

TRAIN THE ML MODEL

```
#import required labraries
import pandas as pd
import numpy as np
from sklearn.preprocessing import MinMaxScaler
from sklearn.metrics import confusion_matrix, accuracy_score
```

In [48]:

```
import matplotlib.pyplot as plt
plt.style.use('ggplot')
%matplotlib inline
```

In []:

```
import os, types
import pandas as pd
from botocore.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
includes your credentials.
# You might want to remove those credentials before you share the notebook.
cos_client = ibm_boto3.client(service_name='s3',
    ibm_api_key_id='MNIUYLIWesIem9Eh5uzuzgRGvqx5IYSqSnhJM_pUKlhH',
    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 = 'modelbuilding-donotdelete-pr-gj3r2r4kwyxkto'
object_key = 'Model Building.ipynb'
```

```
streaming_body_2 = cos_client.get_object(Bucket=bucket,
Key=object_key)['Body']
```

```
# Your data file was loaded into a botocore.response.StreamingBody object.
# Please read the documentation of ibm_boto3 and pandas to learn more about
the possibilities to load the data.
# ibm_boto3 documentation: https://ibm.github.io/ibm-cos-sdk-python/
# pandas documentation: http://pandas.pydata.org/
```

```
import os, types
import pandas as pd
from botocore.client import Config
import ibm_boto3
```

```
def __iter__(self): return 0
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# @hidden_cell
# The following code accesses a file in your IBM Cloud Object Storage. It
includes your credentials.
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cos_client = ibm_boto3.client(service_name='s3',
    ibm_api_key_id='MNIUYLIWesIem9Eh5uzuzgRGvqx5IYSqSnhJM_pUKlhH',
```

```

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 = 'modelbuilding-donotdelete-pr-gj3r2r4kwyxkto'
object_key = 'Model Building.ipynb'

```

```

streaming_body_1 = cos_client.get_object(Bucket=bucket,
Key=object_key)['Body']

```

```

# Your data file was loaded into a botocore.response.StreamingBody object.
# Please read the documentation of ibm_boto3 and pandas to learn more about
the possibilities to load the data.
# ibm_boto3 documentation: https://ibm.github.io/ibm-cos-sdk-python/
# pandas documentation: http://pandas.pydata.org/

```

In [49]:

```

#read the data set
dataset_train=pd.read_csv(r"/content/Dataset-20221111T140216Z-
001.zip",sep=' ', header=None).drop([26,27],axis=1)
col_names=['id', 'cycle',
'setting1','setting2','setting3','s1','s2','s3','s4','s5','s6','s7','s8','s
9','s10','s11','s12','s13','s14','s15','s16','s17','s18','s19','s20','s21']
dataset_train.columns=col_names
print ('Shape of Train dataset: ',dataset_train.shape)
dataset_train.head()

```

Shape of Train dataset: (20631, 26)

Out[49]:

	i	d	c	y	cl	e	set	tin	g1	set	tin	g2	set	tin	g3	s1	s2	s3	s4	s5	.	s1	s1	s1	s1	s	s	s	s	s	s2
0	1	1					-	-								51	64	15	14	1	.	52	23	81	8.	0	3	2	1	3	23.
							0.0	0.0								8.	1.	89.	00.	4.	.	1.	88.	38.	41	.	3	3	0	9.	41
							00	00								67	82	70	60	6	.	66	02	62	95	0	9	8	0.	0	90
							7	4												2	.					3	2	8	0	6	
1	1	2					0.0	-								51	64	15	14	1	.	52	23	81	8.	0	3	2	1	3	23.
							01	0.0								8.	2.	91.	03.	4.	.	2.	88.	31.	43	.	9	3	0	9.	42
							9	00								67	15	82	14	6	.	28	07	49	18	0	2	8	0.	0	36
								3												2	.					3	2	8	0	0	
2	1	3					-	0.0								51	64	15	14	1	.	52	23	81	8.	0	3	2	1	3	23.
							0.0	0.0								8.	2.	87.	04.	4.	.	2.	88.	33.	41	.	3	3	0	8.	34
							04	00								67	35	99	20	6	.	42	03	23	78	0	9	8	0.	9	42
							3	3												2	.					3	0	8	0	5	
3	1	4					0.0	0.0								51	64	15	14	1	.	52	23	81	8.	0	3	2	1	3	23.
							00	00								8.	2.	82.	01.	4.	.	2.	88.	33.	36	.	9	3	0	8.	37
							7	0								67	35	79	87	6	.	86	08	83	82	0	2	8	0.	8	39
																				2	.					3	2	8	0	8	

