## Capacity Building in Seasonal Hydrological Forecasting

Visualization and Statistical Analysis

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## **Pedagogical Objectives**

**i** Learning outcomes

By the end of this module, participants will be able to:

- Create static plots using ggplot2.
- Customize plots (titles, legends, themes, colors).
- Visualize time series, distributions, and comparisons.
- Combine multiple variables in one plot.
- Produce clear, publication-ready visualizations for hydrological datasets.

#### Data visualization in R

Data visualization helps transform raw numbers into insight. In hydrology, visual analysis reveals **seasonal patterns**, **extremes**, and **model accuracy**.

R offers powerful visualization tools:

- Base R graphics → simple, quick.
- $\bullet$   $ggplot2 \rightarrow$  grammar of graphics, flexible and elegant.
- plotly / tmap / leaflet → interactive plots and maps.

## **Visualizing Time Series**

```
library(ggplot2)
flows <- read.csv("data/flows station.csv")</pre>
flows Date <- as. Date (flows Date)
ggplot(flows, aes(x = Date, y = Qobs)) +
  geom_line(color = "steelblue") +
  labs(title = "Daily Streamflow", y = "Flow (m<sup>3</sup>/s)", x = "") +
  theme_minimal()
```

## **Visualizing Time Series**

- Tips
  - Use geom line() for time series.
  - Use geom\_smooth() to highlight long-term trends.

```
ggplot(flows, aes(x = Date, y = Qobs)) +
  geom_line(color = "grey60") +
  geom_smooth(color = "blue", se = FALSE) +
  labs(title = "Streamflow Trend", y = "Flow (m³/s)", x = "") +
  theme_minimal()
```

### **Distribution Analysis**

```
ggplot(flows, aes(x = Qobs)) +
  geom_histogram(bins = 30, fill = "skyblue", color = "black") +
  labs(title = "Distribution of Streamflow", x = "Flow (m³/s)", y =
  theme minimal()
```

## **Distribution Analysis**

```
ggplot(flows, aes(y = Qobs)) +
  geom_boxplot(fill = "orange") +
  labs(title = "Flow Variability", y = "Flow (m³/s)") +
  theme_minimal()
```

## **Comparing Variables**

```
rain <- read.csv("data/precipitation station.csv")</pre>
rain*Date <- as.Date(rain*Date)
merged <- merge(flows, rain, by = "Date")
ggplot(merged, aes(x = Precip, y = Qobs)) +
  geom point(color = "darkgreen") +
  geom smooth(method = "lm", se = FALSE, color = "red") +
  labs(title = "Relationship between Rainfall and Flow",
  theme minimal()
```

## **Comparing Variables**

Interpretation

A positive relationship between rainfall and discharge indicates a direct hydrological response of the basin.

```
obs \leftarrow c(10, 20, 30, 40, 50)
sim \leftarrow c(12, 18, 32, 37, 52)
R2 <- cor(obs, sim)^2
RMSE <- sqrt(mean((obs - sim)^2))</pre>
```

```
# Nash-Sutcliffe Efficiency
NSE <- 1 - (sum((obs - sim)^2) / sum((obs - mean(obs))^2))

# Kling-Gupta Efficiency
r <- cor(obs, sim)
alpha <- sd(sim)/sd(obs)
beta <- mean(sim)/mean(obs)
KGE <- 1 - sqrt((r - 1)^2 + (alpha - 1)^2 + (beta - 1)^2)</pre>
```

data.frame(R2, RMSE, NSE, KGE)

R2 RMSE NSE KGE 1 0.975418 2.236068 0.975 0.9857471

- Hydrological model evaluation
  - $R^2 \rightarrow$  strength of linear relationship.
  - RMSE  $\rightarrow$  overall error magnitude.
  - NSE and KGE → widely used in hydrology for model performance.

## **Visualizing Model Performance**

## Visualizing Model Performance

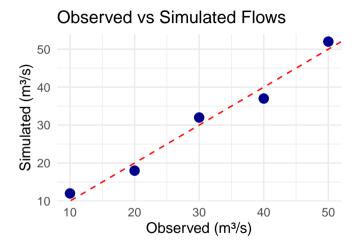


Figure 1: Model performance assesement

#### **Practical Exercises**

#### Exercise 1

- Plot the daily flow series for Station A.
- Add a smoothed trend line and a title.

#### Exercise 2

- Plot the histogram and boxplot of observed flows.
- Comment on the variability.

#### **Practical Exercises**

#### Exercise 3

- Merge rainfall and flow datasets.
- Plot rainfall vs flow scatterplot and compute correlation.

#### Exercise 4

- Given Qobs and Qsim columns in a CSV,
- calculate NSE, RMSE, and KGE using base R formulas.

# THANK YOU FOR YOUR ATTENTATION

