

Problem Statement: Analyzing Weather Data from OpenWeatherMap API Tasks to Perform:

1. Register and obtain API key from OpenWeatherMap.
2. Interact with the OpenWeatherMap API using the API key to retrieve weather data for a specific location.
3. Extract relevant weather attributes such as temperature, humidity, wind speed, and precipitation from the API response.
4. Clean and preprocess the retrieved data, handling missing values or inconsistent formats.
5. Perform data modeling to analyze weather patterns, such as calculating average temperature, maximum/minimum values, or trends over time.
6. Visualize the weather data using appropriate plots, such as line charts, bar plots, or scatter plots, to represent temperature changes, precipitation levels, or wind speed variations.
7. Apply data aggregation techniques to summarize weather statistics by specific time periods (e.g., daily, monthly, seasonal).
8. Incorporate geographical information, if available, to create maps or geospatial visualizations representing weather patterns across different locations.
9. Explore and visualize relationships between weather attributes, such as temperature and humidity, using correlation plots or heatmaps.

API Key: c97ce3f14be7bcf7b544743c2fc6056c

```
In [1]: #Step 1: Import required libraries
import requests
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
from datetime import datetime

plt.style.use('seaborn-v0_8')
```

```
In [2]: #Step 2: Define the API key and location

API_KEY = "c97ce3f14be7bcf7b544743c2fc6056c"
CITY_NAME = "Kochi"
UNITS = "metric"

BASE_URL = "https://api.openweathermap.org/data/2.5/forecast"
```

```
In [3]: #Step 3: Fetch weather data from the API

params = {
    "q": CITY_NAME,
    "appid": API_KEY,
    "units": UNITS
}

response = requests.get(BASE_URL, params=params)

if response.status_code == 200:
    data = response.json()
    print(f"Successfully fetched weather data for {CITY_NAME}")
else:
    print("Failed to fetch data:", response.status_code, response.text)
```

Successfully fetched weather data for Kochi

```
In [4]: # Step 4: Extract relevant weather attributes from the API response

# The forecast data is inside data['list']
weather_list = data['list']

# Create a dataframe from the list
weather_data = pd.json_normalize(weather_list)
```

```

# Select only the relevant columns
weather_data = weather_data[["dt_txt", "main.temp", "main.humidity", "wind.speed",

# Extract weather description
weather_data["weather"] = weather_data["weather"].apply(lambda x: x[0]["description"]

# Convert 'dt_txt' column to datetime type
weather_data["dt_txt"] = pd.to_datetime(weather_data["dt_txt"])

# Rename columns for readability
weather_data.rename(columns={

    "dt_txt": "datetime",
    "main.temp": "temperature",
    "main.humidity": "humidity",
    "wind.speed": "wind_speed",
}, inplace=True)

# Show first few rows
weather_data.head()

```

Out[4]:

| | datetime | temperature | humidity | wind_speed | weather |
|---|---------------------|-------------|----------|------------|-----------------|
| 0 | 2025-10-08 09:00:00 | 26.68 | 73 | 3.67 | light rain |
| 1 | 2025-10-08 12:00:00 | 26.72 | 79 | 3.46 | light rain |
| 2 | 2025-10-08 15:00:00 | 25.74 | 85 | 1.68 | moderate rain |
| 3 | 2025-10-08 18:00:00 | 24.00 | 94 | 2.33 | light rain |
| 4 | 2025-10-08 21:00:00 | 24.08 | 89 | 1.87 | overcast clouds |

In [5]:

```

#Step 5: Clean and preprocess data

print("Missing values before cleaning:")
print(weather_data.isnull().sum())

weather_data.fillna(method='ffill', inplace=True)

print("\nMissing values after cleaning:")
print(weather_data.isnull().sum())

weather_data.info()

```

Missing values before cleaning:

| | |
|-------------|-------|
| datetime | 0 |
| temperature | 0 |
| humidity | 0 |
| wind_speed | 0 |
| weather | 0 |
| dtype: | int64 |

Missing values after cleaning:

| | |
|-------------|-------|
| datetime | 0 |
| temperature | 0 |
| humidity | 0 |
| wind_speed | 0 |
| weather | 0 |
| dtype: | int64 |

