

IBM Recommendation Model To Utilize Energy From Wind Farm

-BY HOROSCOPER"S

SCOPE:-

- Our **Ultimate Aim** Is To **Competitive Real Time** , **Reduce Energy Wastage** And **Losses In Power Grid**.
- By Using Our Model, we Can Predict The **Valuable Time** Of Power Production.
- **Rest Of Plant** Can Be Provided **Frequently** By **Predicting** The **Unworthy Time** By Forecasting Energy, so We Can **control Machine** Break Down's.
- We Can Solve The **Future Energy Needs**.
- As Well **Availability Of Energy** Can Be Determined.

LITERATUER REVIEW:-

➡️ We get into this great paper to gather information about this domain.

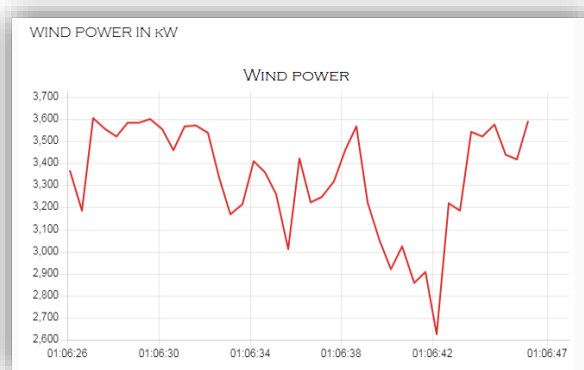
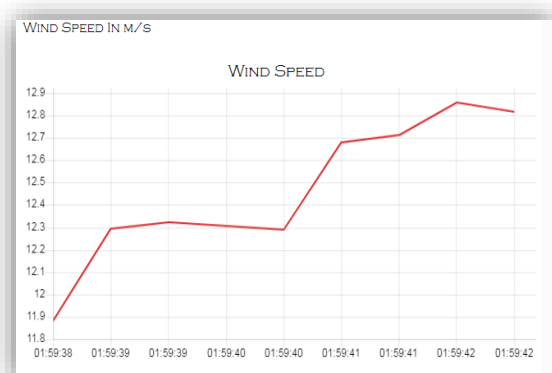
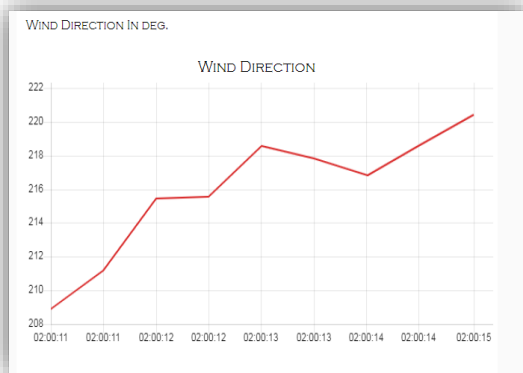
➡️ Below mention's are our aspects from this paper.

Article A New Hybrid Approach to Forecast Wind Power for Large Scale Wind Turbine Data Using Deep Learning with TensorFlow Framework and Principal Component Analysis

