# PREDICTING THE ENERGY OUTPUT OF WIND TURBINE BASED ON WEATHER CONDITIONS



### **INDEX**

- Introduction of Team members
- Road Map
- Advantages and Disadvantages
- Conclusion
- References

### TEAM MEMBERS:

Member 1 and Leader

Name: D.HEMANATHAN

College:HITECH,CBE

Member 2

Name: GURRAM KUSUMA

College: HITECH, CBE

Member 3

Name: D. HEMAMALINI

College: HITECH, CBE

Member 4

Name: S. HEMNATH

College: HITECH, CBE

### ROAD MAP

1. Work Distribution

2. Data Collection

> 4. Model Design

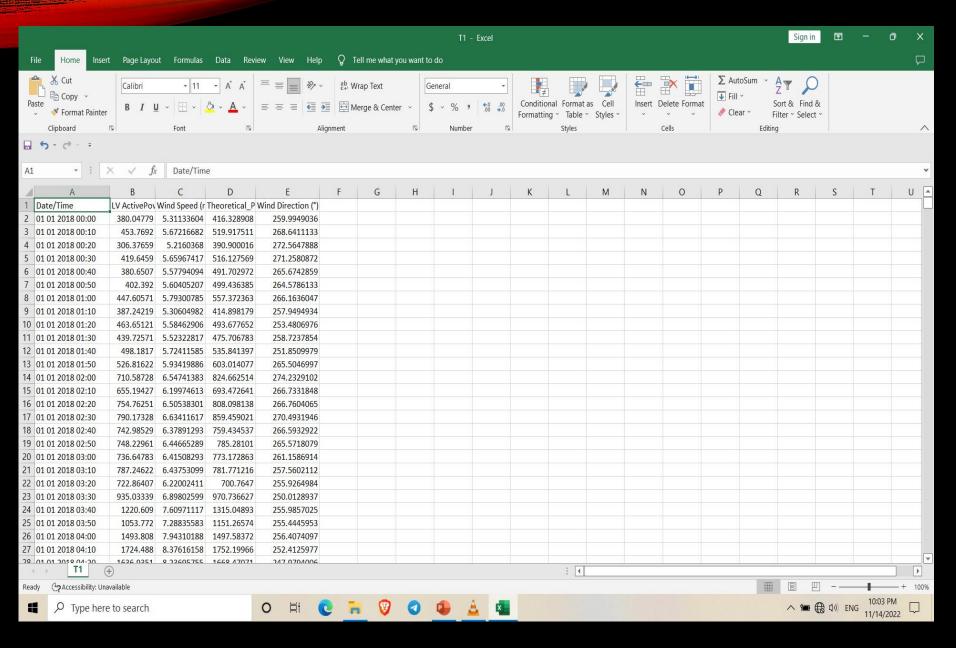
> > 3.Data Preprocessing

> > > 5. Application Desig

### DATA COLLECTION

 We used the wind turbine dataset from Kaggle provided in the description of the problem and downloaded the weather conditions data of the same place and of the same time from the web.

