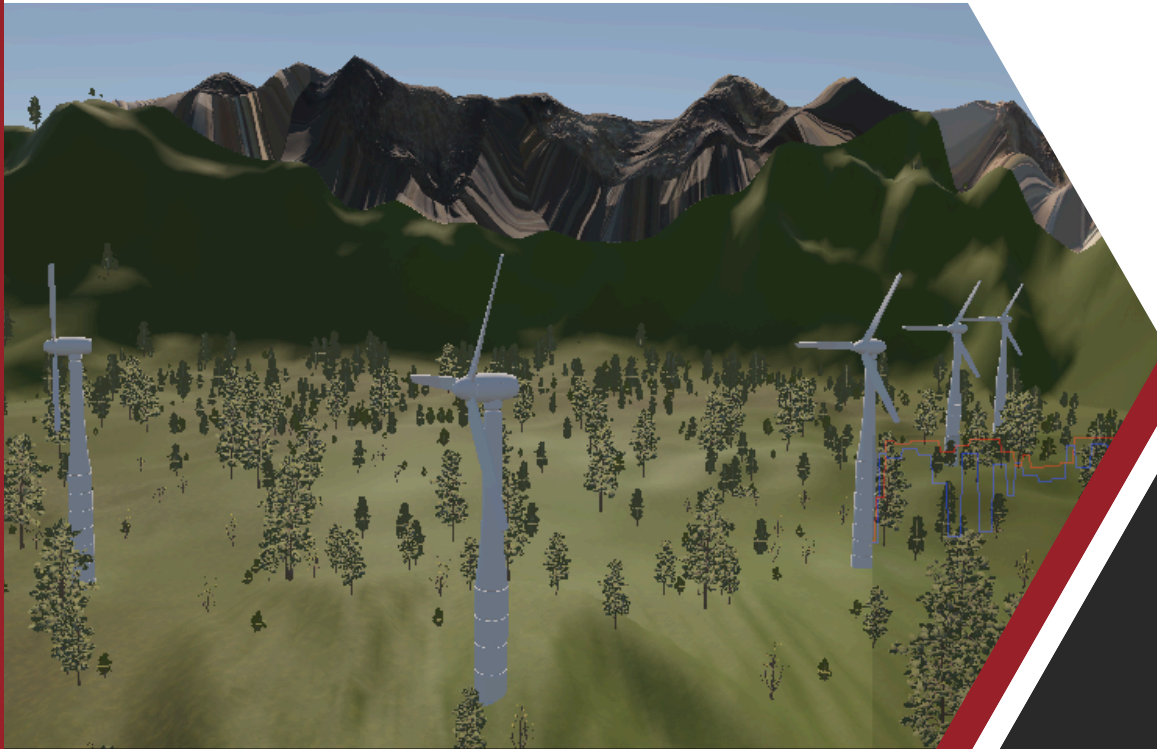




DTCA

WIND FARM DIGITAL TWIN WITH ML INTEGRATION

GROUP - 3



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INTRODUCTION

Wind energy is a key contributor to global sustainable power generation.

Accurate prediction of wind turbine output is essential for monitoring, planning, and digital-twin applications.

