Model building

#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 [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]: c set set set s2i y s1s1s1s1tin s2tin tin s1s3d cl g1 g2g38. 23. 0.0 9. 0.0 4. 0 1 89. 00. 88. 38. 8. 1. 1. 0.0 0. 0.0 8. 23. 0.0 4. 1 2. 91. 03. 88. 31. 0. 0.0 23. 0.0 8. 0.0 4. 8. 2 1 8. 2. 87. 04. 2. 88. 33. 0. 0.0 0.0 0.0 8. 23. 4. 8. 01. 1 8. 2. 82. 2. 88. 33. 0.0 0. 23. 8. 1 8. 2. 82. 06. 2. 88. 33. 0.0

0.0

0.0

 $5 \text{ rows} \times 26 \text{ columns}$

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]:

																				Ou	t[50].
	i d	c y cl e	set tin g1	set tin g2	set tin g3	s1	s2	s3	s4	s 5	•	s1 2	s1 3	s1 4	s1 5	s 1 6	s 1 7	s 1 8	s 1 9	s 2 0	s2 1
0	1	1	0.0 00 7	0.0 00 4	10 0.0	51 8. 67	64 1. 82	15 89. 70	14 00. 60	1 4. 6 2		52 1. 66	23 88. 02	81 38. 62	8. 41 95	0 0 3	3 9 2	2 3 8 8	1 0 0. 0	3 9. 0 6	23. 41 90
1	1	2	0.0 01 9	0.0 00 3	10 0.0	51 8. 67	64 2. 15	15 91. 82	14 03. 14	1 4. 6 2		52 2. 28	23 88. 07	81 31. 49	8. 43 18	0 0 3	3 9 2	2 3 8 8	1 0 0. 0	3 9. 0 0	23. 42 36
2	1	3	0.0 04 3	0.0 00 3	10 0.0	51 8. 67	64 2. 35	15 87. 99	14 04. 20	1 4. 6 2		52 2. 42	23 88. 03	81 33. 23	8. 41 78	0 0 3	3 9 0	2 3 8 8	1 0 0. 0	3 8. 9 5	23. 34 42
3	1	4	0.0 00 7	0.0 00 0	10 0.0	51 8. 67	64 2. 35	15 82. 79	14 01. 87	1 4. 6 2		52 2. 86	23 88. 08	81 33. 83	8. 36 82	0 0 3	3 9 2	2 3 8 8	1 0 0. 0	3 8. 8 8	23. 37 39
4	1	5	0.0 01 9	0.0 00 2	10 0.0	51 8. 67	64 2. 37	15 82. 85	14 06. 22	1 4. 6 2		52 2. 19	23 88. 04	81 33. 80	8. 42 94	0 0 3	3 9 3	2 3 8 8	1 0 0. 0	3 8. 9 0	23. 40 44

 $5 \text{ rows} \times 26 \text{ columns}$

```
', 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
     98
         2
 1
 2
     69
         3
     82 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
       192
       287
 1
   2
       179
 2
   3
 3 4
        189
 4 5
       269
                                                                         In [53]:
pm truth['rtf']=pm truth['more']+rul['max']
pm_truth.head()
                                                                        Out[53]:
```

pm truth=pd.read csv('/content/Aircraft-20221111T140443Z-001.zip',sep='

```
304
      112
       98
            2
                385
 1
       69
 2
            3
                248
                271
 3
       82
       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]:
            set
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                       set
                                                                                              s2
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                                           87
                                                        08
                                                              83
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88.

33.

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23.

id

more

rtf

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          tin
               tin
                    s1 s2
                               s3
                                                                                            t
cl
     g1
          g2
                g3
          00
     01
                                          2
                                                                                  0
           2
```

 $5 \text{ rows} \times 27 \text{ columns}$

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

 $5 \text{ rows} \times 27 \text{ columns}$

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)</pre>
```

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

																			Ou	ıt[56]:
	i d	c y cl e	set tin g1	set tin g2	set tin g3	s1	s2	s3	s4	s5	s1 4	s1 5	s 1 6	s 1 7	s 1 8	s1 9	s2 0	s2 1	t t f	lab el_ bc
0	1	1	0.0 00 7	0.0 00 4	10 0.0	51 8. 67	64 1. 82	15 89. 70	14 00. 60	1 4. 6 2	 81 38. 62	8. 41 95	0 0 3	3 9 2	2 3 8 8	1 0 0. 0	3 9. 0 6	23. 41 90	1 9 1	0
1	1	2	0.0 01 9	0.0 00 3	10 0.0	51 8. 67	64 2. 15	15 91. 82	14 03. 14	1 4. 6 2	81 31. 49	8. 43 18	0 0 3	3 9 2	2 3 8 8	1 0 0. 0	3 9. 0 0	23. 42 36	1 9 0	0
2	1	3	0.0 04 3	0.0 00 3	10 0.0	51 8. 67	64 2. 35	15 87. 99	14 04. 20	1 4. 6 2	81 33. 23	8. 41 78	0 0 3	3 9 0	2 3 8 8	1 0 0. 0	3 8. 9 5	23. 34 42	1 8 9	0
3	1	4	0.0 00 7	0.0 00 0	10 0.0	51 8. 67	64 2. 35	15 82. 79	14 01. 87	1 4. 6 2	 81 33. 83	8. 36 82	0 0 3	3 9 2	2 3 8 8	1 0 0. 0	3 8. 8 8	23. 37 39	1 8 8	0
4	1	5	0.0 01 9	0.0 00 2	10 0.0	51 8. 67	64 2. 37	15 82. 85	14 06. 22	1 4. 6 2	 81 33. 80	8. 42 94	0 0 3	3 9 3	2 3 8 8	1 0 0. 0	3 8. 9 0	23. 40 44	1 8 7	0

 $5 \text{ rows} \times 28 \text{ columns}$

model.fit(x train,y train)

```
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()
```

/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-regr
  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)
                                                                       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']
```