### KAGGLE DATA



### WEATHER DATA

Cli	pboard	<u>[</u> 2]	Font	12	Alignment		Number	Number 5		Styles			Cells		Editing		
	* 1	× ✓ fx	45 F Full-screen Snip														
	Α	В	С	D	E	F	G	Н	1	J	K	L	М	N	0	Р	(
	Time	Temperature	Dew Point	Humidity	Wind	Wind Speed	Wind Gust	Pressure	Precipitation	Condition							
1	12:20 AM	43 F	41 F	93 %	SW	7 mph	0 mph	29.75 in	0.0 in	Fair							
1	12:50 AM	43 F	39 F	87 %	WSW	7 mph	0 mph	29.75 in	0.0 in	Fair							
	1:20 AM	45 F	39 F	81 %	WSW	7 mph	0 mph	29.75 in	0.0 in	Partly Cloudy							
	1:50 AM	45 F	39 F	81 %	SW	7 mph	0 mph	29.75 in	0.0 in	Partly Cloudy							
1	2:20 AM	45 F	41 F	87 %	SW	8 mph	0 mph	29.75 in	0.0 in	Partly Cloudy							
	3:50 AM	45 F	41 F	87 %	SW	6 mph	0 mph	29.75 in	0.0 in	Partly Cloudy							
	4:20 AM	45 F	41 F	87 %	SSW	8 mph	0 mph	29.75 in	0.0 in	Partly Cloudy							
	4:50 AM	45 F	41 F	87 %	SW	10 mph	0 mph	29.75 in	0.0 in	Partly Cloudy							
7	5:20 AM	45 F	41 F	87 %	SW	9 mph	0 mph	29.75 in	0.0 in	Partly Cloudy							
	5:50 AM	45 F	39 F	81 %	SW	9 mph	0 mph	29.72 in	0.0 in	Partly Cloudy							
	6:20 AM	43 F	39 F	87 %	WSW	7 mph	0 mph	29.75 in	0.0 in	Partly Cloudy							
	6:50 AM	41 F	37 F	87 %	SW	7 mph	0 mph	29.72 in	0.0 in	Partly Cloudy							
	7:20 AM	39 F	37 F	93 %	8	6 mph	0 mph	29.72 in	0.0 in	Fair							
4	7:50 AM	41 F	37 F	87 %	SSE	5 mph	0 mph	29.72 in	0.0 in	Fair							
	8:20 AM	41 F	39 F	93 %	SSW	3 mph	0 mph	29.72 in	0.0 in	Fair							
	8:50 AM	43 F	39 F	87 %	8	5 mph	0 mph	29.72 in	0.0 in	Fair							
	9:20 AM	45 F	39 F	81 %	SSW	3 mph	0 mph	29.75 in	0.0 in	Fair							
	9:50 AM	45 F	37 F	76 %	S	6 mph	0 mph	29.75 in	0.0 in	Fair							
	10:20 AM	46 F	39 F	76 %	S	7 mph	0 mph	29.75 in	0.0 in	Fair							
	10:50 AM	46 F	39 F	76 %	S	7 mph	0 mph	29.75 in	0.0 in	Fair							
1	11:20 AM	50 F	39 F	66 %	SSW	8 mph	0 mph	29.72 in	0.0 in	Fair							
	11:50 AM	50 F	37 F	62 %	SSW	8 mph	0 mph	29.72 in	0.0 in	Fair							
	2:20 PM	50 F	37 F	62 %	SSW	8 mph	0 mph	29.72 in	0.0 in	Fair							
4	2:50 PM	50 F	37 F	62 %	SW	10 mph	0 mph	29.70 in	0.0 in	Fair							
	1:20 PM	52 F	37 F	58 %	WSW	10 mph	0 mph	29.70 in	0.0 in	Fair							
	1:50 PM	52 F	37 F	58 %	SW	13 mph	0 mph	29.70 in	0.0 in	Fair							
	2:20 PM	52 F	39 F	62 %	WSW	15 mph	0 mph	29.70 in	0.0 in	Fair							
	Sheet1	Sheet2   Shee							1 4								

Ps Anaconda Pr\_ X Weather Dat\_ 🛆 🖜 (6 U)) FNG

### Data pre-processing

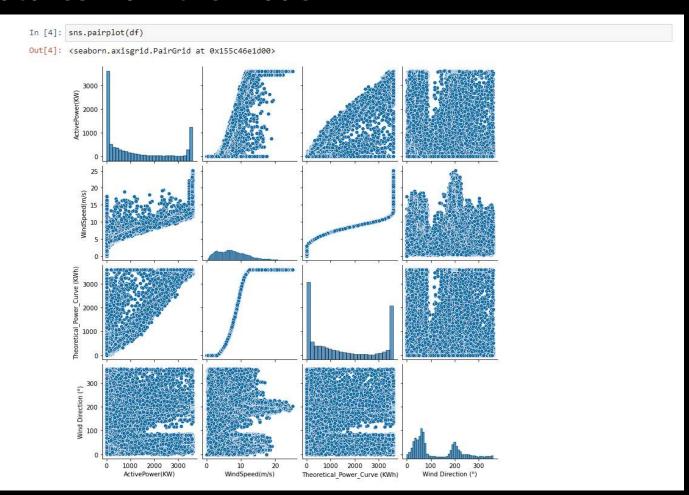
- Activity 1: Taking care of the missing data
- Solution The missing data is less than 3% of the total data so we have filled the missing data by the average value of the respective column data.
- Activity 2: Feature scaling
- Solution \_ Weather data has a scale of 30 minutes while the wind turbine data from Kaggle has a scale of 10 minutes so in order to merge both the data's we have converted each of them into the same scale (i.e. scale of 30 minutes). Then we merged both the data's into one and obtained our final data.