-Mansoor Khan , Tianqi Liu ,* and Farhan Ullah

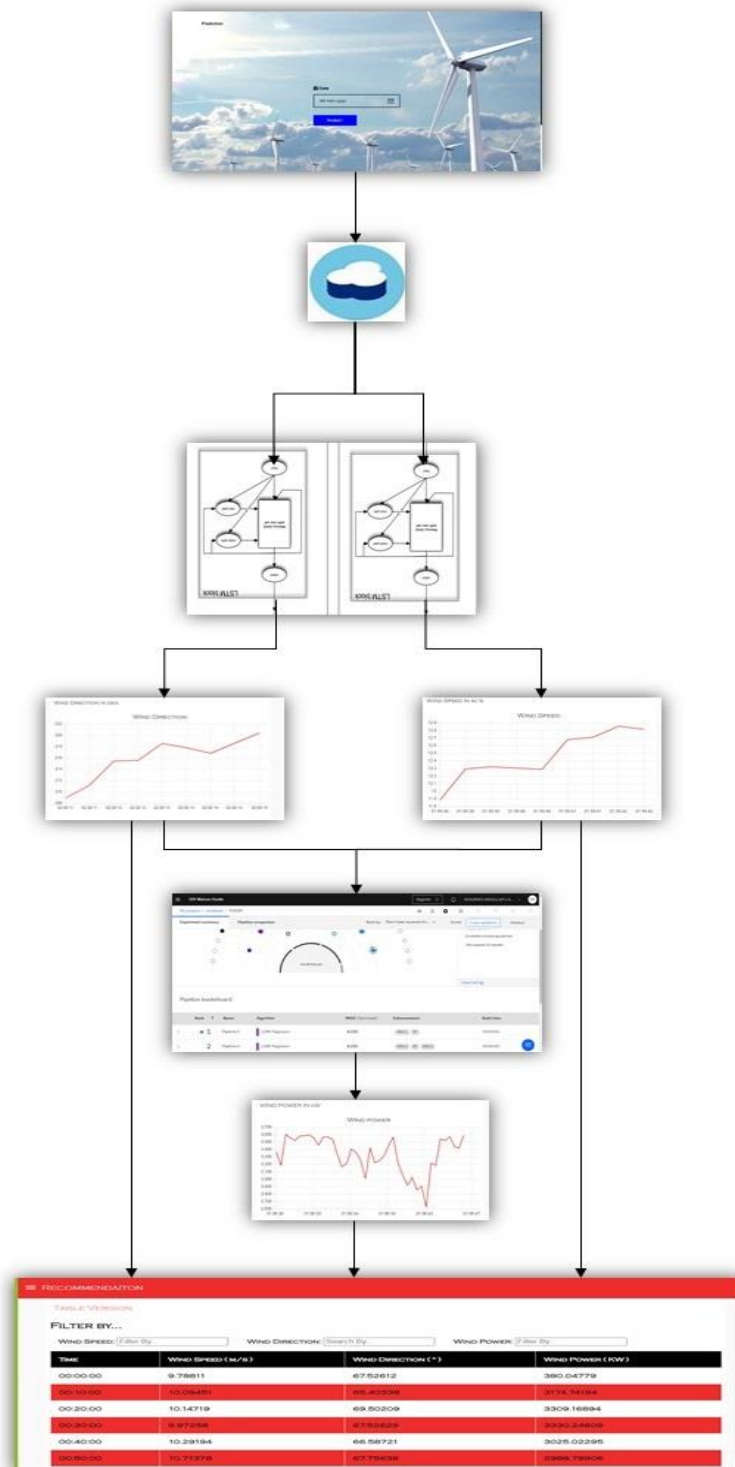
- The principal components are then used as input data to **deep learning** based on the TensorFlow framework. **PCA provides feature extraction and selection.**
- Here, **acc, val_loss, val_acc** represents **accuracy**, validation loss and validation accuracy, respectively. The loss, accuracy and loss, and **validation loss** are **calculated for wind power forecasting.**
- The proposed deep learning algorithm is applied to PCs **to forecast wind power.** The Keras API is used with **TensorFlow to configure** a more reliable **neural network.**

Forecasted Output Of Our Model:-



The Above Forecasting Done For 24 Hrs On Date 28-10-2021

Architecture Flow



Work Process:-

✈ IBM Account Creation.

- Smart Internz assigning template.
- Assigning team.

✈ Fetching Input.

✈ Selection of algorithm.

✈ Data preparation.

✈ Deploying Model.

✈ GUI creation.

Fetching Input:-

➔ Collecting of data from various website.

