

```
In [1]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
import joblib
```

```
In [2]: import os, types
import pandas as pd
from boto3.client import Config
import ibm_boto3

def __iter__(self): return 0

# @hidden_cell
# The following code accesses a file in your IBM Cloud Object Storage. It in
# You might want to remove those credentials before you share the notebook.
cos_client = ibm_boto3.client(service_name='s3',
                              ibm_api_key_id='0j9ybHNI1hQ6b2L9ZiJ1LBRjwWf3tbtFoNAoTtZLrz9z',
                              ibm_auth_endpoint="https://iam.cloud.ibm.com/oidc/token",
                              config=Config(signature_version='oauth'),
                              endpoint_url='https://s3.private.us.cloud-object-storage.appdomain.cloud')

bucket = 'predictingenergyoutputofwindturbi-donotdelete-pr-vvuc1mxciqx0o2'
object_key = 'T1.csv'

body = cos_client.get_object(Bucket=bucket,Key=object_key)['Body']
# add missing __iter__ method, so pandas accepts body as file-like object
if not hasattr(body, "__iter__"): body.__iter__ = types.MethodType( __iter__

df = pd.read_csv(body)
df.head()
```

```
Out[2]:
```

	Date/Time	LV ActivePower (kW)	Wind Speed (m/s)	Theoretical_Power_Curve (KWh)	Wind Direction (°)
0	01 01 2018 00:00	380.047791	5.311336	416.328908	259.994904
1	01 01 2018 00:10	453.769196	5.672167	519.917511	268.641113
2	01 01 2018 00:20	306.376587	5.216037	390.900016	272.564789
3	01 01 2018 00:30	419.645905	5.659674	516.127569	271.258087
4	01 01 2018 00:40	380.650696	5.577941	491.702972	265.674286

```
In [3]: 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
```

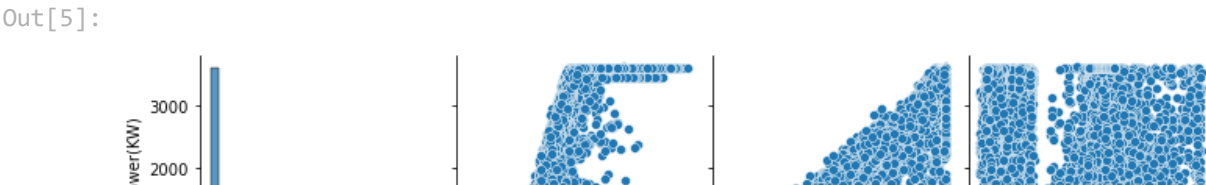
Out[4]:

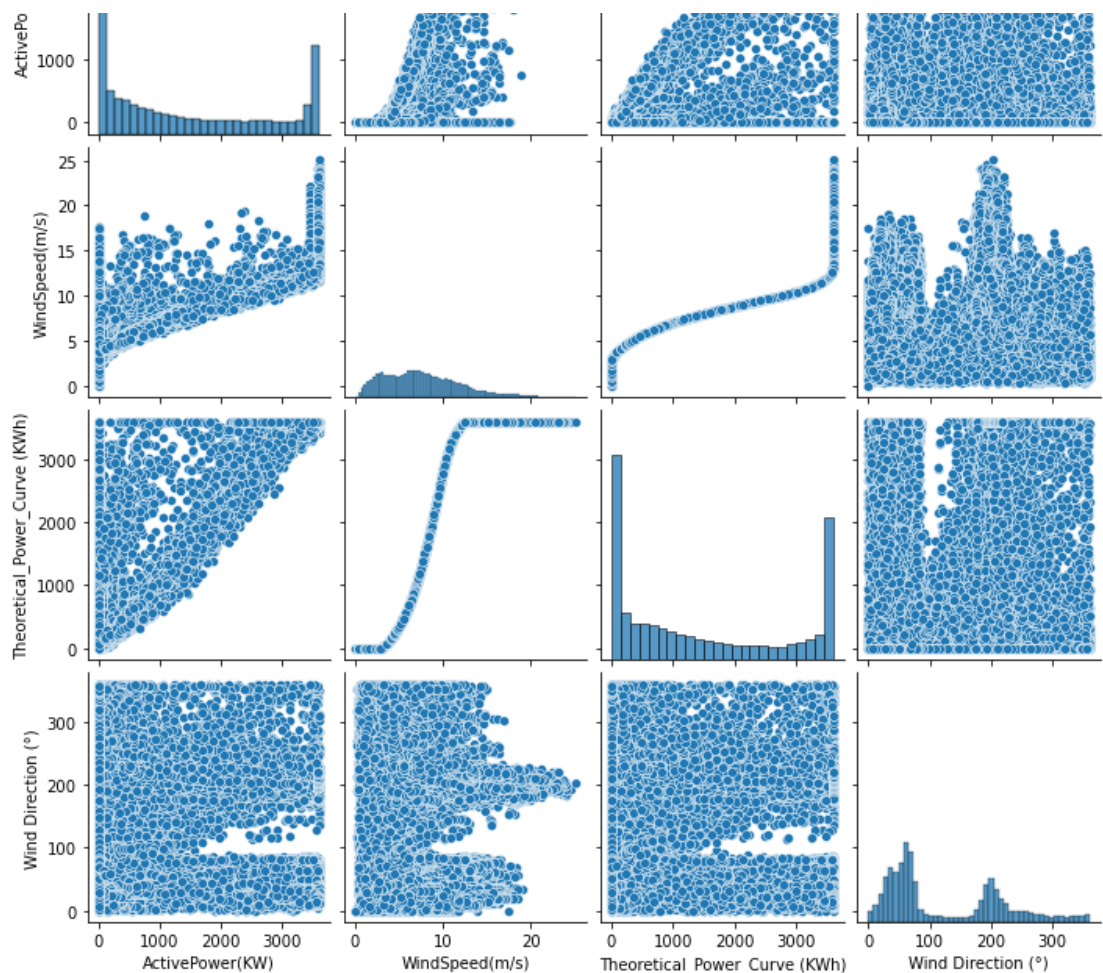
	Time	ActivePower(KW)	WindSpeed(m/s)	Theoretical_Power_Curve (KWh)	Wind Direction (°)
0	01	380.047791	5.311336	416.328908	259.994904
	01				
	2018 00:00				
1	01	453.769196	5.672167	519.917511	268.641113
	01				
	2018 00:10				
2	01	306.376587	5.216037	390.900016	272.564789
	01				
	2018 00:20				
3	01	419.645905	5.659674	516.127569	271.258087
	01				
	2018 00:30				
4	01	380.650696	5.577941	491.702972	265.674286
	01				
	2018 00:40				
...
50525	31	2963.980957	11.404030	3397.190793	80.502724
	12				
	2018 23:10				
50526	31	1684.353027	7.332648	1173.055771	84.062599
	12				
	2018 23:20				
50527	31	2201.106934	8.435358	1788.284755	84.742500
	12				
	2018 23:30				
50528	31	2515.694092	9.421366	2418.382503	84.297913
	12				
	2018 23:40				
50529	31	2820.466064	9.979332	2779.184096	82.274620
	12				
	2018 23:50				

50530 rows × 5 columns

In [5]:

sns.pairplot(df)