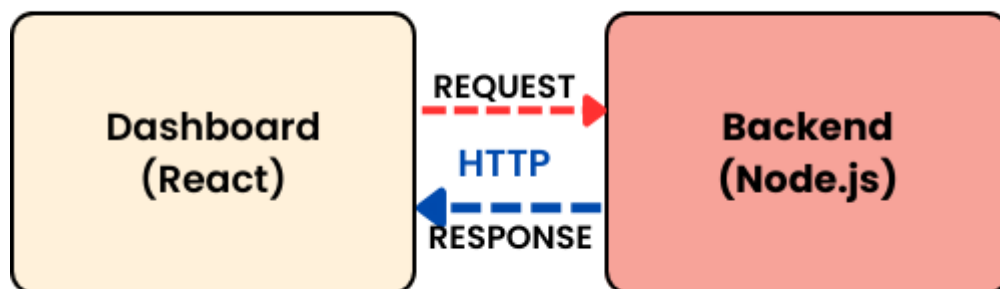
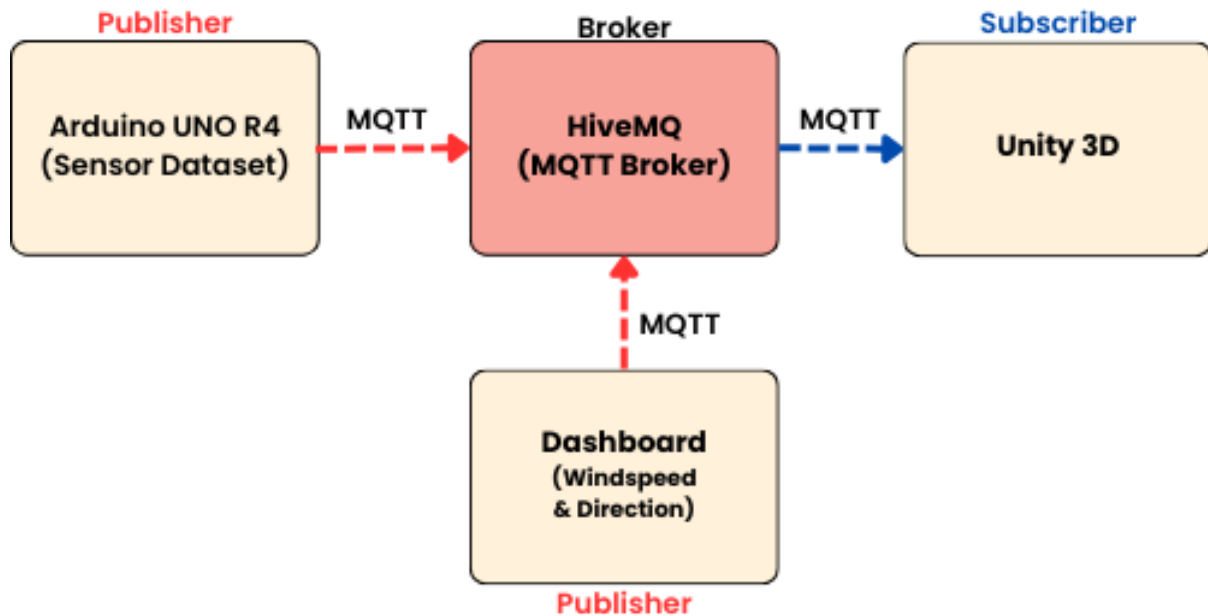
This project develops a machine learning model that predicts turbine Power Generated (in kW) using Wind Speed (m/s) from a real [SCADA data on Kaggle](#).

The model forms the core predictive engine of a Unity-based digital twin, which combines:

- Live IoT sensor data (Arduino → MQTT → Unity)
- A predictive simulator (ML model in ONNX format)
- A 3D visual farm environment (Unity)

The final ML model is exported as ONNX, enabling efficient real-time inference inside Unity.

BLOCK DIAGRAM



ML MODEL

1. **Random Forest** Regression Model used in our project uses an ensemble of multiple decision trees, combining their outputs to produce accurate and stable predictions even in noisy real-world datasets.
2. It captures **nonlinear relationships** between wind speed and power output, making it ideal for SCADA-based wind turbine modeling.
3. The model is **highly robust**, **reduces overfitting** through averaging, and **delivers reliable performance** suitable for deployment in real-time digital twin systems.

Our model configuration:

- `n_estimators=100` - Create 100 different trees
- `max_depth=20` - Each tree can ask up to 20 questions
- `min_samples_split=5` - Need at least 5 examples to create a split
- `min_samples_leaf=2` - Each final answer must have at least 2 examples
- `random_state=42` - Ensures same results every time
- `n_jobs=-1` - Use all CPU cores for faster training

Training process:

The model looks at 40,424 training examples and learns patterns:

- "When wind is 7.5 m/s, power is usually around 1,150 kW"
- "When wind is 12 m/s, power is usually around 2,400 kW"
- And so on...