- Kaggle-

<https://www.kaggle.com/berkerisen/wind-turbine-scada-dataset>

- Github-https://github.com/ShashwatArghode/Wind-Energy-Prediction-using-LSTM/blob/master/AL_WIND_07_12.xlsx

➔ Importing data in **IBM Watson Studio** in a click.

```
In [10]: import types
import pandas as pd
from botocore.client import Config
import ibm_botocore

def __iter__(self): return #

# @hidden_cell
# The following code accesses a file in your IBM Cloud Object Storage. It includes your credentials.
# You might want to remove those credentials before you share the notebook.
client_54c8547f35d43b5d41d68c871da81d6 = ibm_botocore.client(service_name='s3',
    ibm_api_key_id='WcULNdvsl13y128P9SQIeHvJ5Ty-JYw0s5hpA3f1',
    ibm_auth_endpoint='https://iam.cloud.ibm.com/oidc/token',
    config=Config(signature_version='auth'),
    endpoint_url='https://s3.eu-gov-objectstorage.service.networklayer.com')

body = client_54c8547f35d43b5d41d68c871da81d6.get_object(Bucket='hanif-donotdelete-pr-nx1leycl0bvpq',Key='winddirtime.csv')['Body']
# add missing __iter__ method, so pandas accepts body as file-like object
if not hasattr(body, '__iter__'): body.__iter__ = types.MethodType(__iter__, body)

series = pd.read_csv(body,index_col='Date/Time')
series.head()
```

Out[10]:

| Date/Time | Wind Direction (°) |
|------------------|--------------------|
| 01 01 2018 00:00 | 259.994904 |
| 01 01 2018 00:10 | 268.641113 |
| 01 01 2018 00:20 | 272.564789 |
| 01 01 2018 00:30 | 271.250067 |
| 01 01 2018 00:40 | 265.674286 |

Selection Of Algorithm:-

- Choosing **Best algo** . is the **main** process ever in **forecasting**.
- Normally algo. Like **SVM, Arima, Auto Arima** are used for **time series** prediction.
- But in this recommendation model **we tried LSTM** algo, which is one of library of **keras**.
- We have done this **forecasting** in **three segment**, Wind Speed, Direction, Power.
- **Wind Speed** and **Wind Direction** using **Istm** model.
- **Power Forecasting** using **Auto Ai**, the facility provided by our **IBM Watson studio**.

Feature Engineering :-

▶ FOR WIND SPEED AND DIRECTION:-

- To fit our model we done some **preprocessing**,
- Split up data into **train** and **test**.
- Converting the data into **scalar form**, then take **difference of past data**.
- Setting the **epoch**, finding **study rate**, fix the **batch size**, selection of **neurons** and **validate**.

▶ FOR POWER:-

- We done it in **auto AI**.
- For **power** prediction we give the **input** from the **output** of former prediction.

```
SPLITTING OF DATA

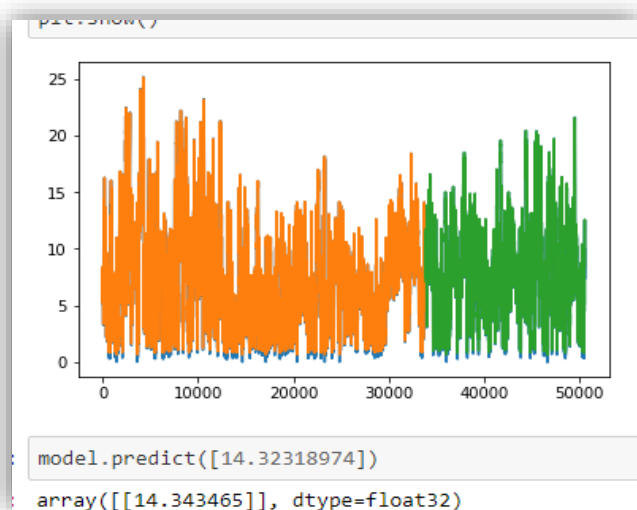
[8]: X = df[['WindSpeed','WindDirection']].values
      print(X)
      y = df['ActivePower'].values
      print(y)
      from sklearn.model_selection import train_test_split

      x_train,x_test,y_train,y_test = train_test_split(X,y, test_size = 0.2)

      [[ 5.31133604 259.99490356]
       [ 5.67216682 268.64111328]
       [ 5.2160368 272.56478882]
       ...
       [ 8.43535805 84.74250031]
       [ 9.42136574 84.2979126 ]
       [ 9.97933197 82.27462006]]
      [ 380.04779053 453.76919556 306.37658691 ... 2201.10693359 2515.6940918
       2820.46606445]
```

