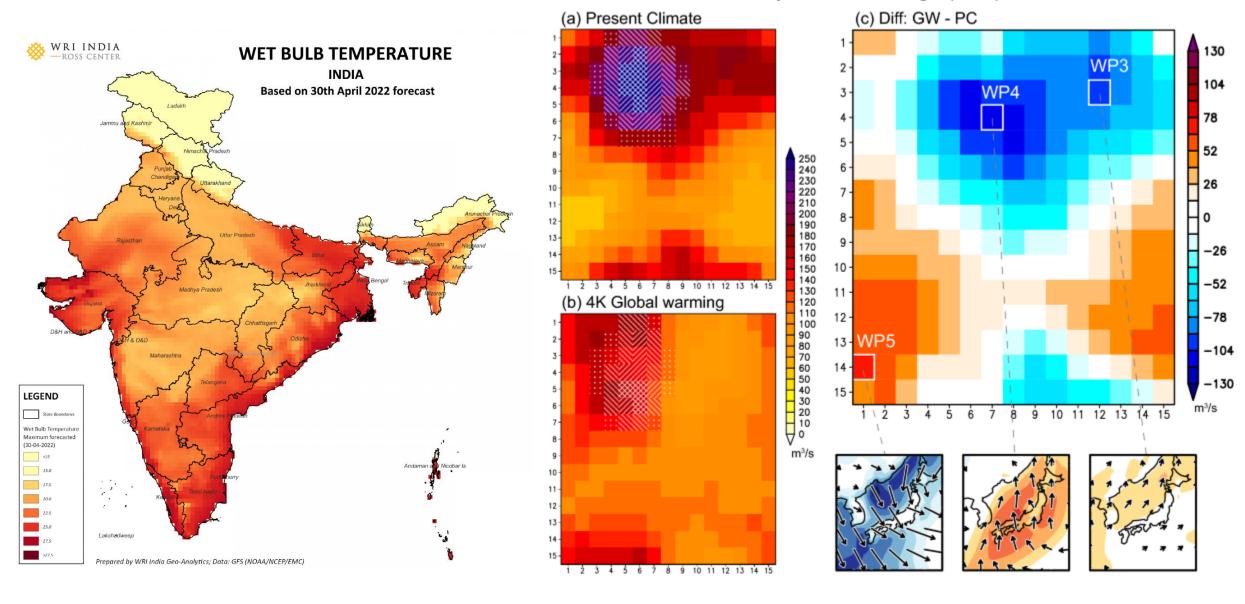
Heatmaps:

- Heatmaps are effective for visualising spatial patterns and distributions of hydro-climatic variables across a geographic area.
- They use colour gradients to represent variations in intensity or magnitude, with warmer colours indicating higher values and cooler colours indicating lower values.
- Heatmaps can be generated for variables like precipitation, temperature, or soil moisture over a specific region or watershed.

				4	567	
January	255	167	123	65	41	
February	244	196	89	52		
March	268	198	119	46	20	
April	321	179	88	31	11	
May	413	152	51	31	4	
June	432	137	42	15	4	
July	567	63	14			
August	550	63	24	8	6	
September	454	114	39	16	7	
October	337	165	81	44	24	
November	238	162	110	65	55	
December	263	176	113	64	35	
< 0.1 0.1 - 4.0 4.1 - 10.0 10.1 - 20.0 > 20.0 Daily precipitation (mm)						

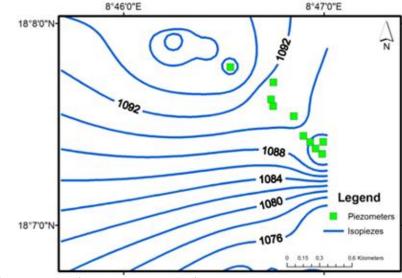
Heatmap for precipitation by month, 1998-2018

Daily-mean discharge (m³/s): MAM

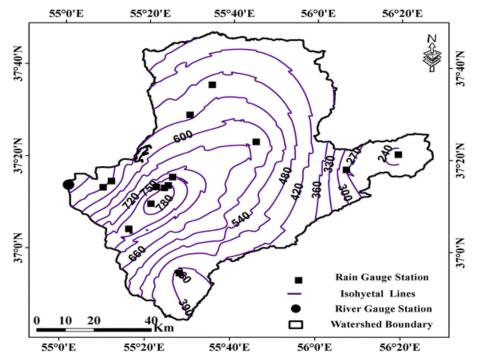


Contour Plots:

- Contour plots are similar to heatmaps but use contour lines to represent isovalues (lines of constant value) of a hydroclimatic variable.
- They provide a visual representation of elevation or gradient changes in the variable across a geographic area.
- Contour plots are commonly used for variables like precipitation, temperature, or groundwater levels.