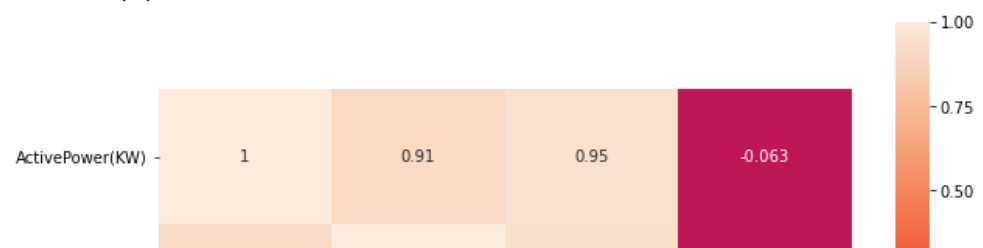
```
In [6]: plt.figure(figsize=(10, 8))
corr = df.corr()
ax = sns.heatmap(corr, vmin = -1,vmax = 1,annot = True)
bottom, top = ax.get_ylim()
ax.set_ylim(bottom + 0.5, top - 0.5)

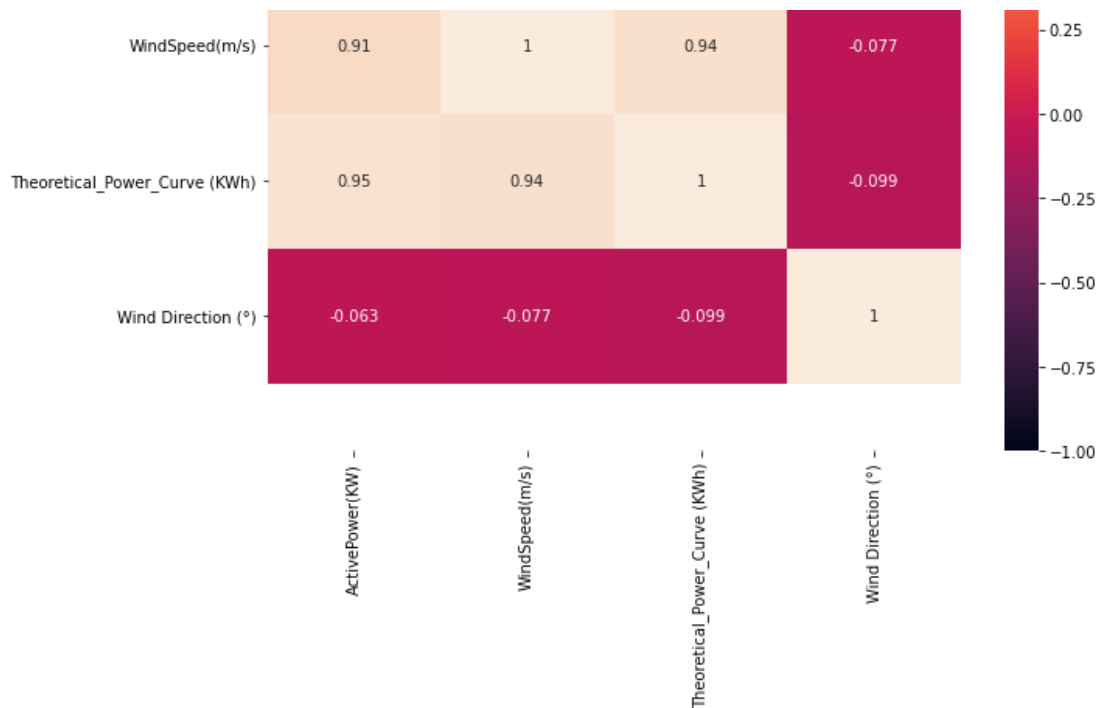
print(corr)
```

	ActivePower(KW)	WindSpeed(m/s)	\
ActivePower(KW)	1.000000	0.912774	
WindSpeed(m/s)	0.912774	1.000000	
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	





```
In [7]: df["Time"] = pd.to_datetime(df["Time"], format = "%d %m %Y %H %M", errors =
```

```
In [8]: y = df["ActivePower(KW)"]
X = df[["Theoretical_Power_Curve (KWh)", "WindSpeed(m/s)"]]

from sklearn.model_selection import train_test_split
train_X, val_X, train_y, val_y = train_test_split(X, y, random_state=0)
```

Model building

```
In [9]: from sklearn.ensemble import RandomForestRegressor
from sklearn.metrics import mean_absolute_error, r2_score
from xgboost import XGBRegressor

forest_model = RandomForestRegressor(n_estimators = 750, max_depth = 4, max_
forest_model.fit(train_X, train_y)
```

```
Out[9]: RandomForestRegressor(max_depth=4, max_leaf_nodes=500, n_estimators=750,
random_state=1)
```

```
In [10]: RandomForestRegressor(max_depth=4, max_leaf_nodes=500, n_estimators=750,
random_state=1)
```

```
Out[10]: RandomForestRegressor(max_depth=4, max_leaf_nodes=500, n_estimators=750,
random_state=1)
```

```
In [11]: power_preds = forest_model.predict(val_X)

print(mean_absolute_error(val_y, power_preds))
print(r2_score(val_y, power_preds))
```

```
164.58015525861344
0.9113496428907649
```

```
In [12]: joblib.dump(forest_model, "power_prediction.sav")
```

```
Out[12]: ['power_prediction.sav']
```

```
In [13]: df
```

```
Out[13]:
```

	Time	ActivePower(KW)	WindSpeed(m/s)	Theoretical_Power_Curve (KWh)	Wind Direction (°)
0	NaT	380.047791	5.311336	416.328908	259.994904
1	NaT	453.769196	5.672167	519.917511	268.641113
2	NaT	306.376587	5.216037	390.900016	272.564789
3	NaT	419.645905	5.659674	516.127569	271.258087
4	NaT	380.650696	5.577941	491.702972	265.674286
...
50525	NaT	2963.980957	11.404030	3397.190793	80.502724
50526	NaT	1684.353027	7.332648	1173.055771	84.062599
50527	NaT	2201.106934	8.435358	1788.284755	84.742500
50528	NaT	2515.694092	9.421366	2418.382503	84.297913
50529	NaT	2820.466064	9.979332	2779.184096	82.274620

