IBM-Recommendation-Model-To-Utilize-Energy-From-Wind-Farm

SCOPE

- ➤OUR ULIMATE AIM IS TO COMPETITIVE REAL TIME, REDUCE ENERGY WASTEAGE 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 FORECATSING ENERGY, WE CAN SOLVE THE FUTURE ENERGY NEEDS.
- > AS WELL AVAILBIALITY OF ENERGY CAN BE DETERMINED.

LITERATUER REVIEW:-

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

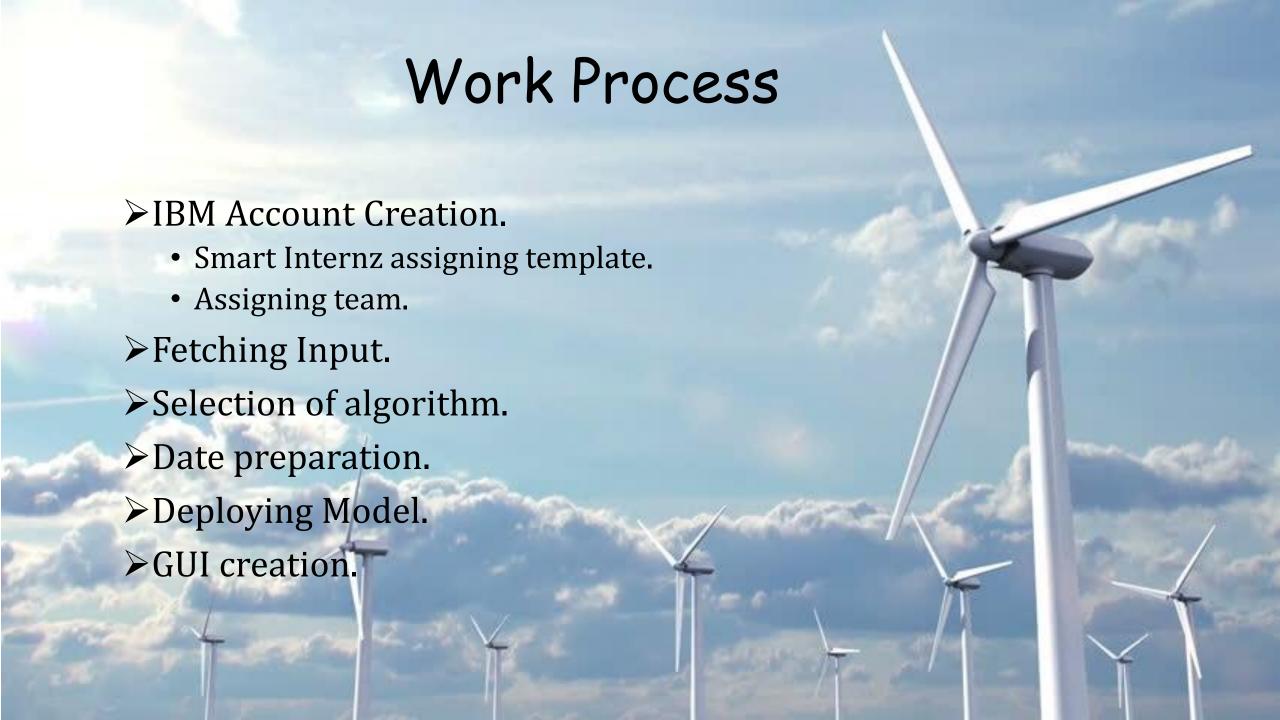
- The principal components are then used as input data to **deep learning** based on the **TensorFlow framework**. PCA provides feature extraction and selection.
- ▶ Here, acc, val_loss, val_acc represents accuracy, validation loss and validation accuracy, respectively. The loss, accuracy and loss, and 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.