## Handle Missing Data and Merging the data

#### DATA PRE-PROCESSING

```
In [2]: import numpy as np
        import pandas as pd
        import matplotlib.pyplot as plt
        import seaborn as sns
        import joblib
In [3]: path = "T1.csv"
        df = pd.read csv(path)
        df.rename(columns={"Date/Time":"Time",
                            "LV ActivePower (kW)": "ActivePower(KW)",
                            "Wind Speed (m/s)": "WindSpeed(m/s)",
                            "Wind Direction(°)": "Wind Direction"},
                            inplace=True)
In [4]: df.isnull().sum()
Out[4]: Time
        ActivePower(KW)
        WindSpeed(m/s)
        Theoretical Power Curve (KWh)
        Wind Direction (°)
        dtype: int64
```

- Activity 3: Visualization of the Data
- Solution- We analyzed the data and studied the impact of different features on output (active power). With the results we achieved, we eradicated all the unnecessary features from the model



```
In [5]: plt.figure(figsize=(10, 8))
         corr = df.corr()
         ax = sns.heatmap(corr, vmin = -1,vmax = 1,annot = True)
        bottom, top = ax.get_ylim()
        ax.set_ylim(bottom + 0.5, top - 0.5)
        print(corr)
                                          ActivePower(KW) WindSpeed(m/s) \
         ActivePower(KW)
                                                 1.000000
                                                                  0.912774
                                                 0.912774
                                                                  1.000000
        WindSpeed(m/s)
         Theoretical_Power_Curve (KWh)
                                                 0.949918
                                                                  0.944209
        Wind Direction (°)
                                                -0.062702
                                                                 -0.077188
                                         Theoretical_Power_Curve (KWh) \
         ActivePower(KW)
                                                                0.949918
        WindSpeed(m/s)
                                                                0.944209
         Theoretical_Power_Curve (KWh)
                                                                1.000000
        Wind Direction (°)
                                                               -0.099076
                                          Wind Direction (°)
         ActivePower(KW)
                                                   -0.062702
        WindSpeed(m/s)
                                                   -0.077188
         Theoretical_Power_Curve (KWh)
                                                   -0.099076
        Wind Direction (°)
                                                    1.000000
                                                                                                      -1.00
                                                                                                      -0.75
                                                      0.91
                                                                      0.95
                    ActivePower(KW)
                                                                                                      -0.50
                                                                                                      -0.25
                    WindSpeed(m/s)
                                       0.91
                                                                      0.94
                                                                                                       0.00
          Theoretical Power Curve (KWh)
                                       0.95
                                                      0.94
                                                                                                       -0.25
                                                                                                       -0.50
                   Wind Direction (°)
                                                                                                       -0.75
                                                                                      Wind Direction (*) -
```

### **Model Design**

- Activity 1: Deciding parameters or features
- **Solution-** With the results we achieved we have selected following as our features for the model :
  - 1.Wind Speed
  - 2.Temperature
  - 3. Humidity
  - 4.Pressure
- Activity 2: Separating Training and Testing data from the data set
- **Solution-** This process will be done inside the model file itself by using the train\_test\_split library in the sklearn package of python.
- Activity 3: Developing the model
- **Solution** We use the boosted trees regressor model to train and evaluate the model using TensorFlow. We decided to train the model using 120 trees to optimize the model and preventing it from underfit also we have used entire batch of training data to train so that model can accurately be trained.

### APPLICATION DESIGN

 Our initial plans included building the website in Html, but we wanted to provide our end-user the solution in single click, . We used Flask and Html framework for designing the UI of the website and Juypter Python for the back-end.

- Activity 1: Build the UI
- **Solution-** We used Flask and Html to design our Website. Flask reduces code development time to huge extent because of its UI and html access. It allows seeing the applied changes almost instantly, without even losing the current application state. Apart from this, development in Flask is very easy because it uses the unique Widget tree model. Everything in Flask UI is just a Libraries. "App route" widget is the root widget of the website and all the other libraries become its children in hierarchial pattern.

```
In [1]: import numpy as np
         from flask import Flask, request, jsonify, render_template
        import joblib
        import requests
In [2]: app = Flask( name )
        model = joblib.load("power_prediction.sav")
In [3]: @app.route('/')
        def home():
            return render_template("intro.html")
        @app.route('/predict')
        def predict():
            return render_template("predict.html")
        @app.route('/windapi', methods=['POST'])
        def windapi():
            city = request.form.get('city')
            apikey = "d8484354b9e388875c48dae8d0d09cd1"
            url = "http://api.openweathermap.org/data/2.5/weather?q=" + city + "&appid=" + apikey
            resp = requests.get(url)
            resp = resp.ison()
            temp = str(resp["main"]["temp"]) +" °C"
            humid = str(resp["main"]["humidity"]) + " %"
            pressure = str(resp["main"]["pressure"]) + " mmHG"
            speed = str(resp["wind"]["speed"]) + " m/s"
            return render_template('predict.html', temp=temp, humid = humid, pressure=pressure, speed = speed)
In [4]: @app.route('/y_predict',methods = ['POST'])
        def y predict():
            x_test = [[float(x) for x in request.form.values()]]
            prediction = model.predict(x_test)
            print(prediction)
            output = prediction[0]
            return render template('predict.html', prediction text = 'The energy predicted is {:.2f} KWh'.format(output))
In [ ]: if __name__ == "__main__";
            app.run(debug=False)
         * Serving Flask app "__main__" (lazy loading)
         * Environment: production
           WARNING: This is a development server. Do not use it in a production deployment.
           Use a production WSGI server instead.
         * Debug mode: off
         * Running on http://127.0.0.1:5000/ (Press CTRL+C to quit)
```