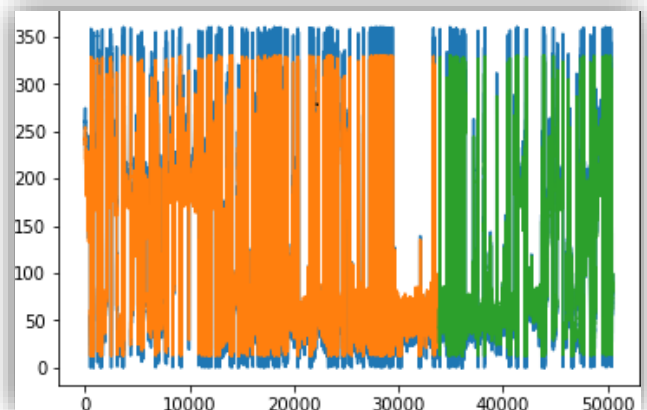
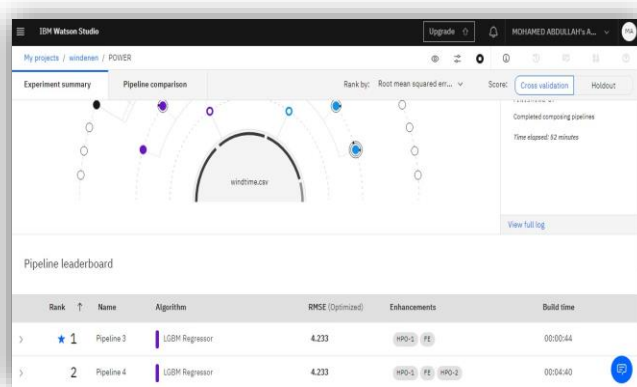

Validation of our Prediction:-

- ➡ The plot shows the **test and train data set fitted with our model.**
- ➡ The RMSE value showing below are **model fitted by power** data which is especially done by **Auto Ai Service** provided by Our **IBM Watson Studio.**



```
import math
# Estimate model performance
trainScore = model.evaluate(trainX, trainY, verbose=0)
print('Train Score: %.2f MSE (%.2f RMSE)' % (trainScore, math.sqrt(trainScore)))
testScore = model.evaluate(testX, testY, verbose=0)
print('Test Score: %.2f MSE (%.2f RMSE)' % (testScore, math.sqrt(testScore)))
```

Train Score: 0.58 MSE (0.76 RMSE)
Test Score: 0.54 MSE (0.74 RMSE)