Piezometric map showing groundwater levels

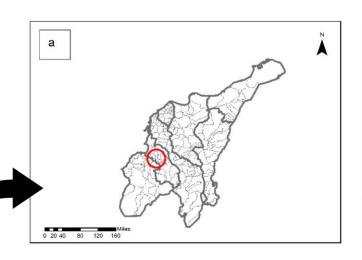


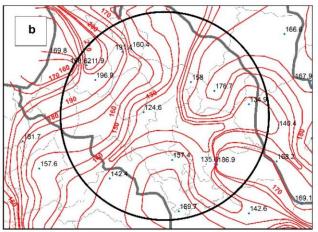
Isohyetal lines depicts the variations in precipitation

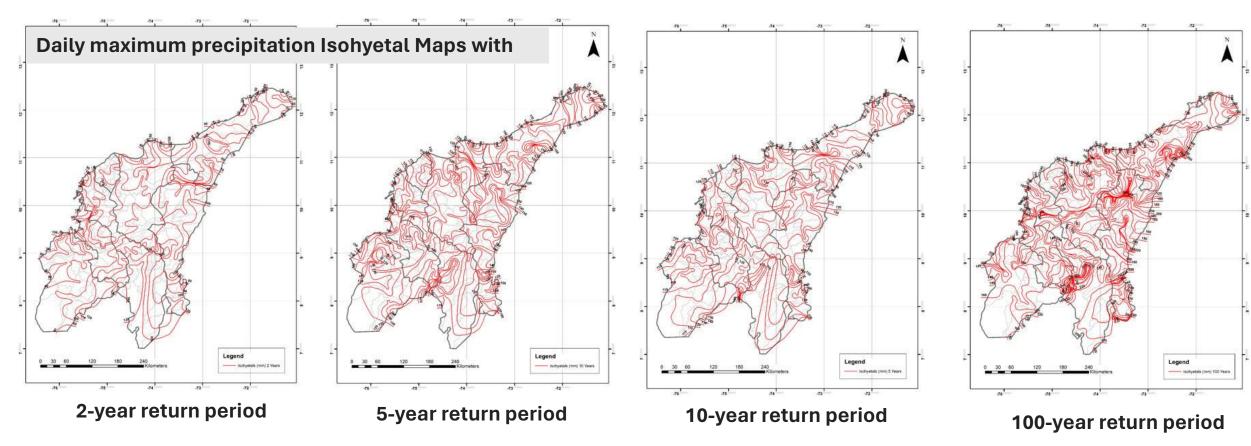
Case Study: Isohyetal Maps of Daily Maximum Rainfall for Different Return Periods for the Colombian Caribbean Region

(Álvarez et.al 2019)

Area showing isohyetal alignment discrepancies (a) isohyetal IDW method (b)







GIS Maps:

- Geographic Information System (GIS)
 maps are powerful tools for visualizing
 hydro-climatic data in a spatial
 context.
- They allow users to overlay different layers of data, such as precipitation, land cover, elevation, and hydrological features, to analyze spatial relationships and patterns.
- GIS maps can be used to create thematic maps, flow maps, or 3D terrain visualizations, depending on the specific application.

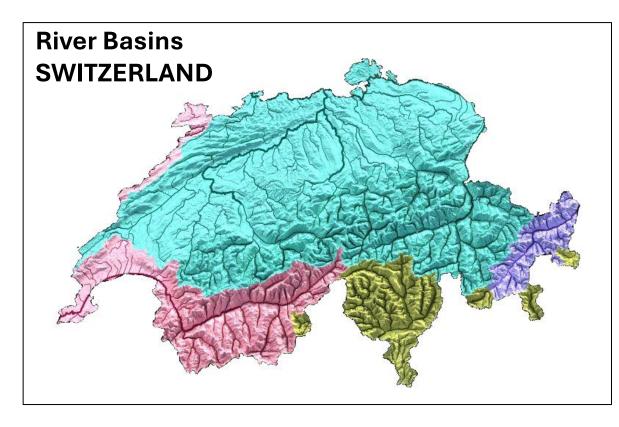


Figure showing the River Basins of Switzerland
Source: ©2024 Milos Popovic

(https:// milospopovic.net)