50530 rows × 5 columns

```
In [14]: import pickle
pickle.dump(forest_model, open("model.pkl", "wb"))
```

```
In [15]: !pip install -U ibm-watson-machine-learning
```

Requirement already satisfied: ibm-watson-machine-learning in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (1.0.257)
Requirement already satisfied: tabulate in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from ibm-watson-machine-learning) (0.8.9)
Requirement already satisfied: lomond in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from ibm-watson-machine-learning) (0.3.3)
Requirement already satisfied: urllib3 in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from ibm-watson-machine-learning) (1.26.7)
Requirement already satisfied: requests in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from ibm-watson-machine-learning) (2.26.0)
Requirement already satisfied: packaging in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from ibm-watson-machine-learning) (21.3)
Requirement already satisfied: pandas<1.5.0,>=0.24.2 in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from ibm-watson-machine-learning) (1.3.4)
Requirement already satisfied: ibm-cos-sdk==2.11.* in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from ibm-watson-machine-learning) (2.11.0)
Requirement already satisfied: importlib-metadata in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from ibm-watson-machine-learning) (4.8.2)
Requirement already satisfied: certifi in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from ibm-watson-machine-learning) (2022.9.24)
Requirement already satisfied: jmespath<1.0.0,>=0.7.1 in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from ibm-watson-machine-learning) (0.10.0)

hon-3.9/lib/python3.9/site-packages (from ibm-cos-sdk==2.11.*->ibm-watson-machine-learning) (0.10.0)
Requirement already satisfied: ibm-cos-sdk-core==2.11.0 in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from ibm-cos-sdk==2.11.*->ibm-watson-machine-learning) (2.11.0)
Requirement already satisfied: ibm-cos-sdk-s3transfer==2.11.0 in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from ibm-cos-sdk==2.11.*->ibm-watson-machine-learning) (2.11.0)
Requirement already satisfied: python-dateutil<3.0.0,>=2.1 in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from ibm-cos-sdk-core==2.11.0->ibm-cos-sdk==2.11.*->ibm-watson-machine-learning) (2.8.2)
Requirement already satisfied: pytz>=2017.3 in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from pandas<1.5.0,>=0.24.2->ibm-watson-machine-learning) (2021.3)
Requirement already satisfied: numpy>=1.17.3 in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from pandas<1.5.0,>=0.24.2->ibm-watson-machine-learning) (1.20.3)
Requirement already satisfied: six>=1.5 in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from python-dateutil<3.0.0,>=2.1->ibm-cos-sdk-core==2.11.0->ibm-cos-sdk==2.11.*->ibm-watson-machine-learning) (1.15.0)
Requirement already satisfied: idna<4,>=2.5 in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from requests->ibm-watson-machine-learning) (3.3)
Requirement already satisfied: charset-normalizer~=2.0.0 in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from requests->ibm-watson-machine-learning) (2.0.4)
Requirement already satisfied: zipp>=0.5 in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from importlib-metadata->ibm-watson-machine-learning) (3.6.0)
Requirement already satisfied: pyparsing!=3.0.5,>=2.0.2 in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from packaging->ibm-watson-machine-learning) (3.0.4)

```
In [16]: from ibm_watson_machine_learning import APIClient
import json
import numpy as np
```

```
In [17]: wml_credentials = {
          "apikey": "Ilik-kKvZ4Lwtruh-17B12FS5IEFz0Ujhq8533qqw09Z",
          "url": "https://us-south.ml.cloud.ibm.com"
        }
```

```
In [18]: wml_client = APIClient(wml_credentials)
```

```
In [19]: wml_client.spaces.list()
```

Note: 'limit' is not provided. Only first 50 records will be displayed if the number of records exceed 50

ID	NAME	CREATED
f1f8ff94-56cd-4e00-a0c9-f8ae5161c050	wind turbine	2022-11-15T04:21:43.273Z

```
In [20]: SPACE_ID = "f1f8ff94-56cd-4e00-a0c9-f8ae5161c050"
```

```
In [21]: wml_client.set.default_space(SPACE_ID)
```

```
Out[21]: 'SUCCESS'
```

```
In [22]: wml_client.software_specifications.list()
```