ACCURACY SCORES:

- Training Set Accuracy: 94.75% ($R^2 = 0.9475$)
- Testing Set Accuracy: 89.20% ($R^2 = 0.8920$)
- Model Generalization: Excellent

ERROR METRICS:

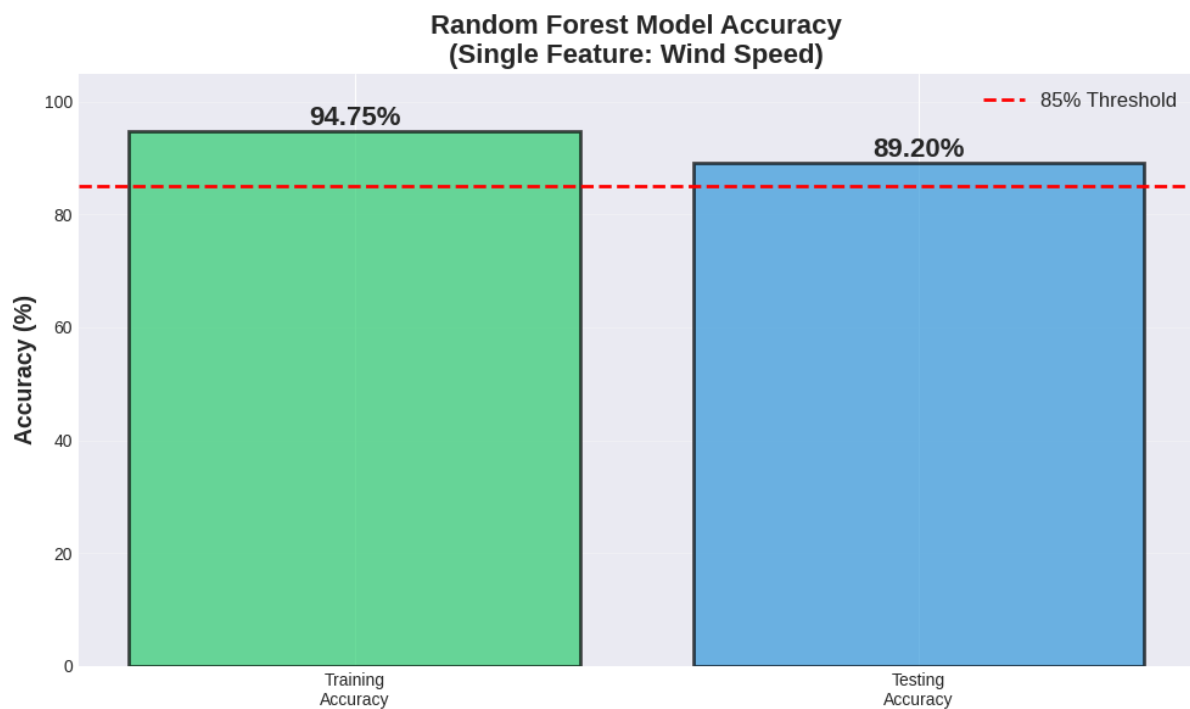
1) TRAINING SET:

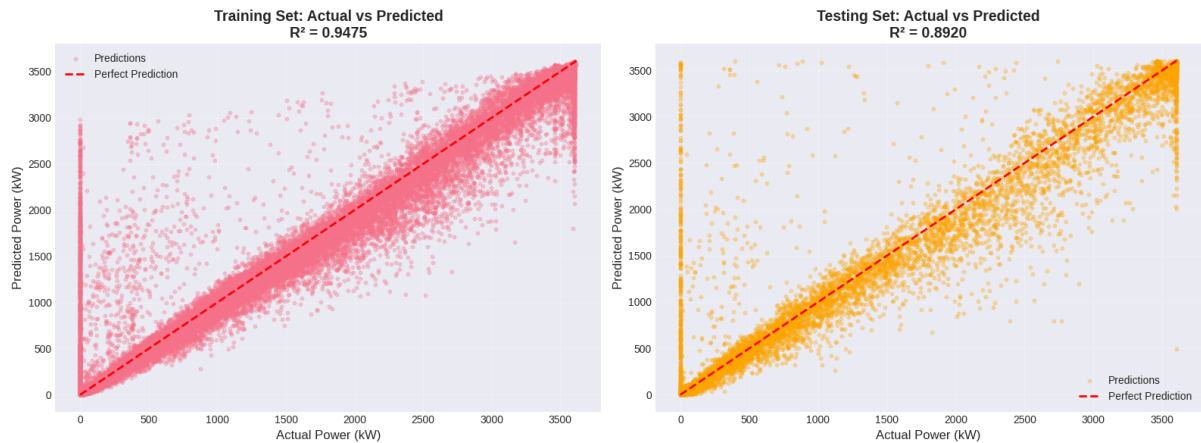
- Root Mean Square Error (RMSE): 301.12 kW
- Mean Absolute Error (MAE): 124.85 kW
- R^2 Score: 0.9475