GIS Maps:

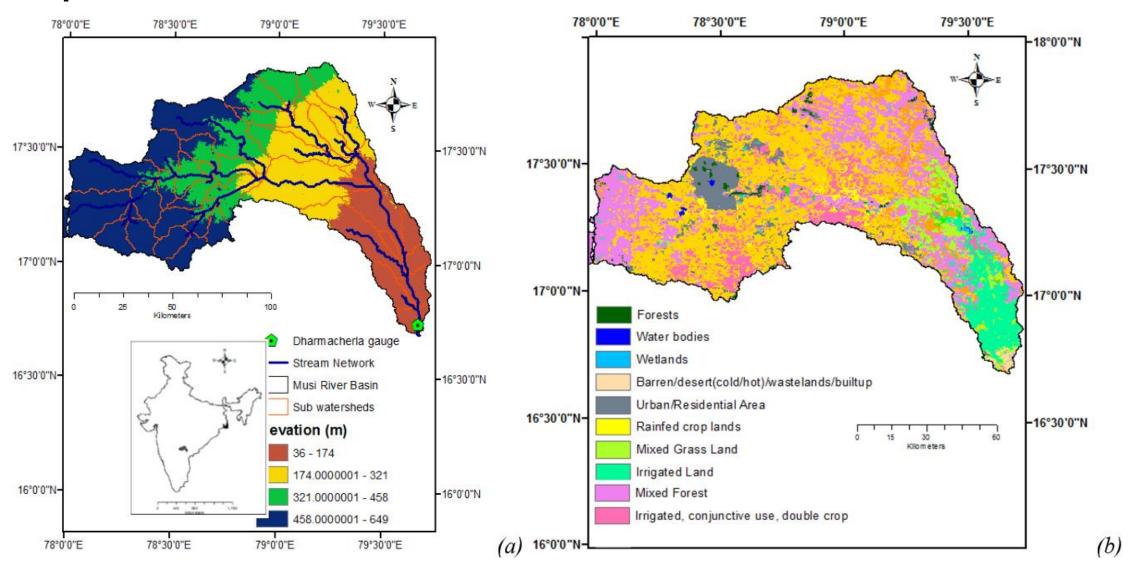
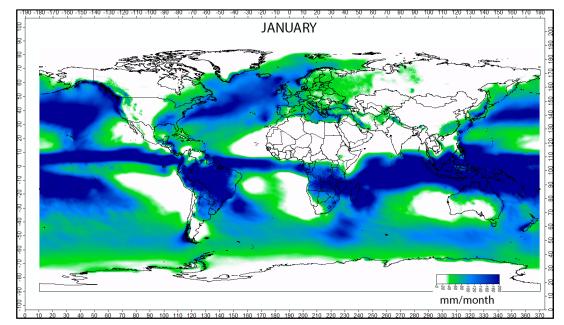


Figure showing the Stream network, sub watersheds, elevation (a) and Land use characteristics (b) of Musi River Basin

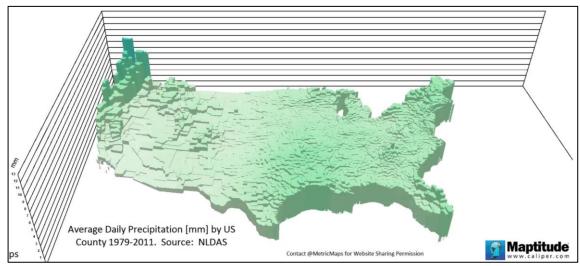
Source: Jothiprakash et al. 2017

Animated Visualizations:

- Animated visualizations are effective for illustrating temporal changes or dynamic processes in hydro-climatic data.
- They can be used to show seasonal variations, trends over time, or the progression of extreme events like storms or droughts.
- Animation techniques include timelapse sequences, animated line graphs, or spatial animations of changing variables over a geographic area.



