

# DAI Assignment-1 (23114023 – Darsh Jain)

## 1. Data Preprocessing

### 1.1 Data Loading

- The dataset is loaded from weatherAUS.csv using `pandas.read_csv()`, with the Date column parsed as a datetime index.

### 1.2 Handling Missing Values

- The code prints the number of missing values for each column.
- Imputation techniques, including interpolation, are applied.
- Some variables undergo direct removal if they have too many missing values.

### 1.3 Data Summary

- The dataset's shape is printed to understand the number of rows and columns.
  - `data.describe()` provides statistical summaries (mean, min, max, percentiles).
  - `data.info()` helps identify categorical and numerical variables.
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## 2. Exploratory Data Analysis (EDA)

### 2.1 Outlier Detection and Removal

- The Interquartile Range (IQR) method is used to identify and remove outliers.
- The code initially removes outliers but later replaces them with the lower or upper bound instead.

### 2.2 Frequency Distribution

- Histograms and bar plots visualize the frequency distributions of categorical and numerical variables.

### 2.3 Correlation Analysis

- A **heatmap** visualizes the correlation matrix of numerical features.
- Highly correlated variables are identified.

## 3. Feature Engineering: Use of Trigonometric Transformations

### 3.1 Sine and Cosine Transformations

- The notebook applies **sine** and **cosine** transformations to wind direction features:
  - WindGustDirCos, WindGustDirSin
  - WindDir9amCos, WindDir9amSin
  - WindDir3pmCos, WindDir3pmSin

- This transformation helps preserve the circular nature of wind direction data while making it suitable for numerical analysis.
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## 4. Multivariate Analysis

### 4.1 Pair Plots

- `sns.pairplot()` is used to explore relationships between multiple numerical variables.

### 4.2 Scatter Plots (with Multiple Variables)

- A scatter plot is created with:
  - **X-axis:** Temp3pm
  - **Y-axis:** Humidity3pm
  - **Hue:** Location
  - **Style:** WindGustDirCos
  - **Size:** Pressure3pm

### 4.3 Grouped Comparisons

- Box plots and violin plots are used to compare numerical variables across different categorical values.
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## 5. Visualization Techniques

### 5.1 Heatmaps

- Correlation among numerical variables is visualized using a **heatmap**.

### 5.2 Box Plots and Bar Plots

- Box plots visualize distributions and detect outliers.
  - Bar plots compare numerical variables against categorical labels.
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## 6. Conclusion and Insights

- The dataset underwent significant preprocessing, including outlier handling and imputation.
- Exploratory analysis revealed relationships between different weather variables.
- Various visualization techniques helped uncover patterns and trends.
- Multivariate techniques such as scatter plots, pair plots, and heatmaps provided deep insights into the interactions between multiple features.