2) TESTING SET:

- Root Mean Square Error (RMSE): 429.23 kW
- Mean Absolute Error (MAE): 181.72 kW
- R^2 Score: 0.8920

MODEL EXPLAINS 89.20% OF VARIANCE IN POWER GENERATION





Visualization In Unity

Digital Twin – Unity Implementation

1). Wind Turbine System

- TurbineController.cs rotates blades based on wind speed
- TurbineController.cs handles yaw → smoothly faces WindZone direction
- Works with all sources: CSV (CsvPlaybackManager.cs), MQTT (MqttWindClient.cs), manual mode (SimulationManager.cs)

2). Wind Environment

- WindManager.cs updates WindZone speed + direction
- WindManager.cs drives particle direction & color
- WindParticles VFX uses elongated particles for wind visualization

3). Data Input Modes (Controlled by SimulationManager.cs)

- CSV playback (CsvPlaybackManager.cs)
- MQTT–Arduino (MqttWindClient.cs)
- MQTT–Website (MqttWindClient.cs)
- Manual mode using inspector booleans in SimulationManager.cs

4). CSV Playback System (CsvPlaybackManager.cs)

- Reads dataset (timestamp, wind speed, direction, active power)
- 144 samples per day
- Interpolates between samples
- Adjustable simulation time scaling (SecondsPerDay)

5). ONNX Power Prediction (OnnxPowerPredictor.cs)

- Loads model.onnx from StreamingAssets
- Feeds wind speed to ONNX model
- Outputs predictedPower
- Accessible globally for graph + logging

6). Power Comparison Logging (OnnxPowerPredictor.cs)

- Creates power_comparison.csv
- Logs: DateTime, WindSpeed, WindDirection, ActivePower, PredictedPower
- Syncs logging with index from CsvPlaybackManager.cs or MQTT timestamps

7). MQTT Integration (MqttWindClient.cs)

- Uses MQTTnet + MQTTnet.Extensions.ManagedClient DLLs
- Secure TLS connection (port 8883)
- Subscribes to JSON payload { windSpeed, windDirection, activePower, timestamp }
- Sends received values to SimulationManager.cs

8). Simulation Manager (SimulationManager.cs)

- Master selection of data sources: CSV / MQTT / Manual
- Feeds wind speed & direction to WindManager.cs
- Feeds actual power to graph via PowerGraphUI.cs
- Coordinates ONNX prediction per frame

9). Real-Time Graph Visualization (PowerGraphUI.cs)