Forecasting output









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 watson studion in a click.

```
import pandas as pd
from botocore.client import Config
def iter (self): return 0
client 64c85547f35d43b5b41b68c871da81d6 = ibm_boto3.client(service_name='s3',
    ibm api key id='HcWLNSvsllJ6yiZ0P9bQiEHBvJ5Ty-JY00oShzpA1FL1'
    ibm auth endpoint="https://iam.cloud.ibm.com/oidc/token",
    config=Config(signature_version='oauth'),
    endpoint url='https://s3.eu-geo.objectstorage.service.networklayer.com')
body = client_64c85547f35d43b5b41b68c871da81d6.get_object(Bucket='hanif-donotdelete-pr-nx1leyc1obvpqp',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()
                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 00:30 271.258087
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 lstm model.
- ➤ Power Forescating using **AutoAi**, The facility provied by **our IBM Watson studio**.

Future Engineering

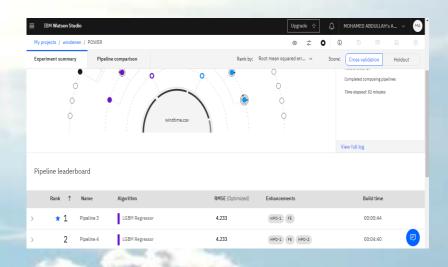
FOR WIND SPEED AND DIRECTION:-

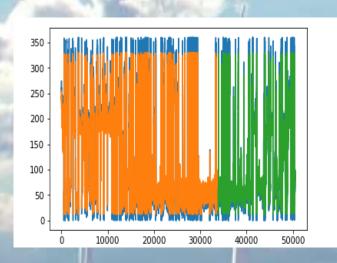
- To fit our model we done some **preprocessing**.
- >Split up data into train and test.
- > Converting the data into scalar form, then take difference of past data.
- Setting the epoch ,finding study rate, fix the batch size, selection of neurons and validate.

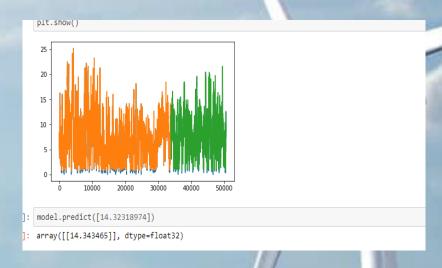
FOR POWER:-

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

Validation of our Prediction







import math

Estimate model performance

trainScore = model.evaluate(trainX, trainY, verbose=@
print('Train Score: %.2f MSE (%.2f RMSE)' % (trainSco
testScore = model.evaluate(testX, testY, verbose=0)
print('Test Score: %.2f MSE (%.2f RMSE)' % (testScore)

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

Deployment

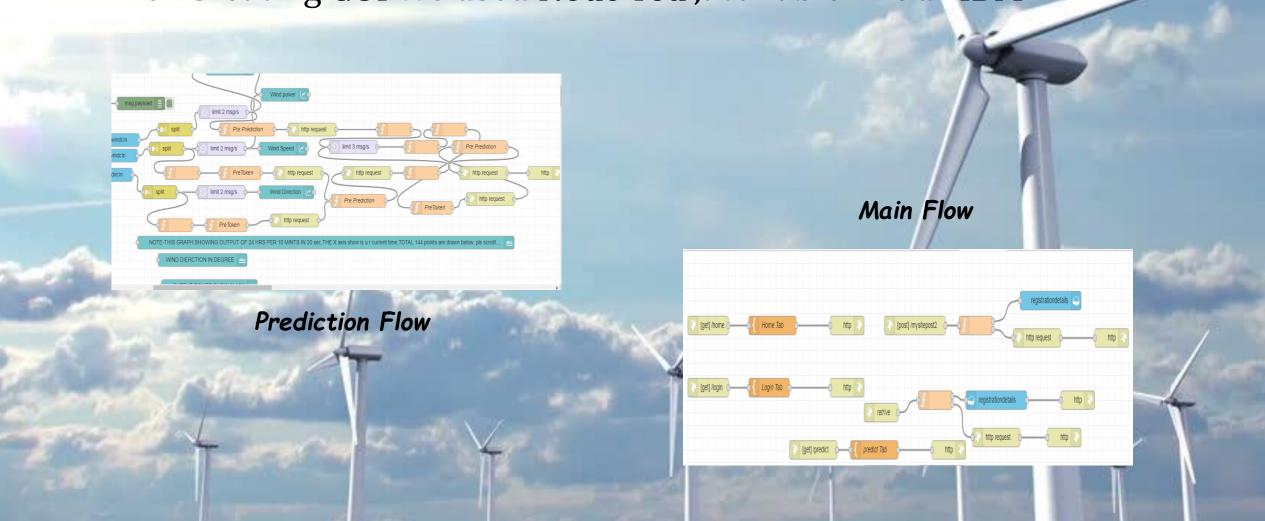
- ➤ IBM Watson Studio made Deployment easy in few steps.
- For deployment purpose, we created a service credentials in Watson machine learning service.

After creating credentials load the model in service provider, We get the guid.

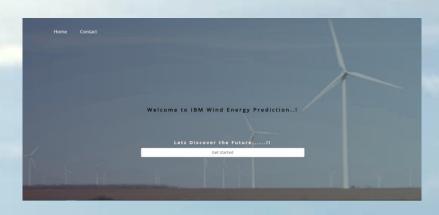
```
from watson machine learning client import WatsonMachineLearningAPIClient
wml credentials = {
  apikey": "mMtFox3bgn8ZwUo80xuT 7uegZO6HNwmdj55cde-x4Tg",
  "iam_apikey_description": "Auto-generated for key 22942c5d-09db-432c-aaa8-3769965759b9",
  "iam apikey name": "Service credentials-2",
  "iam role crn": "crn:v1:bluemix:public:iam::::serviceRole:Writer".
  "iam serviceid crn": "crn:v1:bluemix:public:iam-identity::a/1af47d91ccdc47489f9955bf340a1583::serviceid:ServiceId-a1a64fbd-7e6c-4f1b-9e
  "instance_id": "fdc0cfe6-e3a5-4958-af00-d1453d727e98",
   url": "https://eu-gb.ml.cloud.ibm.com
client = WatsonMachineLearningAPIClient(wml_credentials)
       client.repository.ModelMetaNames.AUTHOR NAME: 'Mohamed Abdullah',
       client.repository.ModelMetaNames.AUTHOR_EMAIL: 'abdullahmohamed2540@gmail.com',
        client.repository.ModelMetaNames.NAME: 'Prediction of Wind Direction',
        client.repository.ModelMetaNames.FRAMEWORK_NAME: 'tensorflow',
        client.repository.ModelMetaNames.FRAMEWORK VERSION: '1.15'.
        client.repository.ModelMetaNames.FRAMEWORK_LIBRARIES:[{'name':'keras', 'version':'2.2.4'}]
model_result_path = 'direction_prediction.h5'
1stm model.save(model result path)
!tar -zcvf 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

For Creating GUI We used Node-red, available in Our IBM.



Our GUI With IBM



Home Tab





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.