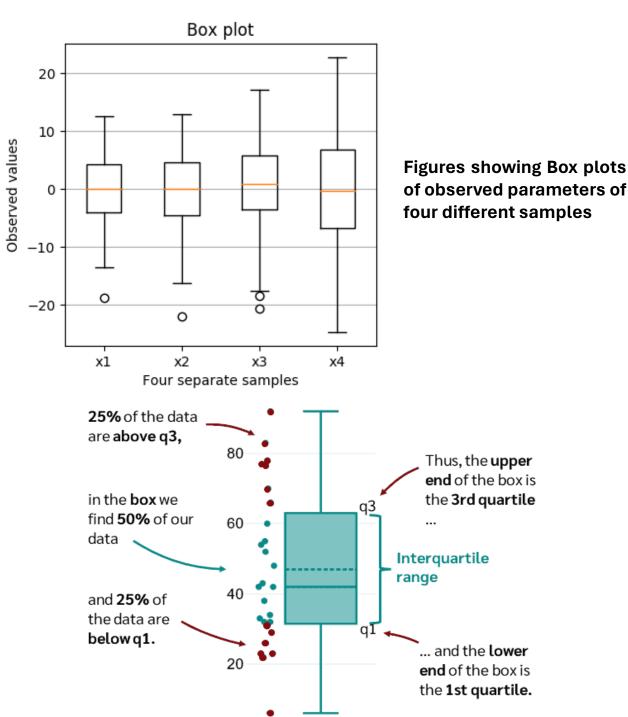
Average Precipitation variation of the year 2016 Source :https://en.m.wikiversity.org/



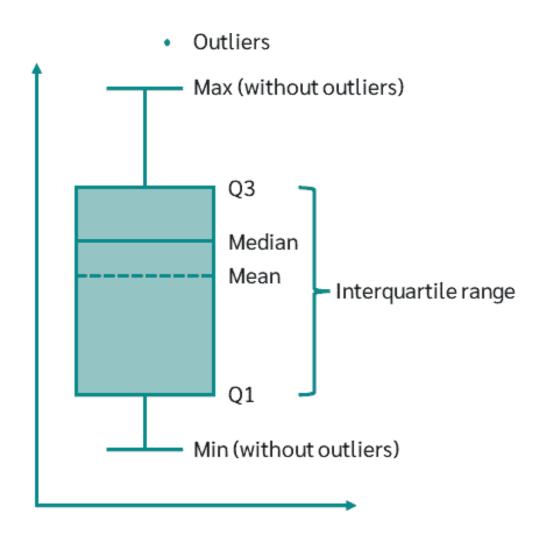
Average Daily precipitation of US during 1979-2011 Source : NLDAS

Box Plots:

- Box plots are useful for visualizing the distribution and variability of hydroclimatic variables.
- They display summary statistics such as median, quartiles, and outliers, providing insights into the central tendency and spread of the data.
- Box plots are particularly useful for comparing the distribution of variables between different groups or categories.



Box Plots:



The box indicates the range in which the middle 50% of all data lies

Thus, the lower end of the box is the 1st quartile and the upper end is the 3rd quartile

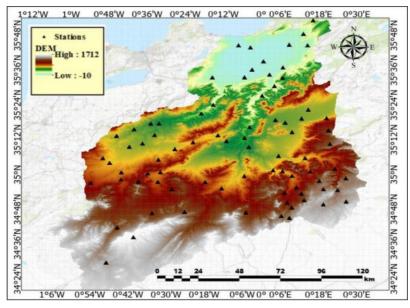
Between Q1 and Q3, is the interquartile range

In the boxplot, the solid line indicates the median and the dashed line indicates the mean.

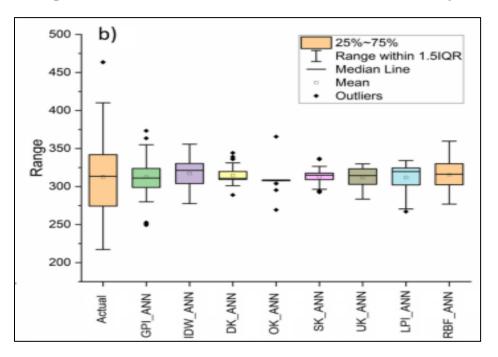
The T-shaped whiskers go to the last point, which is still within 1.5 times the interquartile range.

Points that are further away are considered extreme values (outliers).