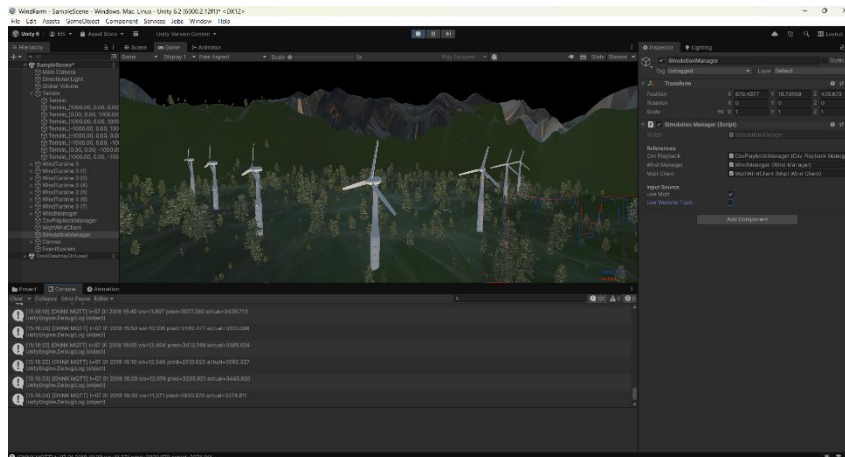
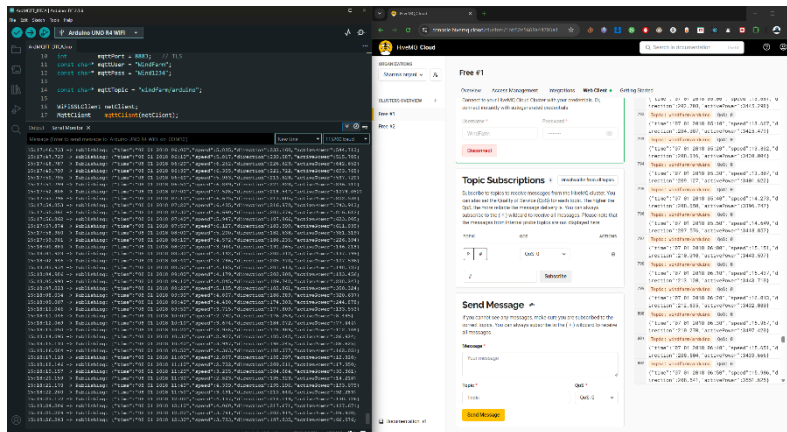
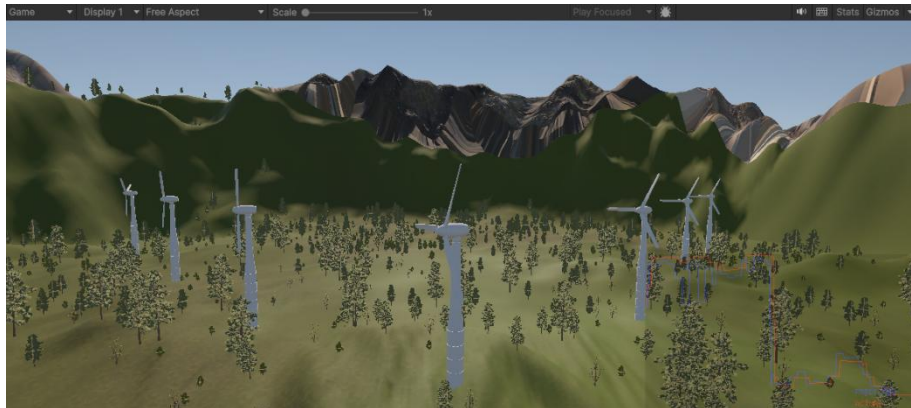
- Draws predicted vs actual power
- Uses runtime texture for graph
- Interpolates plotted points
- Adds legend text: "Predicted", "Actual"

10). Project Output

- Animated wind turbines reacting to real wind data
- Real-time ML prediction (via OnnxPowerPredictor.cs)
- Real-time telemetry via MQTT (MqttWindClient.cs)
- Real-time energy graph (PowerGraphUI.cs)
- CSV logging (OnnxPowerPredictor.cs) for MATLAB/Python analysis

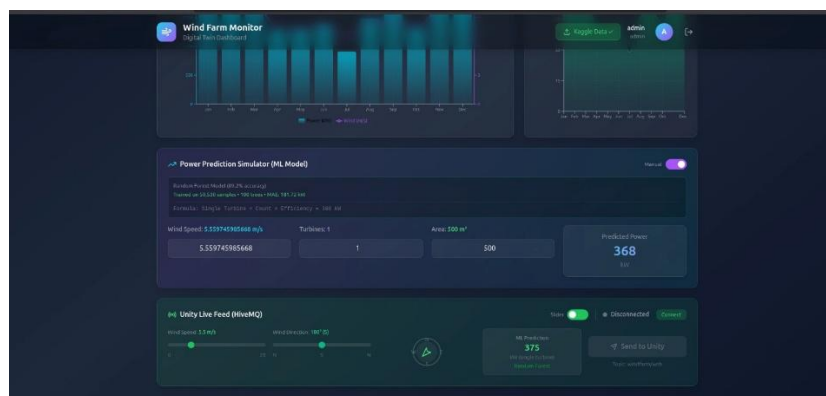
RESULTS

1). Unity 3D



2). Dashboard

The screenshot shows a web browser window with the URL 'localhost:3073'. The page has a dark purple gradient background. In the center, there is a white 'Create Account' form for 'Jon Wind Farm Monitor'. The form includes fields for 'Username' (containing 'USER1'), 'Email' (containing 'user1@gmail.com'), 'Password', and 'Confirm Password'. A 'Sign Up' button is at the bottom, and a link for 'Already have an account? Sign in' is below it.



**THANK
YOU**