<class 'pandas.core.frame.DataFrame'>

RangeIndex: 40 entries, 0 to 39

Data columns (total 5 columns):

| # | Column | Non-Null Count | Dtype |
|---|-------------|----------------|----------------|
| 0 | datetime | 40 non-null | datetime64[ns] |
| 1 | temperature | 40 non-null | float64 |
| 2 | humidity | 40 non-null | int64 |
| 3 | wind_speed | 40 non-null | float64 |
| 4 | weather | 40 non-null | object |

dtypes: datetime64[ns](1), float64(2), int64(1), object(1)

memory usage: 1.7+ KB

```
C:\Users\Prafull Satle\AppData\Local\Temp\ipykernel_15176\2416937720.py:6: FutureWarning: DataFrame.fillna with 'method' is deprecated and will raise in a future version. Use obj.ffill() or obj.bfill() instead.  
    weather_data.fillna(method='ffill', inplace=True)
```

```
In [6]: #Step 6: Perform basic data modeling and descriptive analysis
```

```
print("Basic Weather Statistics:")  
print(weather_data[["temperature", "humidity", "wind_speed"]].describe())  
  
avg_temp = weather_data["temperature"].mean()  
print(f"\n Average Temperature in {CITY_NAME}: {avg_temp:.2f} °C")
```

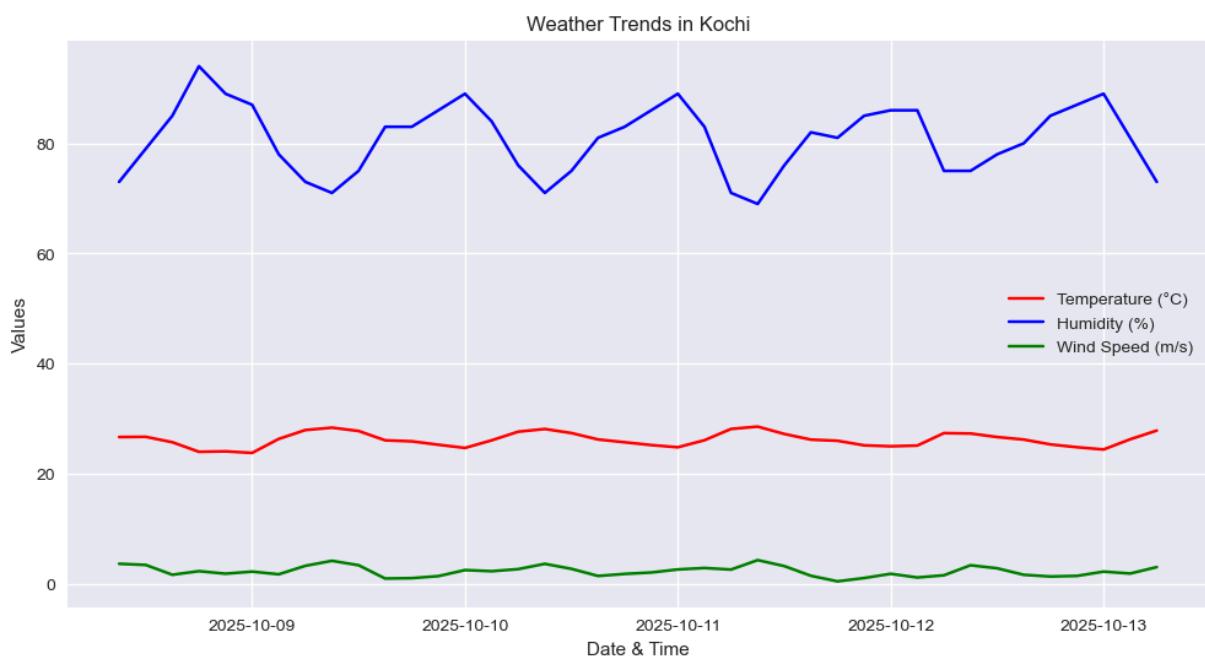
Basic Weather Statistics:

| | temperature | humidity | wind_speed |
|-------|-------------|-----------|------------|
| count | 40.000000 | 40.000000 | 40.000000 |
| mean | 26.215250 | 80.800000 | 2.303500 |
| std | 1.294769 | 6.247871 | 0.940877 |
| min | 23.800000 | 69.000000 | 0.480000 |
| 25% | 25.207500 | 75.000000 | 1.557500 |
| 50% | 26.155000 | 81.500000 | 2.250000 |
| 75% | 27.337500 | 86.000000 | 2.942500 |
| max | 28.570000 | 94.000000 | 4.340000 |