	i	c	set	set	set	s1	s2	s3	s4	s	.	s1	s1	s1	s1	s	s	s	s	s	s2
	d	y	tin	tin	tin					5	.	2	3	4	5	1	1	1	1	2	1
		e	g1	g2	g3											6	7	8	9	0	
			-	-		51	64	15	14	1	.	52	23	81	8.	0	3	2	1	3	23.
4	1	5	0.0	0.0	10	8.	2.	82.	06.	4.	.	2.	88.	33.	42	.	9	3	0	8.	40
			01	00	0.0	67	37	85	22	6	.	19	04	80	94	0	3	8	0.	9	44
			9	2						2	.					3		8	0	0	

5 rows × 26 columns

In [50]:

```
dataset_test=pd.read_csv('/content/Dataset-20221111T140216Z-001.zip',sep='
',header=None).drop([26,27],axis=1)
dataset_test.columns=col_names
# dataset_test.head()
print('Shape of Test dataset:',dataset_train.shape)
dataset_train.head()
Shape of Test dataset: (20631, 26)
```

Out[50]:

	i	c	set	set	set	s1	s2	s3	s4	s	.	s1	s1	s1	s1	s	s	s	s	s	s2
	d	y	tin	tin	tin					5	.	2	3	4	5	1	1	1	1	2	1
		e	g1	g2	g3											6	7	8	9	0	
0	1	1	-	-		51	64	15	14	1	.	52	23	81	8.	0	3	2	1	3	23.
			0.0	0.0	10	8.	1.	89.	00.	4.	.	1.	88.	38.	41	.	9	3	0	9.	41
			00	00	0.0	67	82	70	60	6	.	66	02	62	95	0	2	8	0.	0	90
			7	4						2	.					3		8	0	6	
1	1	2	0.0	-		51	64	15	14	1	.	52	23	81	8.	0	3	2	1	3	23.
			01	0.0	10	8.	2.	91.	03.	4.	.	2.	88.	31.	43	.	9	3	0	9.	42
			9	00	0.0	67	15	82	14	6	.	28	07	49	18	0	2	8	0.	0	36
			3	3						2	.					3		8	0	0	
2	1	3	-	0.0		51	64	15	14	1	.	52	23	81	8.	0	3	2	1	3	23.
			0.0	0.0	10	8.	2.	87.	04.	4.	.	2.	88.	33.	41	.	9	3	0	8.	34
			04	00	0.0	67	35	99	20	6	.	42	03	23	78	0	0	8	0.	9	42
			3	3						2	.					3		8	0	5	
3	1	4	0.0	0.0		51	64	15	14	1	.	52	23	81	8.	0	3	2	1	3	23.
			00	00	10	8.	2.	82.	01.	4.	.	2.	88.	33.	36	.	9	3	0	8.	37
			7	0	0.0	67	35	79	87	6	.	86	08	83	82	0	2	8	0.	8	39
										2	.					3		8	0	8	
4	1	5	-	-		51	64	15	14	1	.	52	23	81	8.	0	3	2	1	3	23.
			0.0	0.0	10	8.	2.	82.	06.	4.	.	2.	88.	33.	42	.	9	3	0	8.	40
			01	00	0.0	67	37	85	22	6	.	19	04	80	94	0	3	8	0.	9	44
			9	2						2	.					3		8	0	0	

5 rows × 26 columns

In [51]:

```
pm_truth=pd.read_csv('/content/Aircraft-20221111T140443Z-001.zip',sep='
',header=None).drop([1],axis=1)
pm_truth.columns=['more']
pm_truth['id']=pm_truth.index+1
pm_truth.head()
```

Out[51]:

	more	id
0	112	1
1	98	2
2	69	3
3	82	4
4	91	5

In [52]:

```
#pre-process the dataset
rul=pd.DataFrame (dataset_test.groupby ('id')
['cycle'].max()).reset_index()
rul.columns=['id','max']
rul.head()
```