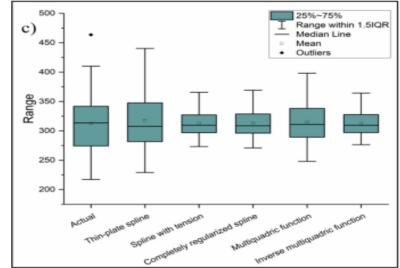
Spatial variation of annual precipitation in the Macta basin (Achite et al. 2023)

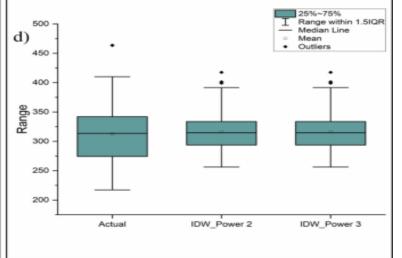






Prediction results to ANN-based precipitation forecast interpolations. **b** Boxplot plots

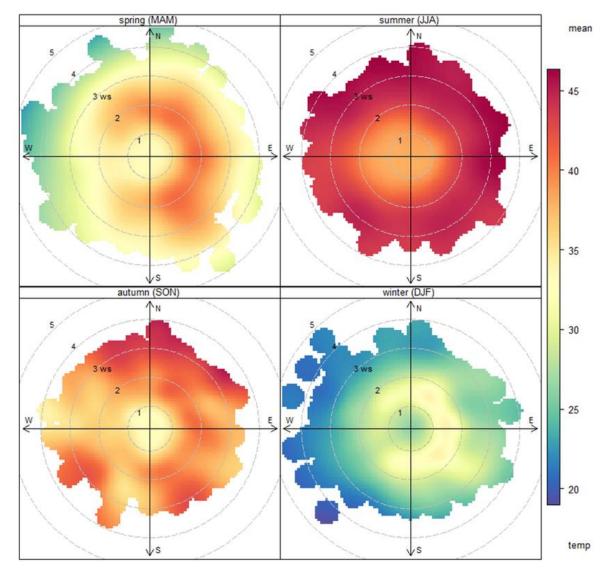




Boxplot graphs of annual precipitation

Radial Plots:

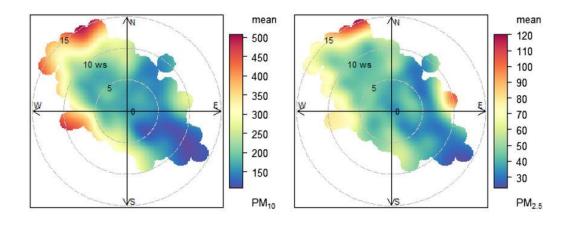
- Radial plots, also known as polar plots, are unconventional but effective for visualizing cyclic or periodic patterns in hydro-climatic data.
- They represent data around a circular axis, with each data point plotted at a specific angle and distance from the center.
- Radial plots are often used to analyze diurnal or seasonal variations in variables like temperature, wind speed, or solar radiation.

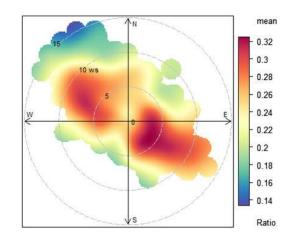


Polar plot of wind speed, wind direction and temperature using data from January 2014 to September 2015 in Makkah

Source: Munir et al. 2016.

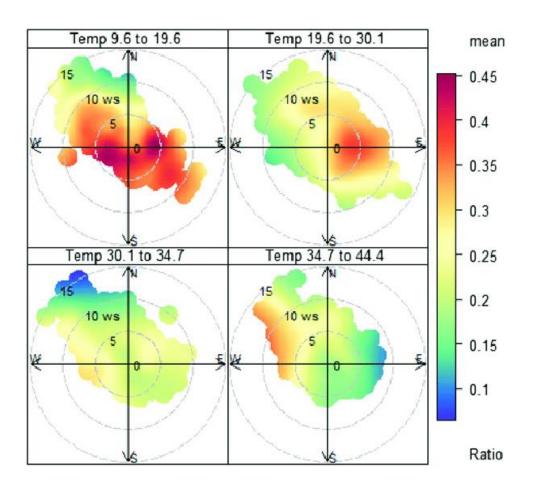
Radial Plots Characterizing temporal variability with meteorological parameters in Bahrain (Coskuner et al. 2018):





Bivariate polar plots of wind speed, wind direction against PM 2.5 and PM 10 concentrations and their ratios

Source: Coskuner et al. 2018



Bivariate polar plots showing the combined effect of wind speed, wind direction, and temperature on the ratios of PM 2.5 /PM 10 in Bahrain

Source: Coskuner et al. 2018