Average Temperature in Kochi: 26.22 °C

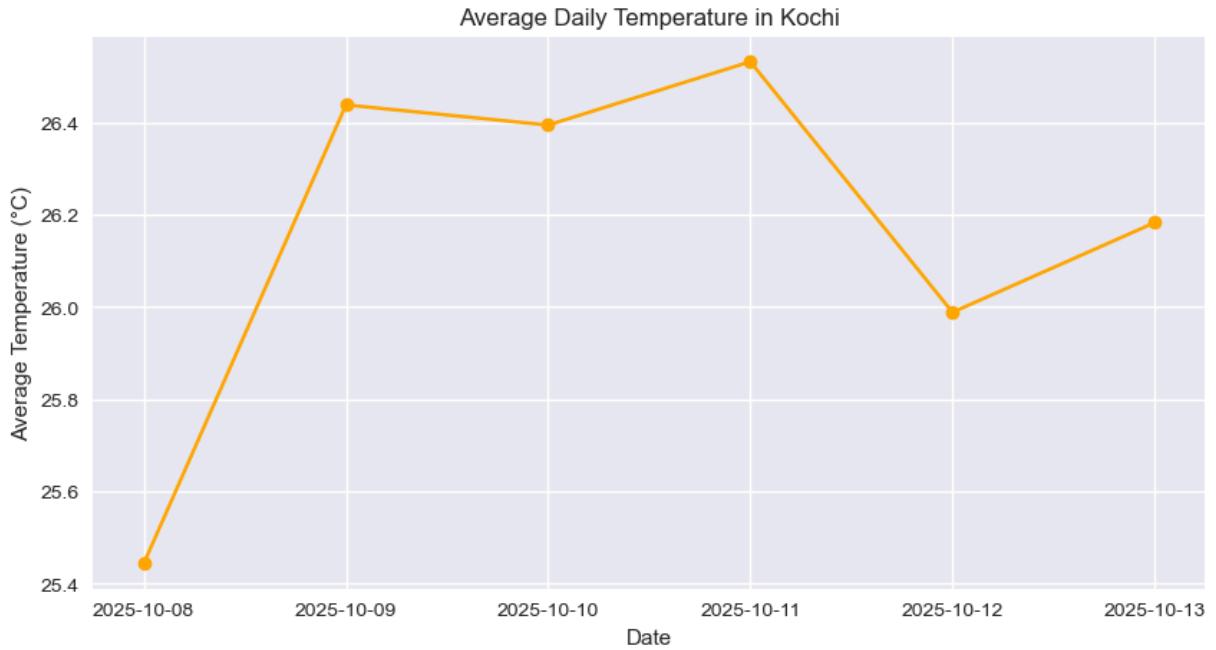
```
In [7]: #Step 7: Visualize temperature, humidity, and wind speed trends over time
```

```
plt.figure(figsize=(12, 6))  
plt.plot(weather_data["datetime"], weather_data["temperature"], label="Temperature")  
plt.plot(weather_data["datetime"], weather_data["humidity"], label="Humidity (%)")  
plt.plot(weather_data["datetime"], weather_data["wind_speed"], label="Wind Speed (m/s)")  
plt.title(f"Weather Trends in {CITY_NAME}")  
plt.xlabel("Date & Time")  
plt.ylabel("Values")  
plt.legend()  
plt.grid(True)  
plt.show()
```



```
In [8]: #Step 8: Aggregate data by day to analyze daily averages
```

```
weather_data["date"] = weather_data["datetime"].dt.date  
daily_avg = weather_data.groupby("date")[["temperature", "humidity", "wind_speed"]]  
  
plt.figure(figsize=(10, 5))  
plt.plot(daily_avg["date"], daily_avg["temperature"], marker='o', color='orange')  
plt.title(f"Average Daily Temperature in {CITY_NAME}")  
plt.xlabel("Date")  
plt.ylabel("Average Temperature (°C)")  
plt.grid(True)  
plt.show()
```



```
In [9]: #Step 9: Explore relationships between weather attributes
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```
plt.figure(figsize=(6, 5))
sns.heatmap(weather_data[["temperature", "humidity", "wind_speed"]].corr(), annot=True)
plt.title("Correlation Between Weather Attributes")
plt.show()
```