Out[52]:

	id	max
0	1	192
1	2	287
2	3	179
3	4	189
4	5	269

In [53]:

```
pm_truth['rtf']=pm_truth['more']+rul['max']
pm_truth.head()
```

Out[53]:

	more	id	rtf
0	112	1	304
1	98	2	385
2	69	3	248
3	82	4	271
4	91	5	360

In [54]:

```
#calculate time to failure
pm_truth.drop('more', axis=1, inplace=True)
dataset_test=dataset_test.merge(pm_truth,on=['id'],how='left')
dataset_test['ttf']=dataset_test['rtf'] - dataset_test['cycle']
dataset_test.drop('rtf', axis=1, inplace=True)
dataset_test.head()
```

Out[54]:

id			category			set1	set2	set3	s1	s2	s3	s4	s5	.	s1	s1	s1	s1	s1	s1	s2	s2	s2	tf
			tin1	tin2	tin3									.	3	4	5	6	7	8	9	0	1	
0	1	1	-	-		51	64	15	14	1	.	23	81	8.	0		2	1	3		23.	3		
			0.0	0.0	10	8.	1.	89.	00.	4.	.	88.	38.	41	.	3	3	0	9.	41	0			
			00	00	0.0	67	82	70	60	6	.	02	62	95	0	9	8	0	0	6	90	0		
			7	4					2					3										
1	1	2	0.0	-		51	64	15	14	1	.	23	81	8.	0		2	1	3		23.	3		
			01	0.0	10	8.	2.	91.	03.	4.	.	88.	31.	43	.	9	3	0	9.	42	0			
			9	00	0.0	67	15	82	14	6	.	07	49	18	0	2	8	0	0	0	36	2		
				3					2					3										
2	1	3	-	0.0		51	64	15	14	1	.	23	81	8.	0		2	1	3		23.	3		
			0.0	0.0	10	8.	2.	87.	04.	4.	.	88.	33.	41	.	9	3	0	8.	34	0			
			04	00	0.0	67	35	99	20	6	.	03	23	78	0	9	8	0	9	42	1			
			3	3				2					3	0			5							
3	1	4	0.0	0.0		51	64	15	14	1	.	23	81	8.	0		2	1	3		23.	3		
			00	00	10	8.	2.	82.	01.	4.	.	88.	33.	36	.	9	3	0	8.	37	0			
			7	0	0.0	67	35	79	87	6	.	08	83	82	0	2	8	0	8	39	0			
									2					3										
4	1	5	-	-	10	51	64	15	14		.	23	81	8.		3					23.	2		
			0.0	0.0	0.0	8.	2.	82.	06.	1	.	88.	33.	42	0	9	2	1	3	40	9			
						67	37	85	22	4.	.	04	80	94	.	3	3	0	8.	44	9			

i	d	c	y	cl	e	set	tin	g1	set	tin	g2	set	tin	g3	s1	s2	s3	s4	s5	.	s1	s1	s1	s	s	s	s1	s2	s2	t
																					3	4	5	1	1	1	9	0	1	f
						01	00								6									0	8	0.	9			
						9	2								2									3	8	0	0			

5 rows × 27 columns

In [55]:

```
dataset_train['ttf']=dataset_train.groupby
(['id'])['cycle'].transform(max)-dataset_train['cycle']
dataset_train.head()
```

Out[55]:

