

WEATHER DATA ANALYSIS REPORT

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SECTION-D

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INTRODUCTION

Weather analysis is a crucial task in understanding climatic patterns and predicting future weather conditions. By analyzing weather data, we can extract meaningful insights that help in various domains such as agriculture, disaster management, and urban planning.

METHODOLOGY

The methodology followed in this report includes:

1. Data Collection: The dataset `weather_data.csv` is used for analysis.

2. Data Preprocessing:

- Load the dataset using Pandas.
- Convert the Date column to datetime format.
- Handle missing values.

3. Exploratory Data Analysis (EDA):

- Display the first few rows of the dataset.
- Generate descriptive statistics.
- Identify missing values.

4. Data Visualization:

- Temperature trends over time.
- Rainfall distribution.
- Wind speed distribution.
- Correlation heatmap of weather variables.

CODE IMPLEMENTATION

```
import pandas as pd

import matplotlib.pyplot as plt

import seaborn as sns


# Load the weather data CSV file

data = pd.read_csv('weather_data.csv')


# Display the first few rows of the data

print(data.head())


# Basic Data Information

print("\nData Info:")

print(data.info())


# Descriptive statistics of numerical columns

*print("\nDescriptive Statistics:")

print(data.describe())


# Check for missing values

print("\nMissing Values:")

print(data.isnull().sum())


# Convert 'Date' column to datetime format if it exists

if 'Date' in data.columns:

    data['Date'] = pd.to_datetime(data['Date'])

else:

    print("\nWarning: 'Date' column is missing!")
```

1. Plot Temperature Over Time

if 'Date' in data.columns and 'Temperature' in data.columns:

```
plt.figure(figsize=(10, 6))

sns.lineplot(x=data['Date'], y=data['Temperature'], label='Temperature (°C)', color='orange')

plt.title('Temperature Over Time')

plt.xlabel('Date')

plt.ylabel('Temperature (°C)')

plt.xticks(rotation=45)

plt.legend()

plt.grid(True)

plt.tight_layout()

plt.show()
```

else:

```
print("\nSkipping Temperature Plot: Missing 'Date' or 'Temperature' column.")
```

2. Plot Rainfall Distribution

if 'Rainfall' in data.columns:

```
plt.figure(figsize=(8, 5))

sns.histplot(data['Rainfall'].dropna(), kde=True, color='blue', bins=20)

plt.title('Rainfall Distribution')

plt.xlabel('Rainfall (mm)')

plt.ylabel('Frequency')

plt.grid(True)

plt.tight_layout()

plt.show()
```

else:

```
print("\nSkipping Rainfall Distribution Plot: 'Rainfall' column missing.")
```

3. Plot Wind Speed Distribution

if 'WindSpeed' in data.columns:

```
plt.figure(figsize=(8, 5))

sns.histplot(data['WindSpeed'].dropna(), kde=True, color='green', bins=20)

plt.title('Wind Speed Distribution')

plt.xlabel('Wind Speed (km/h)')

plt.ylabel('Frequency')

plt.grid(True)

plt.tight_layout()

plt.show()
```

else:

```
print("\nSkipping Wind Speed Distribution Plot: 'WindSpeed' column missing.")
```

4. Correlation Heatmap

required_columns = {'Temperature', 'Rainfall', 'WindSpeed', 'Humidity'}

if required_columns.issubset(data.columns):

```
plt.figure(figsize=(10, 7))

correlation_matrix = data[list(required_columns)].corr()

sns.heatmap(correlation_matrix, annot=True, cmap='coolwarm', fmt='.2f')

plt.title('Correlation Heatmap of Weather Variables')

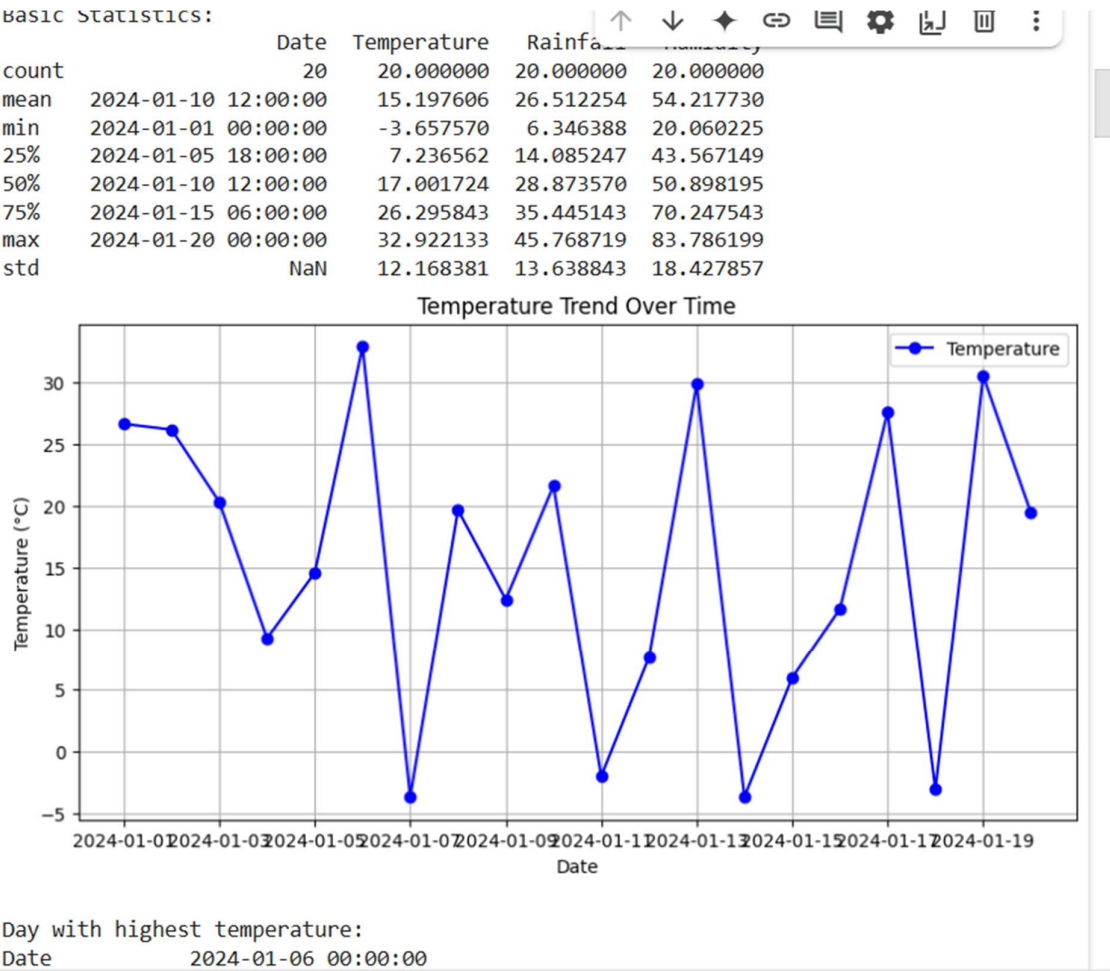
plt.tight_layout()

plt.show()
```

else:

```
print("\nSkipping Correlation Heatmap: Missing one or more required columns.")
```

OUTPUT



REFERENCES

- Seaborn Documentation:
<https://seaborn.pydata.org/>
- Matplotlib Documentation:
<https://matplotlib.org/stable/contents.html>
- Pandas Documentation:
<https://pandas.pydata.org/>
- Dataset source: Provided weather data
(weather_data.csv)