- Flask also has a huge library support for design and performance along with detailed documentation of each on <u>Flask Libraies</u>. It has no database
- A great emphasis has been put on developing the elegant UI of the website. We have used Brown theme. Through our Website we can predict the power generated by windmill for next 72 hours (on hourly basis) on any user input coordinates and current weather condition.
- Also a database of 50 cities in the India has been fed to the app which can be exploited using "Search for weather condition" button. User can search for the name of the cities and find the power that can be generated there. The app also provides the next 72 hour prediction of weather conditions at that particular coordinate.

### HTML CODE FOR INTRO

```
<html>
        <head>
        <title>Wind Energy Prediction</title>
                .header {
                        top:0px;
                        margin:0px;
                         left: 0px;
                         right: 0px;
                         position: fixed;
                         background: #6c493a;
                         color: white;
                         overflow: hidden;
                         padding-bottom: 30px;
                         font-size: 2.25vw;
                        width: 100%;
                         padding-left:0px;
                        text-align: center;
                         padding-top:20px;
                }
                .second{
                         top:80px;
                        bottom:0px;
                        margin:0px;
                         left: 0px;
                         right: 0px;
                         position: fixed;
                         padding: 0px;
                        width: 100%;
                         background-image:url(https://c1.wallpaperflare.com/preview/623/531/630/596ca965a2e3f.jpg);
                         background-repeat:no-repeat;
                         background-size: contain;
                .inside{
                         top:80px;
                         bottom:0px;
                         margin:0px;
                         left: 45%;
                         right: 0%;
                         position: fixed;
                         padding-left: 40px;
```

```
position: fixed;
                       padding-left: 40px;
                       padding-top:8%;
                       padding-right:40px;
                       background-color: #F2D19A;
                       font-family:Georgia, serif;
                       color:black;
                       font-size:20px;
                       text-align: justify;
               .myButton{
                         border: none;
                         text-align: center;
                         cursor: pointer;
                         text-transform: uppercase;
                         outline: none;
                         overflow: hidden;
                         color: #fff;
                         font-weight: 700;
                         font-size: 12px;
                         background-color: #6c493a;
                         padding: 10px 15px;
                         margin: 0 auto;
                         box-shadow: 0 5px 15px rgba(0,0,0,0.20);
       </style>
       </head>
       <body>
               <div class="header">Predicting The Energy Output Of Wind Turbine Based On Weather Condition</div>
               <div class="second">
                       <div class="inside">Renewable energy, such as wind and solar energy, plays an increasing role in the supply of energy worldwide. This trend will continue because glo
However, levels of production of wind energy are hard to predict as they rely on potentially unstable weather conditions present at the wind farm. In particular, wind speed is crucial for e
               <a href="{{url for('predict')}}"><button type="button" class="myButton" >Want to predict the energy??</button></a>
               </div>
               </div>
       </body>
</html>
```

### Advantages and Disadvantages

#### **ADVANTAGES**

- Weather Underground Services provide very accurate Historical Weather Data which increased the accuracy of model.
- Website is more convenient to use due to zero storage.
- With Choosing city, Website can accurately predict power output using weather condition.

#### <u>DISADVANTAGES</u>

- Weather API is paid and the free version provide limited API requests per day.
- Android Website can be deployed on IBM Cloud.
- No free server available on IBM Cloud for deploying Backend.

### Conclusion

• In this project, we used Weather Underground services (subsidiary of IBM) to get accurate historical weather data. For merging this data with Windmill data we learned some Data Analysis concepts. We analyzed several ML models, and chose Random Tree Regressor to develop this model. This project gave us deep insight about the Flutter framework. We integrated the app with model and Weather API using REST API and Flask Back-end. The accuracy of Random Tree Regressor model for this project is 85%.

### References

Wind Power Data:
 <u>https://www.kaggle.com/berkerisen/wind-turbine-scada-dataset</u>

Weather Data

https://www.wunderground.com

Data Science

https://www.youtube.com/watch?v=CmorAWRsCAw&list=PLeo1K3hjS3uuASpe-1LjfG5f14Bnozjwy

Climacell API

https://www.climacell.co/weather-api/

Flask

https://flask.palletsprojects.com/en/1.1.x/

Flutter

https://flutter.dev/