i	d	c	y	cl	e	set	tin	g1	set	tin	g2	set	tin	g3	s1	s2	s3	s4	s5	.	s1	s1	s1	s	s	s	s1	s2	s2	t
																					3	4	5	1	1	1	9	0	1	f
0	1	1				-	-								51	64	15	14	1		23	81	8.	0	3	2	1	3	23.	1
						0.0	0.0			10					8.	1.	89.	00.	4.	.	88.	38.	41	.	3	3	0	9.	41	9
						00	00			0.0					67	82	70	60	6	.	02	62	95	0	2	8	0.	0	90	1
						7	4												2	.				3		0	6			
1	1	2				0.0	-			10					51	64	15	14	1		23	81	8.	0	3	2	1	3	23.	1
						01	0.0			0.0					8.	2.	91.	03.	4.	.	88.	31.	43	.	3	3	0	9.	42	9
						9	3			0.0					67	15	82	14	6	.	07	49	18	0	2	8	0.	0	36	0
																			2	.				3		0	0			
2	1	3				-	0.0			10					51	64	15	14	1		23	81	8.	0	3	2	1	3	23.	1
						0.0	0.0			0.0					8.	2.	87.	04.	4.	.	88.	33.	41	.	3	3	0	8.	34	8
						04	3			0.0					67	35	99	20	6	.	03	23	78	0	0	8	0.	9	42	9
						3													2	.				3		0	5			
3	1	4				0.0	0.0			10					51	64	15	14	1		23	81	8.	0	3	2	1	3	23.	1
						00	0.0			0.0					8.	2.	82.	01.	4.	.	88.	33.	36	.	3	3	0	8.	37	8
						7	0			0.0					67	35	79	87	6	.	08	83	82	0	2	8	0.	8	39	8
																			2	.				3		0	8			
4	1	5				-	-			10					51	64	15	14	1		23	81	8.	0	3	2	1	3	23.	1
						0.0	0.0			0.0					8.	2.	82.	06.	4.	.	88.	33.	42	.	3	3	0	8.	40	8
						01	00			0.0					67	37	85	22	6	.	04	80	94	0	3	8	0.	9	44	7
						9	2												2	.				3		0	0			

5 rows × 27 columns

In [56]:

```
df_train=dataset_train.copy()
df_test=dataset_test.copy ()
period=30
df_train['label_bc']=df_train['ttf'].apply(lambda x: 1 if x <= period else
0)
```

```
df_test['label_bc']= df_test['ttf'].apply(lambda x: 1 if x <= period else
0)
df_train.head()
```

Out[56]:

	i	c	set	set	set	s1	s2	s3	s4	s5	.	s1	s1	s	s	s	s1	s2	s2	t	lab
	d	y	tin	tin	tin						.	4	5	1	1	1	9	0	1	t	el_
		e	g1	g2	g3						.			6	7	8				f	bc
0	1	1	-	-	10	51	64	15	14	1	.	81	8.	0	3	2	1	3	23.	1	
			0.0	0.0	0.0	8.	1.	89.	00.	4.	.	38.	41	.	9	8	0.	9.	41	9	0
			7	4		67	82	70	60	2	.	62	95	3	2	8	0	6	90	1	
1	1	2	0.0	-	10	51	64	15	14	1	.	81	8.	0	3	2	1	3	23.	1	
			01	0.0	0.0	8.	2.	91.	03.	4.	.	31.	43	.	9	3	0	9.	42	9	0
			9	00		67	15	82	14	6	.	49	18	0	2	8	0.	0	36	0	
				3						2	.			3		8	0	0			
2	1	3	-	0.0	10	51	64	15	14	1	.	81	8.	0	3	2	1	3	23.	1	
			0.0	0.0	0.0	8.	2.	87.	04.	4.	.	33.	41	.	9	3	0	8.	34	8	0
			04	00		67	35	99	20	6	.	23	78	0	0	8	0.	9	42	9	
			3	0						2	.			3		8	0	5			
3	1	4	0.0	0.0	10	51	64	15	14	1	.	81	8.	0	3	2	1	3	23.	1	
			00	0.0	0.0	8.	2.	82.	01.	4.	.	33.	36	.	9	3	0	8.	37	8	0
			7	0		67	35	79	87	6	.	83	82	0	2	8	0.	8	39	8	
										2	.			3		8	0	8			
4	1	5	-	-	10	51	64	15	14	1	.	81	8.	0	3	2	1	3	23.	1	
			0.0	0.0	0.0	8.	2.	82.	06.	4.	.	33.	42	.	9	3	0	8.	40	8	0
			01	0.0		67	37	85	22	6	.	80	94	0	3	8	0.	9	44	7	
			9	0						2	.			3		8	0	0			

5 rows × 28 columns

In [58]:

```
x=df_train.iloc[:, :-1].values
y=df_train.iloc[:, -1].values
from sklearn.model_selection import train_test_split
x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.3,random_state=3)
y
```