Deployment:-

- **IBM Watson Studio** made **Deployment** easy in few steps.
- For deployment purpose, we created a service credentials in **Watson machine learning service**.
- After creating credentials **load the model** in service Provider ,We **get the guid**
- that's all its done

```
from watson_machine_learning_client import WatsonMachineLearningAPIClient

wml_credentials = {
    "apikey": "wMfKx0hgmD2Aa00uF_Tug209WwmdJ55cde-x6Tg",
    "iam_apikey_description": "Auto-generated for key 22942c5d-09db-412c-aaab-370000375868",
    "iam_apikey_name": "Service-credentials-2",
    "iam_role_crn": "crn:ibm:iam:public:iam::service-role-writer",
    "iam_serviceid_crn": "crn:ibm:iam:public:iam-identity:a1af47091ccdc47489f9053b9340a3183::serviceid:serviceid-a2a0f9d-79bc-4f1b-b925-799288ee2c1a",
    "instance_id": "f4d6cf6b-45e5-4958-af08-d343b7274000",
    "url": "https://eu-gb.ml.cloud.ibm.com"
}

client = WatsonMachineLearningAPIClient(wml_credentials)

metadata = {
    client.repository.ModelMetadataNames.AUTHOR_NAME: 'Muhammed Abdullah',
    client.repository.ModelMetadataNames.AUTHOR_EMAIL: 'abdullahmuhammed214@gmail.com',
    client.repository.ModelMetadataNames.NAME: 'Prediction of kind Direction',
    client.repository.ModelMetadataNames.FRAMEWORK_NAME: 'tensorflow',
    client.repository.ModelMetadataNames.FRAMEWORK_VERSION: '1.15',
    client.repository.ModelMetadataNames.FRAMEWORK_LIBRARIES: [{"name": 'keras', 'version': '2.2.4'}]
}

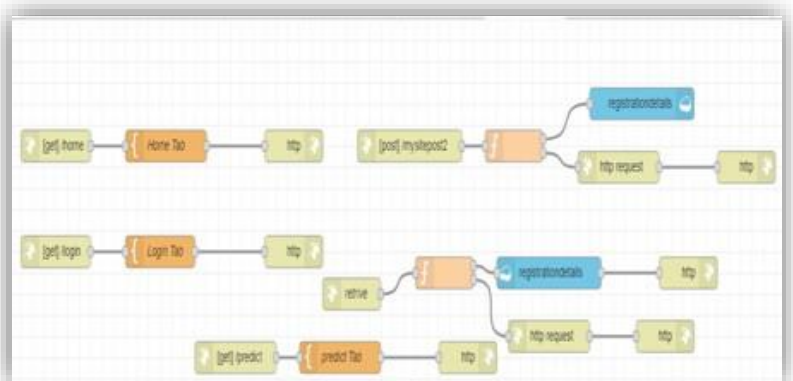
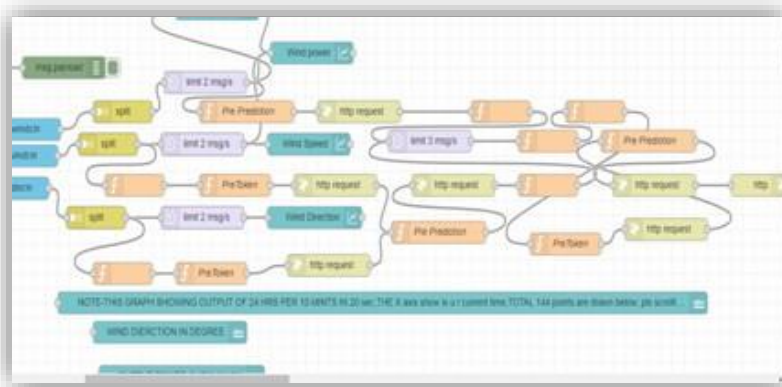
model_result_path = 'direction_prediction.h5'
lstm_model.save(model_result_path)

!tar -czvf direction_prediction.tgz direction_prediction.h5
direction_prediction.h5

stored_data = client.repository.store_model(model='direction_prediction.tgz', meta_props=metadata, training_data=X, training_target=y)
```

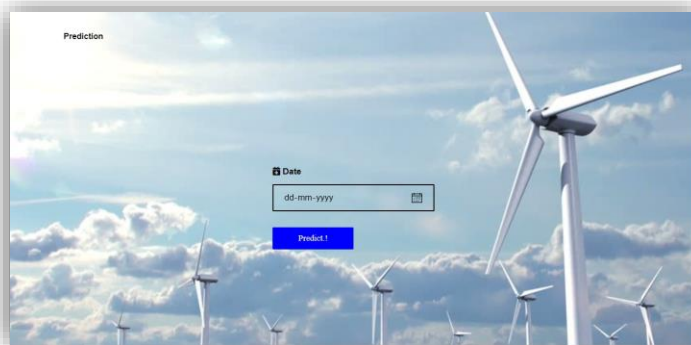
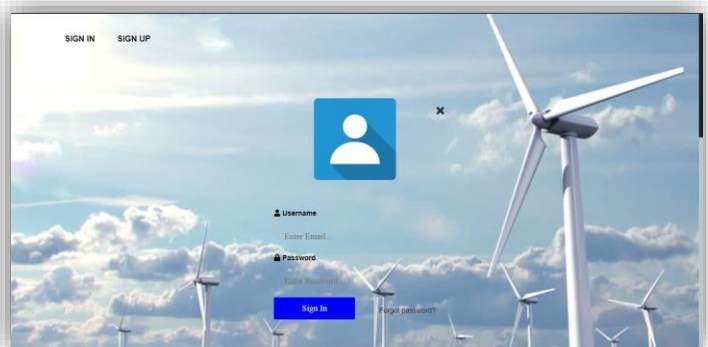
GUI Creation:-

