Task: Build a Smart Weather Forecasting AI Agent

Objective

Design and implement an **Al-powered weather agent** capable of understanding user queries, generating weather forecasts, and presenting the output in both **tabular** and **chart** formats based on user instructions.

© Key Goals

Your AI Weather Agent should:

- 1. Ingest and process historical weather data (CSV or API-based).
- 2. Forecast weather for upcoming days (using ML/statistical methods).
- 3. Respond to natural language questions, e.g.:
 - "What's the weather like for the next 3 days in Mumbai or any city or country?"
 - "Show me the temperature forecast for this week as a chart."
 - "Summarize last week's weather in table format."

4. Render results in:

- Natural language (text)
- o **Tables** (e.g., temperature, humidity by day)
- Charts (line/bar chart of forecast data)

K Functional Requirements

1. Data Ingestion

- Use publicly available historical weather data (e.g., NOAA, Kaggle, or CSV file provided).
- o Optional: Connect to a live weather API like OpenWeatherMap.

2. Forecasting

- Forecast weather conditions (temperature, humidity, etc.) for the next few days.
- You may use simple statistical models (e.g., moving average) or ML models (e.g., linear regression, Prophet, etc.).

3. Query Handling

- o Agent should understand user queries and determine:
 - Location (if applicable)
 - Duration (e.g., next 3 days, last 7 days)
 - Output format (chart, table, text)

4. Output Presentation

- Text: Descriptive summary.
- o Table: Use pandas/HTML tables to show data.
- o Chart: Use matplotlib or plotly to generate forecast charts dynamically.

Tech Stack (Recommended)

You are free to choose the tools, but here's a suggested stack:

- Backend: Python with FastAPI or Flask.
- Data: CSV, Pandas
- Forecasting: scikit-learn, Prophet, statsmodels, or any ML/time-series package
- Visualization: matplotlib, seaborn, or plotly
- Optional LLM-based Query Understanding: transformers, langehain, etc.

Deliverables

- 1. \bigcirc Codebase with instructions to run locally.
- 2. Sample dataset or API integration.
- 3. API or CLI to test the agent's responses.
- 4. Sample queries and outputs:
 - o Sample natural language questions.
 - Corresponding outputs: chart images, tables, or text.

Page 15 Bonus (Optional)

- Multi-city support.
- Z Streamed/chart-first responses.
- Add simple authentication (API key or JWT).
- See LLM for parsing query intent (e.g., HuggingFace models).

• Basic web interface to test the agent.

Timeline

- Expected Time: 2–3 days.
- Please ensure that the submission is clean, documented, and easy to run.