Out[58]:

```
array([0, 0, 0, ..., 1, 1, 1])
```

In [59]:

```
from sklearn.linear_model import LogisticRegression
model=LogisticRegression()
model.fit(x_train,y_train)

/usr/local/lib/python3.7/dist-packages/sklearn/linear_model/_logistic.py:81
8: ConvergenceWarning: lbfgs failed to converge (status=1):
STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.
```

Increase the number of iterations (max_iter) or scale the data as shown in:
<https://scikit-learn.org/stable/modules/preprocessing.html>
Please also refer to the documentation for alternative solver options:
https://scikit-learn.org/stable/modules/linear_model.html#logistic-regression

```
extra_warning_msg=_LOGISTIC_SOLVER_CONVERGENCE_MSG,
```

Out[59]:

```
LogisticRegression()
```

In [65]:

```
#Check the metrics of the model
from sklearn.metrics import accuracy_score
y_predlog=model.predict(x_train)
accuracy_score(y_predlog,y_train)
```

Out[65]:

```
1.0
```

In [61]:

```
y_pred_test=model.predict(x_test)
accuracy_score(y_pred_test,y_test)
```

Out[61]:

```
0.9998384491114701
```

In [62]:

```
from sklearn.metrics import confusion_matrix
cm1=confusion_matrix(y_test,y_pred_test)
cm1
```

Out[62]:

```
array([[5297,    1],
       [    0,  892]])
```

In [64]:

```
#saving the model
import joblib
joblib.dump(model, "engine_model.sav")
```

Out[64]:

```
['engine_model.sav']
```

deploy model

In [1]:

```
!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: 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: packaging in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from ibm-watson-machine-learning) (21.3)
```

```
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: 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: tabulate in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from ibm-watson-machine-learning) (0.8.9)

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: requests in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from ibm-watson-machine-learning) (2.26.0)

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-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: jmespath<1.0.0,>=0.7.1 in /opt/conda/envs/Python-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-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: 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: 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: 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 [2]:

```
from ibm_watson_machine_learning import APIClient
```

In [3]:

```
wml_credentials = {
    "url": "https://us-south.ml.cloud.ibm.com",
    "apikey" : "hNzkDg_2UyFOhp8N3SorKaqB14tQpUAT1Ci6RpWIyXTC"
}
```

In [14]:

```
wml_client = APIClient(wml_credentials)
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
c25e08e2-22e3-4932-bf81-77c30fadbe65  model                                2022-11-17T16:11:08.2
79Z
29fcbf45-6183-447c-8e34-758b3c2c8945  models                              2022-11-17T16:10:23.8
98Z
cfd83c6d-f976-45d7-abd8-194e088f12a6  model building                      2022-11-17T16:06:54.8
40Z
79c79f6b-2a3e-4fea-9dbc-bf7631a57192  model building                      2022-11-17T15:54:15.9
60Z
-----
---

```

```

In [15]:
space_id = "c25e08e2-22e3-4932-bf81-77c30fadbe65"

```

```

In [16]:
wml_client.set.default_space(space_id)

```

```

Out[16]:
'SUCCESS'

```

```

In [8]:
wml_client.software_specifications.list()

```

```

-----
NAME                                ASSET_ID                                TYPE
default_py3.6                      0062b8c9-8b7d-44a0-a9b9-46c416adcbd9  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         09f4cff0-90a7-5899-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-4240baled5f7	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
runtime-22.2-py3.10-xc	5e8cddff-db4a-5a6a-b8aa-2d4af9864dab	base
autoai-kb_3.1-py3.7	632d4b22-10aa-5180-88f0-f52dfb6444d7	base

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

In [9]:

```
MODEL_NAME = 'machine_learning'
DEPLOYMENT_NAME = 'machine_deploy'
MACHINE_MODEL = 'MODEL'
```

In [10]:

```
software_spec_uid =
wml_client.software_specifications.get_id_by_name('runtime-22.1-py3.9')
```

In [13]:

```
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 []:

```
model_details = wml_client.repository.store_model(
    model=MACHINE_MODEL,
    meta_props=model_props,
    training_data=x_train,
    training_target=y_train
)
```