- We Done Our GUI With **Node-red**, it Is One Of **Service** Provide By **Our IBM**
- Below Shown Node Are Our **Prediction Node** And Our **GUI Tab Nodes**



Our GUI With IBM

👉 below shown image' s are our different tabs in our GUI.



HIGHLIGHT' S: -

- We Predicting **Wind Speed, Wind Direction** In **separate algo.**, and giving the **output of this prediction** to predict **power**, however real time scenario happens like this, so our model should be best to **competitive with real time.**
- It can also **recommend** the time to **Utilize Power from grid.**
- So that we can give **rest period** to grid and **Save Energy.**
- **24 hours** forecast with **10 minutes interval**, just in **20 seconds.**
- For easy understanding of users “ **Recommended Table Version** ” are provided.
- To view forecast in better way, **Search** and **Filter** options are added.

REPORT ON -SBSPS_Challenge_2238-IBM Recommendation Model To Utilize Energy From Wind

RECOMMENDATION

TABLE VERSION

FILTER BY...

WIND SPEED: WIND DIRECTION: WIND POWER:

| TIME | WIND SPEED (M/S) | WIND DIRECTION (°) | WIND POWER (KW) |
|----------|--------------------|----------------------|-------------------|
| 00:00:00 | 9.78811 | 67.52612 | 380.04779 |
| 00:10:00 | 10.09451 | 65.40338 | 3174.74194 |
| 00:20:00 | 10.14719 | 69.50209 | 3309.16894 |
| 00:30:00 | 9.97258 | 67.52629 | 3330.24609 |
| 00:40:00 | 10.29194 | 66.58721 | 3025.02295 |
| 00:50:00 | 10.71378 | 67.75638 | 2988.78906 |
| ----- | ----- | ----- | ----- |

RECOMMENDATION

| TIME | WIND SPEED (M/S) | WIND DIRECTION (°) | WIND POWER (KW) |
|----------|--------------------|----------------------|-------------------|
| 00:00:00 | 9.78811 | 67.52612 | 380.04779 |
| 00:10:00 | 10.09451 | 65.40338 | 3174.74194 |
| 00:20:00 | 10.14719 | 69.50209 | 3309.16894 |
| 00:30:00 | 9.97258 | 67.52629 | 3330.24609 |
| 00:40:00 | 10.29194 | 66.58721 | 3025.02295 |
| 00:50:00 | 10.71378 | 67.75638 | 2988.78906 |
| 01:00:00 | 10.51755 | 65.80247 | 1261.66296 |
| 01:10:00 | 10.3856 | 66.7978 | 1307.474 |
| 01:20:00 | 10.22741 | 64.7504 | 1483.552 |
| | | | |

Future Plans: -

- ▶ To give **chart bots** to **help more** for the service users.
- ▶ **Prediction** of **weather** also help to maintain **more stability** in prediction.
- ▶ All **screen compatibility** will be launched soon.

To Access our GUI:-

➤ <https://node-red-itpxr.eu-gb.mybluemix.net/home>

To Appreciate Our Work:-

➤ <https://node-red-itpxr.eu-gb.mybluemix.net/home#contact>

**SPECIAL THANKS
TO: -**

IBM Platform And SMARTINTERNZ

by:-

Team -Horoscooper' s

MOHAMED HANEEF I

MOHAMED ABDULLAH K