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Team ID	LTVIP2025TMIDS40838
Project Name	Weather-Based Prediction of Wind Turbine Energy Output: A Next-Generation Approach to Renewable Energy
Maximum Marks	3 Marks

# **Phase 5: Project Executable Files (Expanded)**

#### 1. Overview

This phase documents all executable components of the Wind Turbine Energy Prediction project, including:

- Flask Web Application for user interaction
- Jupyter Notebooks for data preprocessing, model training, testing, and deployment

The goal is to provide a fully functional system that predicts wind energy output based on real-time or historical weather data.

# 2. Flask Web Application

#### **Purpose:**

• To allow end-users to input weather parameters and receive predicted wind turbine energy output in real-time.

#### **Features:**

- **Input:** Wind speed, temperature, humidity, air pressure, and optional date/time for historical prediction
- **Output:** Estimated energy generation (kWh or MW)
- Interface: Clean and user-friendly web interface with responsive design
- **Integration:** Connects directly to the trained machine learning model

#### Workflow:

- 1. User enters weather parameters in the web form.
- 2. Flask backend receives the input and preprocesses it.
- 3. Preprocessed data is fed into the trained ML model.
- 4. Model predicts energy output and returns the result on the web page.

#### **Technology Stack:**

- Python (Flask framework)
- HTML/CSS for interface
- Pickle/Joblib for loading trained model
- Bootstrap or simple CSS for layout

# 3. Jupyter Notebooks

#### a) model\_training.ipynb

- **Purpose:** Preprocess the data and train machine learning models.
- Contents:
  - o Importing libraries (Pandas, NumPy, Scikit-learn)
  - Data cleaning, normalization, feature engineering
  - Model selection (Linear Regression, Random Forest, Gradient Boosting, XGBoost)
  - Evaluation using RMSE, MAE, R<sup>2</sup>
  - Saving the best-performing model using Pickle/Joblib

### b) model\_testing.ipynb

Purpose: Test the trained model on unseen data to validate predictions.

#### • Contents:

- Loading the saved model
- o Input test data for prediction
- Compare predicted vs actual energy output
- Generate plots for visual comparison (scatter plots, line charts)
- Performance metrics calculation for model evaluation

### c) Flask\_App.ipynb (or app.py)

• Purpose: Integrate the trained model into a Flask application for deployment.

#### • Contents:

- Flask app setup (Flask(\_\_name\_\_))
- o Routes for home page and prediction endpoint
- o Input form processing and validation
- Model prediction and output rendering
- Launching the app locally for testing (app.run(debug=True))

## 4. File Structure (Example)

```
Turbine_Project/

|-- notebooks/
|-- model_training.ipynb
|-- model_testing.ipynb
|-- Flask_App.ipynb
|-- templates/
|-- templates/
|-- static/
|-- style.css
```

$\vdash$	- saved_model/
	└─ wind_model.pkl
L	README md

### **5. Expected Outcome**

- Fully functional web application capable of predicting wind turbine energy output.
- Reproducible machine learning workflow from data preprocessing to model deployment.
- Interactive interface for stakeholders to input weather data and get predictions.
- Clear documentation of model performance, preprocessing steps, and deployment setup.

### 6. Additional Notes

- The Flask app can be deployed on local machines or cloud platforms (e.g., Heroku).
- The notebooks allow for continuous model improvement by retraining with new data.
- The project is designed to be **scalable**, allowing integration with multiple turbines or farms.