NAME	ASSET_ID	TYPE
default_py3.6	0062b8c9-8b7d-44a0-a9b9-46c416adcbdb9	base
kernel-spark3.2-scala2.12	020d69ce-7ac1-5e68-ac1a-31189867356a	base
pytorch-onnx_1.3-py3.7-edt	069ea134-3346-5748-b513-49120e15d288	base
scikit-learn_0.20-py3.6	09c5a1d0-9c1e-4473-a344-eb7b665ff687	base
spark-mllib_3.0-scala_2.12	09f4cfff-90a7-5899-b9ed-1ef348aebdee	base
pytorch-onnx_rt22.1-py3.9	0b848dd4-e681-5599-be41-b5f6fccc6471	base
ai-function_0.1-py3.6	0cdb0f1e-5376-4f4d-92dd-da3b69aa9bda	base
shiny-r3.6	0e6e79df-875e-4f24-8ae9-62dcc2148306	base
tensorflow_2.4-py3.7-horovod	1092590a-307d-563d-9b62-4eb7d64b3f22	base
pytorch_1.1-py3.6	10ac12d6-6b30-4ccd-8392-3e922c096a92	base
tensorflow_1.15-py3.6-ddl	111e41b3-de2d-5422-a4d6-bf776828c4b7	base
autoai-kb_rt22.2-py3.10	125b6d9a-5b1f-5e8d-972a-b251688ccf40	base
runtime-22.1-py3.9	12b83a17-24d8-5082-900f-0ab31fbfd3cb	base
scikit-learn_0.22-py3.6	154010fa-5b3b-4ac1-82af-4d5ee5abbc85	base
default_r3.6	1b70aec3-ab34-4b87-8aa0-a4a3c8296a36	base
pytorch-onnx_1.3-py3.6	1bc6029a-cc97-56da-b8e0-39c3880dbbe7	base
kernel-spark3.3-r3.6	1c9e5454-f216-59dd-a20e-474a5cdf5988	base
pytorch-onnx_rt22.1-py3.9-edt	1d362186-7ad5-5b59-8b6c-9d0880bde37f	base
tensorflow_2.1-py3.6	1eb25b84-d6ed-5dde-b6a5-3fbdf1665666	base
spark-mllib_3.2	20047f72-0a98-58c7-9ff5-a77b012eb8f5	base
tensorflow_2.4-py3.8-horovod	217c16f6-178f-56bf-824a-b19f20564c49	base
runtime-22.1-py3.9-cuda	26215f05-08c3-5a41-a1b0-da66306ce658	base
do_py3.8	295addb5-9ef9-547e-9bf4-92ae3563e720	base
autoai-ts_3.8-py3.8	2aa0c932-798f-5ae9-abd6-15e0c2402fb5	base
tensorflow_1.15-py3.6	2b73a275-7cbf-420b-a912-eae7f436e0bc	base
kernel-spark3.3-py3.9	2b7961e2-e3b1-5a8c-a491-482c8368839a	base
pytorch_1.2-py3.6	2c8ef57d-2687-4b7d-acce-01f94976dac1	base
spark-mllib_2.3	2e51f700-bca0-4b0d-88dc-5c6791338875	base
pytorch-onnx_1.1-py3.6-edt	32983cea-3f32-4400-8965-dde874a8d67e	base
spark-mllib_3.0-py37	36507ebe-8770-55ba-ab2a-eafe787600e9	base
spark-mllib_2.4	390d21f8-e58b-4fac-9c55-d7ceda621326	base
autoai-ts_rt22.2-py3.10	396b2e83-0953-5b86-9a55-7ce1628a406f	base
xgboost_0.82-py3.6	39e31acd-5f30-41dc-ae44-60233c80306e	base
pytorch-onnx_1.2-py3.6-edt	40589d0e-7019-4e28-8daa-fb03b6f4fe12	base
pytorch-onnx_rt22.2-py3.10	40e73f55-783a-5535-b3fa-0c8b94291431	base
default_r36py38	41c247d3-45f8-5a71-b065-8580229facf0	base
autoai-ts_rt22.1-py3.9	4269d26e-07ba-5d40-8f66-2d495b0c71f7	base
autoai-obm_3.0	42b92e18-d9ab-567f-988a-4240ba1ed5f7	base
pmml-3.0_4.3	493bcb95-16f1-5bc5-bee8-81b8af80e9c7	base
spark-mllib_2.4-r_3.6	49403dff-92e9-4c87-a3d7-a42d0021c095	base
xgboost_0.90-py3.6	4ff8d6c2-1343-4c18-85e1-689c965304d3	base
pytorch-onnx_1.1-py3.6	50f95b2a-bc16-43bb-bc94-b0bed208c60b	base
autoai-ts_3.9-py3.8	52c57136-80fa-572e-8728-a5e7cbb42cde	base
spark-mllib_2.4-scala_2.11	55a70f99-7320-4be5-9fb9-9edb5a443af5	base
spark-mllib_3.0	5c1b0ca2-4977-5c2e-9439-ffd44ea8ffe9	base
autoai-obm_2.0	5c2e37fa-80b8-5e77-840f-d912469614ee	base
spss-modeler_18.1	5c3cad7e-507f-4b2a-a9a3-ab53a21dee8b	base
cuda-py3.8	5d3232bf-c86b-5df4-a2cd-7bb870a1cd4e	base
autoai-kb_3.1-py3.7	632d4b22-10aa-5180-88f0-f52dfb6444d7	base
pytorch-onnx_1.7-py3.8	634d3cdc-b562-5bf9-a2d4-ea90a478456b	base

