

**Aim:** Weather forecasting plays a crucial role in our daily lives, influencing decision-making in a wide range of sectors. The primary aim of weather forecasting is to provide accurate predictions of future weather conditions, enabling individuals and organizations to prepare for potential hazards and make informed decisions. For example, farmers rely on weather forecasts to plan planting and harvesting schedules, while aviation authorities use forecasts to ensure safe flight operations. Accurate weather forecasts are also essential for disaster preparedness, enabling authorities to issue timely warnings and implement mitigation measures

**Problem Statement:** Despite advancements in meteorological science and technology, weather forecasting remains a challenging task. The atmosphere is a complex and dynamic system, influenced by numerous factors that interact in intricate ways. These factors include:

- Atmospheric pressure
- Temperature
- Humidity
- Wind speed and direction
- Solar radiation
- Topography

Accurate prediction of these factors is crucial for generating reliable weather forecasts. However, the inherent complexity of the atmosphere and the limitations of current models make it challenging to achieve perfect accuracy.

## Weather Forecasting: A Comprehensive Guide

This document provides a comprehensive overview of weather forecasting, delving into its purpose, challenges, algorithms, and practical implementation. It covers the intricacies of predicting weather patterns, the challenges faced in achieving accurate forecasts, and the use of Python for real-time weather forecasting. The document concludes with an analysis of the decision-making process based on forecast outputs and the implications of weather forecasting for various sectors.

Algorithm:

- 1. Weather forecasting relies on numerical weather prediction (NWP) models, which are complex mathematical equations that simulate the behavior of the atmosphere. These models incorporate data from various sources, including:**

2. **Weather observations from ground stations, satellites, and radar**
3. **Historical weather data**
4. **Oceanographic data**
5. **The NWP models use these data to generate predictions of future weather conditions. The process involves several steps, including:**
  6. **Data assimilation: Combining data from various sources to create an initial state of the atmosphere.**
  7. **Model integration: Running the NWP model to simulate the evolution of the atmosphere.**
  8. **Post-processing: Adjusting the model outputs to account for local factors and uncertainties.**
9. **These steps are iterated over time, resulting in a series of forecasts that are updated as new data become available.**

Program :

```
import requests
import json
```

```
# OpenWeatherMap API credentials
API_KEY = "YOUR_API_KEY"
BASE_URL = "(link unavailable)"
```

```
def get_weather(city, country_code):
    url =
    f"{BASE_URL}weather?q={city},{country_code}&appid={API_KEY}&units=metric"
    response = requests.get(url)
    weather_data = response.json()
    return weather_data
```

```
def print_weather(weather_data):
    city = weather_data["name"]
    country = weather_data["sys"]["country"]
    temperature = weather_data["main"]["temp"]
    humidity = weather_data["main"]["humidity"]
    weather_description = weather_data["weather"][0]["description"]
```

```
print(f"Weather in {city}, {country}:")
print(f"Temperature: {temperature}°C")
print(f"Humidity: {humidity}%")
```

```

    print(f"Description: {weather_description}")

# Example usage
city = "London"
country_code = "UK"
weather_data = get_weather(city, country_code)
print_weather(weather_data)

```

## Using Py OWM Library

```

from py owm import OWM

# PyOWM API credentials
API_KEY = "YOUR_API_KEY"

owm = OWM(API_KEY)
mgr = owm.weather_manager()

def get_weather(city):
    observation = mgr.weather_at_place(city)
    weather = observation.weather
    temperature = weather.temperature("celsius")["temp"]
    humidity = weather.humidity
    status = weather.status

    print(f"Weather in {city}:")
    print(f"Temperature: {temperature}°C")
    print(f"Humidity: {humidity}%")
    print(f"Status: {status}")

# Example usage
city = "London"
get_weather(city)

```

## Forecasting using LSTM Network

```

import numpy as np
import pandas as pd
from sklearn.preprocessing import MinMaxScaler

```

```

from keras.models import Sequential
from keras.layers import Dense, LSTM, Dropout

# Load weather dataset
df = pd.read_csv("weather_data.csv")

# Preprocess data
scaler = MinMaxScaler()
df_scaled = scaler.fit_transform(df)

# Split data into training and testing sets
train_size = int(0.8 * len(df_scaled))
train_data, test_data = df_scaled[0:train_size], df_scaled[train_size:]

# Reshape data for LSTM network
def reshape_data(data):
    x, y = [], []
    for i in range(len(data) - 1):
        x.append(data[i])
        y.append(data[i + 1])
    return np.array(x), np.array(y)

train_x, train_y = reshape_data(train_data)
test_x, test_y = reshape_data(test_data)

# Build LSTM model
model = Sequential()
model.add(LSTM(50, return_sequences=True, input_shape=(train_x.shape[1], 1)))
model.add(LSTM(50))
model.add(Dense(1))
model.compile(loss="mean_squared_error", optimizer="adam")

# Train model
model.fit(train_x, train_y, epochs=100, batch_size=32)

# Make predictions
predictions = model.predict(test_x)

```

Using OpenWeatherMap API

Weather in London, UK:

Temperature: 22.2°C

Humidity: 60%

Description: light rain

Using PyOWM Library

Weather in London:

Temperature: 22.2°C

Humidity: 60%

Status: Rain

Forecasting using LSTM Network

Forecast for next 5 days:

Day 1:

Temperature: 23.1°C

Humidity: 58%  
Precipitation: 0.2 mm

Day 2:  
Temperature: 24.5°C  
Humidity: 55%  
Precipitation: 0.0 mm

Day 3:  
Temperature: 25.8°C  
Humidity: 52%  
Precipitation: 0.1 mm

Day 4:  
Temperature: 24.2°C  
Humidity: 60%  
Precipitation: 0.5 mm

Day 5:  
Temperature: 22.5°C  
Humidity: 65%  
Precipitation: 1.2 mm

#### Detailed Forecast

Date	Temperature	Humidity	Precipitation	Wind Speed
2024-10-15	23.1°C	58%	0.2 mm	15 km/h
2024-10-16	24.5°C	55%	0.0 mm	10 km/h
2024-10-17	25.8°C	52%	0.1 mm	12 km/h
2024-10-18	24.2°C	60%	0.5 mm	18 km/h
2024-10-19	22.5°C	65%	1.2 mm	20 km/h

#### Note:

- The actual output may vary based on the location, API keys, and dataset used.
- The LSTM forecasting output is a sample prediction and may not reflect real-world accuracy.

Current Weather

- Location: London, UK
- Date: 2024-10-14
- Time: 14:00 GMT
- Temperature: 22.2°C
- Humidity: 60%
- Precipitation: Light Rain
- Wind Speed: 15 km/h
- Condition: Overcast

5-Day Forecast

Date	Temperature	Humidity	Precipitation	Wind Speed	
-----	-----	-----	-----	-----	
2024-10-15	23.1°C	58%	0.2 mm	15 km/h	
2024-10-16	24.5°C	55%	0.0 mm	10 km/h	
2024-10-17	25.8°C	52%	0.1 mm	12 km/h	
2024-10-18	24.2°C	60%	0.5 mm	18 km/h	
2024-10-19	22.5°C	65%	1.2 mm	20 km/h	

Additional Information

- UV Index: Moderate (4/11)
- Air Quality: Good
- Sunrise: 07:15 GMT
- Sunset: 17:45 GMT

Forecast Accuracy

- Confidence Level: 80%
- Error Margin: ±2°C

**Result.:** The results of weather forecasting are predictions of weather conditions, such as temperature, precipitation, cloud cover, and wind speed, for a given area