Note: Only first 50 records were displayed. To display more use 'limit' parameter.

save and Deploy the model


```
In [23]: import sklearn
sklearn.__version__
```

Out[23]: '1.0.2'

```
In [24]: MODEL_NAME = 'DemoModel_MLR'
DEPLOYMENT_NAME = 'Wind turbine'
DEMO_MODEL = forest_model
```

```
In [25]: software_spec_uid = wml_client.software_specifications.get_id_by_name('runti
```

```
In [26]: # setup model meta
model_props = {
    wml_client.repository.ModelMetaNames.NAME: MODEL_NAME,
    wml_client.repository.ModelMetaNames.TYPE: 'scikit-learn_1.0',
    wml_client.repository.ModelMetaNames.SOFTWARE_SPEC_UID: software_spec_uid
}
```

```
In [27]: #save model
model_details = wml_client.repository.store_model(
    model=DEMO_MODEL,
    meta_props=model_props,
    training_data=train_X,
    training_target=train_y
)
```

```
In [28]: model_details
```

```
Out[28]: {'entity': {'hybrid_pipeline_software_specs': [],
  'label_column': 'ActivePower(KW)',
  'schemas': {'input': [{'fields': [{'name': 'Theoretical_Power_Curve (KW
h)',
  'type': 'float64'},
  {'name': 'WindSpeed(m/s)', 'type': 'float64'}]},
  'id': '1',
  'type': 'struct'}]},
  'output': [],
  'software_spec': {'id': '12b83a17-24d8-5082-900f-0ab31fbfd3cb',
  'name': 'runtime-22.1-py3.9'},
  'type': 'scikit-learn_1.0'},
  'metadata': {'created_at': '2022-11-16T16:01:18.347Z',
  'id': '59ecfa2f-181a-47a9-930d-648f237de348',
  'modified_at': '2022-11-16T16:01:22.204Z',
  'name': 'DemoModel_MLR',
  'owner': 'IBMid-668000E1AW',
  'resource_key': '4c7ea175-3c08-4b77-92d3-b966c6ac9624',
  'space_id': 'f1f8ff94-56cd-4e00-a0c9-f8ae5161c050'},
  'system': {'warnings': []}}
```

```
In [29]: model_id = wml_client.repository.get_model_id(model_details)
```

```
In [30]: model_id
```

Out[30]: '59ecfa2f-181a-47a9-930d-648f237de348'


```
In [31]: #deploy in purpose
deployment_props = {
    wml_client.deployments.ConfigurationMetaNames.NAME:DEPLOYMENT_NAME,
    wml_client.deployments.ConfigurationMetaNames.ONLINE:{}
}
```

```
In [32]: deployment = wml_client.deployments.create(
    artifact_uid=model_id,
    meta_props=deployment_props
)
```

```
#####
#####
```

Synchronous deployment creation for uid: '59ecfa2f-181a-47a9-930d-648f237de348' started

```
#####
#####
```

initializing

Note: online_url is deprecated and will be removed in a future release. Use serving_urls instead.

ready

```
-----
-----
Successfully finished deployment creation, deployment_uid='8e0aa5c6-3b8f-46d6-a314-bf5b6a73d66d'
-----
-----
```

In []:

In []:

In []: