

```
In [81]: import numpy as np
import pandas as pd
import warnings
train_file = r'D:\python_project\logistic_regression\train_data_set.csv'
test_file = r'D:\python_project\logistic_regression\test_data_set.csv'
mt_train = pd.read_csv(train_file)
mt_test = pd.read_csv(test_file)
```

```
In [82]: mt_train.columns.nunique()
mt_test.columns.nunique()
```

Out[82]: 11

```
In [83]: mt_test.head()
```

	created_at	entry_id	current	voltage	vibration	temperature	latitude	longitude	relay_st
0	2022-04-05T13:30:41+05:30	1	2.223148	223.14752	0	37.36264	NaN	NaN	
1	2022-04-05T13:30:58+05:30	2	2.213905	224.07919	0	37.48474	NaN	NaN	
2	2022-04-05T13:31:14+05:30	3	2.333177	223.81500	0	37.60684	NaN	NaN	
3	2022-04-05T13:31:30+05:30	4	2.097085	224.11105	0	37.24054	NaN	NaN	
4	2022-04-05T13:31:47+05:30	5	2.249357	223.86400	0	37.36264	NaN	NaN	

```
In [84]: mt_train.head()
```

	created_at	entry_id	current	voltage	vibration	temperature	latitude	longitude	relay_st
0	2022-03-15T17:17:51+05:30	1	1.938089	238.64746	0	37.11844	NaN	NaN	
1	2022-03-15T17:18:07+05:30	2	2.352603	215.62500	0	37.36264	NaN	NaN	
2	2022-03-15T17:18:24+05:30	3	3.654926	152.06055	0	37.24054	NaN	NaN	
3	2022-03-15T17:18:41+05:30	4	2.420011	195.74707	0	37.11844	NaN	NaN	
4	2022-03-15T17:19:00+05:30	5	2.632627	177.10449	0	37.11844	NaN	NaN	

```
In [85]: mt_train['data']='train'
```

```
mt_test['data']='test'  
mt_all = pd.concat([mt_train,mt_test],axis=0)
```

In [86]: `mt_all.nunique()`

```
Out[86]: created_at      952  
entry_id        759  
current         952  
voltage         943  
vibration       2  
temperature     142  
latitude         0  
longitude        0  
relay_status     2  
humidity         37  
failure          2  
data             2  
dtype: int64
```

In [87]: `mt_all.head(1392)`

```
Out[87]:   created_at  entry_id  current  voltage  vibration  temperature  latitude  longitude  relay  
0    2022-03-  
     15T17:17:51+05:30      1  1.938089  238.64746      0  37.11844      NaN      NaN  
1    2022-03-  
     15T17:18:07+05:30      2  2.352603  215.62500      0  37.36264      NaN      NaN  
2    2022-03-  
     15T17:18:24+05:30      3  3.654926  152.06055      0  37.24054      NaN      NaN  
3    2022-03-  
     15T17:18:41+05:30      4  2.420011  195.74707      0  37.11844      NaN      NaN  
4    2022-03-  
     15T17:19:00+05:30      5  2.632627  177.10449      0  37.11844      NaN      NaN  
...  
188   2022-04-  
      05T14:45:50+05:30    189  2.447289  211.85481      0  43.83394      NaN      NaN  
189   2022-04-  
      05T14:46:06+05:30    190  2.425245  212.24246      0  44.07814      NaN      NaN  
190   2022-04-  
      05T14:46:22+05:30    191  4.281932  238.45860      0  45.12915      NaN      NaN  
191   2022-04-  
      05T14:46:38+05:30    192  3.826854  401.23560      0  47.26988      NaN      NaN  
192   2022-04-  
      05T14:46:54+05:30    193  4.181952  250.38970      0  45.63972      NaN      NaN
```

952 rows × 12 columns

In [88]: `mt_all.drop(['created_at','entry_id','relay_status','longitude','latitude'],axis=1,inpl`

```
mt_all.head()
```

```
Out[88]:
```

	current	voltage	vibration	temperature	humidity	failure	data
0	1.938089	238.64746	0	37.11844	25	yes	train
1	2.352603	215.62500	0	37.36264	25	no	train
2	3.654926	152.06055	0	37.24054	46	no	train
3	2.420011	195.74707	0	37.11844	26	no	train
4	2.632627	177.10449	0	37.11844	26	no	train

```
In [89]:
```

```
mt_all['vibration']=(mt_all['vibration']==1).astype(int)
mt_all['vibration'].unique()
```

```
Out[89]: array([0, 1])
```

```
In [90]:
```

```
mt_all['failure']=(mt_all['failure']=='yes').astype(int)
#mt_all['failure']=pd.to_numeric(mt_all['failure'],errors='coerce')
mt_all['failure'].unique()
```

```
Out[90]: array([1, 0])
```

```
In [91]:
```

```
mt_all['failure'].unique()
```

```
Out[91]: array([1, 0])
```

```
In [92]:
```

```
for col in mt_all.columns:
    if col=='data' or mt_all[col].isnull().sum()==0:continue
    mt_all.loc[mt_all[col].isnull(),col]=mt_all.loc[mt_all['data']=='train',col].mean()
```

```
In [93]:
```

```
mt_train = mt_all[mt_all['data']=='train']
del mt_train['data']
mt_test = mt_all[mt_all['data']=='test']
mt_test.drop(['data','failure'],axis=1,inplace=True)
```

C:\Users\Omkar250\anaconda3\lib\site-packages\pandas\core\frame.py:4308: SettingWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy
return super().drop()

```
In [94]:
```

```
from sklearn.linear_model import LogisticRegression
```

```
In [95]:
```

```
params = {'class_weight':['balanced'],'penalty':['l1','l2'],
'C':np.linspace(0.0001,1000,10)}
```

```
In [96]: model = LogisticRegression(fit_intercept=True)

In [97]: from sklearn.model_selection import GridSearchCV

In [98]: grid_search = GridSearchCV(model, param_grid=params, cv=10, scoring='roc_auc', n_jobs=-1)

In [99]: x_train = mt_train.drop('failure', axis=1)
y_train = mt_train['failure']

In [100...]: grid_search.fit(x_train, y_train)

C:\Users\Omkar250\anaconda3\lib\site-packages\sklearn\model_selection\_search.py:918: UserWarning: One or more of the test scores are non-finite: [      nan  0.82535887
nan  0.98152174      nan  0.98244306
      nan  0.98291925      nan  0.98128364      nan  0.98271222
      nan  0.98196687      nan  0.98220497      nan  0.98312629
      nan  0.98291925]
warnings.warn(
C:\Users\Omkar250\anaconda3\lib\site-packages\sklearn\linear_model\_logistic.py:763: 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
n_iter_i = _check_optimize_result()

Out[100...]: GridSearchCV(cv=10, estimator=LogisticRegression(), n_jobs=-1,
                     param_grid={'C': array([1.000000e-04, 1.111112e+02, 2.222223e+02, 3.333334e
+02,
                     4.444445e+02, 5.555556e+02, 6.666667e+02, 7.777778e+02,
                     8.888889e+02, 1.000000e+03]),
                     'class_weight': ['balanced'],
                     'penalty': ['l1', 'l2']},
                     scoring='roc_auc')

In [101...]: grid_search.best_estimator_

Out[101...]: LogisticRegression(C=888.8889, class_weight='balanced')

In [102...]: mt_train['failure'].nunique()
grid_search.cv_results_

Out[102...]: {'mean_fit_time': array([0.00290363, 0.05640724, 0.0026552 , 0.09638443, 0.00205066,
       0.07529125, 0.00274177, 0.07637553, 0.00223219, 0.06401577,
       0.00239375, 0.06791797, 0.00214696, 0.08126185, 0.00219271,
       0.07892537, 0.00209434, 0.07617378, 0.0023937 , 0.07993965]),
 'std_fit_time': array([0.00015297, 0.01308862, 0.00043239, 0.01876899, 0.00016699,
       0.00655309, 0.00116188, 0.00711257, 0.00039607, 0.01110674,
       0.00066166, 0.01249217, 0.00032261, 0.00944136, 0.00039957,
       0.00947944, 0.00036288, 0.00777443, 0.00091423, 0.00716464]),
 'mean_score_time': array([0.          , 0.00815263, 0.          , 0.01017141, 0.          ,
       0.00708137, 0.          , 0.00628347, 0.          , 0.00579731,
       0.          , 0.00589328, 0.          , 0.00688162, 0.          , 0.          ]),
```

```

0.00804198, 0. , 0.00602436, 0. , 0.00563202]),),
'std_score_time': array([0. , 0.00083781, 0. , 0.00725848, 0. ,
0.00175321, 0. , 0.0006386 , 0. , 0.00125311,
0. , 0.00157448, 0. , 0.00082822, 0. ,
0.00406582, 0. , 0.00079393, 0. , 0.00116036]),),
'param_C': masked_array(data=[0.0001, 0.0001, 111.1112000000001, 111.11120000000001,
222.2223000000002, 222.2223000000002, 333.3334,
333.3334, 444.4445, 444.4445, 555.5556, 555.5556,
666.6667, 666.6667, 777.7778000000001,
777.7778000000001, 888.8889, 888.8889, 1000.0, 1000.0],
mask=[False, False, False, False, False, False, False,
False, False, False, False, False, False, False,
False, False, False],),
fill_value='?',
dtype=object),
'param_class_weight': masked_array(data=['balanced', 'balanced', 'balanced', 'balance
d',
'balanced', 'balanced', 'balanced', 'balanced',
'balanced', 'balanced', 'balanced', 'balanced',
'balanced', 'balanced', 'balanced', 'balanced'],
mask=[False, False, False, False, False, False, False,
False, False, False, False, False, False, False,
False, False, False],),
fill_value='?',
dtype=object),
'param_penalty': masked_array(data=['l1', 'l2', 'l1', 'l2', 'l1', 'l2', 'l1', 'l2', 'l1',
'l2', 'l1', 'l2', 'l1', 'l2', 'l1', 'l2', 'l1', 'l2',
'l1', 'l2'],
mask=[False, False, False, False, False, False, False, False,
False, False, False, False, False, False, False, False,
False, False, False],),
fill_value='?',
dtype=object),
'params': [{"C": 0.0001, "class_weight": "balanced", "penalty": "l1"}, { "C": 0.0001, "class_weight": "balanced", "penalty": "l2"}, {"C": 111.1112000000001, "class_weight": "balanced", "penalty": "l1"}, {"C": 111.1112000000001, "class_weight": "balanced", "penalty": "l2"}, {"C": 222.2223000000002, "class_weight": "balanced", "penalty": "l1"}, {"C": 222.2223000000002, "class_weight": "balanced", "penalty": "l2"}, {"C": 333.3334, "class_weight": "balanced", "penalty": "l1"}, {"C": 333.3334, "class_weight": "balanced", "penalty": "l2"}, {"C": 444.4445, "class_weight": "balanced", "penalty": "l1"}, {"C": 444.4445, "class_weight": "balanced", "penalty": "l2"}, {"C": 555.5556, "class_weight": "balanced", "penalty": "l1"}, {"C": 555.5556, "class_weight": "balanced", "penalty": "l2"}, {"C": 666.6667, "class_weight": "balanced", "penalty": "l1"}, {"C": 666.6667, "class_weight": "balanced", "penalty": "l2"}, {"C": 777.7778000000001, "class_weight": "balanced", "penalty": "l1"}, {"C": 777.7778000000001, "class_weight": "balanced", "penalty": "l2"}, {"C": 888.8889, "class_weight": "balanced", "penalty": "l1"}, {"C": 888.8889, "class_weight": "balanced", "penalty": "l2"}, {"C": 1000.0, "class_weight": "balanced", "penalty": "l1"}, {"C": 1000.0, "class_weight": "balanced", "penalty": "l2"}],),
'split0_test_score': array([ nan, 0.99585921, nan, 0.99378882, nan,
0.99585921, nan, 0.99585921, nan, 0.99378882,
nan, 0.99378882, nan, 0.99585921, nan,
0.99585921, nan, 0.99792961, nan, 0.99585921]),),
'split1_test_score': array([nan, 1., nan, 1., nan, 1., nan, 1., nan, 1.,
1., nan, 1., nan, 1., nan, 1.]),
'split2_test_score': array([nan, 1., nan, 1., nan, 1., nan, 1., nan, 1.,
1., nan, 1., nan, 1.]),)

```

```

'split3_test_score': array([      nan,  0.73809524,      nan,  1.        ,
    1.        ,      nan,  1.        ,      nan,  1.        ,
    nan,  1.        ,      nan,  1.        ,      nan,
    1.        ,      nan,  1.        ,      nan,  1.        ]),
'split4_test_score': array([nan,  1.,  nan,  1.,  nan,  1.,  nan,  1.,  nan,  1.,
    nan,
    1.,  nan,  1.,  nan,  1.]),
'split5_test_score': array([      nan,  0.98571429,      nan,  1.        ,
    1.        ,      nan,  1.        ,      nan,  1.        ,
    nan,  1.        ,      nan,  1.        ,      nan,
    1.        ,      nan,  1.        ,      nan,  1.        ]),
'split6_test_score': array([      nan,  0.98571429,      nan,  1.        ,
    1.        ,      nan,  1.        ,      nan,  1.        ,
    nan,  1.        ,      nan,  1.        ,      nan,
    1.        ,      nan,  1.        ,      nan,  1.        ]),
'split7_test_score': array([nan,  1.,  nan,  1.,  nan,  1.,  nan,  1.,  nan,  1.,
    nan,
    1.,  nan,  1.,  nan,  1.]),
'split8_test_score': array([      nan,  0.50714286,      nan,  0.98809524,
    0.9952381 ,      nan,  1.        ,      nan,  0.98571429,
    nan,  1.        ,      nan,  0.99047619,      nan,
    0.99285714,      nan,  1.        ,      nan,  1.        ]),
'split9_test_score': array([      nan,  0.0410628 ,      nan,  0.83333333,
    0.83333333,      nan,  0.83333333,      nan,  0.83333333,
    nan,  0.83333333,      nan,  0.83333333,      nan,
    0.83333333,      nan,  0.83333333,      nan,  0.83333333]),
'mean_test_score': array([      nan,  0.82535887,      nan,  0.98152174,
    0.98244306,      nan,  0.98291925,      nan,  0.98128364,
    nan,  0.98271222,      nan,  0.98196687,      nan,
    0.98220497,      nan,  0.98312629,      nan,  0.98291925]),
'std_test_score': array([      nan,  0.30438346,      nan,  0.04954152,
    0.04973443,      nan,  0.04987725,      nan,  0.04951507,
    nan,  0.04982738,      nan,  0.04963234,      nan,
    0.04967828,      nan,  0.0499348 ,      nan,  0.04987725]),
'rank_test_score': array([20, 10, 17, 8, 15, 5, 13, 2, 19, 9, 12, 4, 14, 7, 16,
    6, 18,
    1, 11, 2])}

```

In [103...]

```

def report(results, n_top=3):
    for i in range(1, n_top + 1):
        candidates = np.flatnonzero(results['rank_test_score'] == i)
        for candidate in candidates:
            print("Model with rank: {0}".format(i))
            print("Mean validation score: {0:.6f} (std: {1:.6f})".format(
                results['mean_test_score'][candidate],
                results['std_test_score'][candidate]))
            print("Parameters: {0}".format(results['params'][candidate]))
            print("")

```

In [104...]

```
report(grid_search.cv_results_, 3)
```

```

Model with rank: 1
Mean validation score: 0.983126 (std: 0.049935)
Parameters: {'C': 888.8889, 'class_weight': 'balanced', 'penalty': 'l2'}

```

```

Model with rank: 2
Mean validation score: 0.982919 (std: 0.049877)
Parameters: {'C': 333.3334, 'class_weight': 'balanced', 'penalty': 'l2'}

```

```

Model with rank: 2
Mean validation score: 0.982919 (std: 0.049877)

```

```
Parameters: {'C': 1000.0, 'class_weight': 'balanced', 'penalty': 'l2'}
```

```
In [105...]: test_prediction = grid_search.predict_proba(mt_test)
```

```
In [106...]: test_prediction
```

```
Out[106...]: array([[9.81105344e-01, 1.88946561e-02],  
[9.78083905e-01, 2.19160945e-02],  
[9.74408974e-01, 2.55910259e-02],  
[9.81867360e-01, 1.81326395e-02],  
[9.76969607e-01, 2.30303933e-02],  
[9.86168930e-01, 1.38310697e-02],  
[9.81344741e-01, 1.86552589e-02],  
[9.83758099e-01, 1.62419008e-02],  
[9.83244606e-01, 1.67553938e-02],  
[9.79377946e-01, 2.06220543e-02],  
[9.89307585e-01, 1.06924149e-02],  
[9.85626923e-01, 1.43730767e-02],  
[9.72945867e-01, 2.70541334e-02],  
[9.84986309e-01, 1.50136910e-02],  
[9.76298213e-01, 2.37017868e-02],  
[9.89457039e-01, 1.05429611e-02],  
[9.80636745e-01, 1.93632551e-02],  
[9.82757467e-01, 1.72425332e-02],  
[9.84820278e-01, 1.51797220e-02],  
[9.90403553e-01, 9.59644733e-03],  
[9.87243805e-01, 1.27561952e-02],  
[9.90800646e-01, 9.19935420e-03],  
[9.93581387e-01, 6.41861326e-03],  
[9.93224753e-01, 6.77524686e-03],  
[9.88180468e-01, 1.18195321e-02],  
[9.93070183e-01, 6.92981716e-03],  
[9.76655255e-01, 2.33447450e-02],  
[9.82636101e-01, 1.73638985e-02],  
[9.87855736e-01, 1.21442644e-02],  
[9.83851686e-01, 1.61483136e-02],  
[9.87190949e-01, 1.28090515e-02],  
[9.86368830e-01, 1.36311697e-02],  
[9.87401217e-01, 1.25987833e-02],  
[9.87047485e-01, 1.29525147e-02],  
[9.88259824e-01, 1.17401764e-02],  
[9.79701706e-01, 2.02982940e-02],  
[9.79510762e-01, 2.04892376e-02],  
[9.90579723e-01, 9.42027652e-03],  
[9.93753303e-01, 6.24669661e-03],  
[9.85413350e-01, 1.45866500e-02],  
[9.93425330e-01, 6.57466969e-03],  
[9.86796229e-01, 1.32037715e-02],  
[9.86367979e-01, 1.36320205e-02],  
[9.90565718e-01, 9.43428167e-03],  
[9.86397107e-01, 1.36028930e-02],  
[9.79775898e-01, 2.02241021e-02],  
[9.82619491e-01, 1.73805088e-02],  
[9.88287059e-01, 1.17129410e-02],  
[9.83931883e-01, 1.60681174e-02],  
[9.90635576e-01, 9.36442440e-03],  
[9.78909022e-01, 2.10909781e-02],  
[9.92854627e-01, 7.14537336e-03],  
[9.83175522e-01, 1.68244783e-02],  
[9.87033997e-01, 1.29660033e-02],  
[9.93006109e-01, 6.99389147e-03],
```

[9.91799522e-01, 8.20047790e-03],
[9.84013681e-01, 1.59863191e-02],
[9.91268444e-01, 8.73155572e-03],
[9.91887872e-01, 8.11212824e-03],
[9.89131353e-01, 1.08686474e-02],
[9.93719288e-01, 6.28071186e-03],
[9.92410850e-01, 7.58914987e-03],
[9.89109046e-01, 1.08909541e-02],
[9.95395866e-01, 4.60413405e-03],
[9.86725962e-01, 1.32740377e-02],
[9.91026590e-01, 8.97340988e-03],
[9.95542215e-01, 4.45778541e-03],
[9.89139291e-01, 1.08607092e-02],
[9.96117939e-01, 3.88206148e-03],
[9.96164629e-01, 3.83537095e-03],
[9.90984993e-01, 9.01500695e-03],
[9.84356963e-01, 1.56430368e-02],
[9.91764190e-01, 8.23580971e-03],
[9.89382982e-01, 1.06170179e-02],
[9.90187805e-01, 9.81219523e-03],
[9.95068980e-01, 4.93102020e-03],
[9.91114377e-01, 8.88562278e-03],
[9.88534339e-01, 1.14656611e-02],
[9.88324753e-01, 1.16752467e-02],
[9.97182061e-01, 2.81793856e-03],
[9.88797002e-01, 1.12029976e-02],
[9.94188417e-01, 5.81158324e-03],
[9.89916046e-01, 1.00839538e-02],
[9.87646056e-01, 1.23539436e-02],
[9.94454136e-01, 5.54586440e-03],
[9.88588882e-01, 1.14111179e-02],
[9.88282198e-01, 1.17178017e-02],
[9.94084187e-01, 5.91581264e-03],
[9.93086048e-01, 6.91395247e-03],
[9.95378419e-01, 4.62158122e-03],
[9.87093313e-01, 1.29066865e-02],
[9.94219738e-01, 5.78026210e-03],
[9.93267171e-01, 6.73282860e-03],
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```

```
In [107... grid_search.classes_
```

```
Out[107... array([0, 1])
```

```
In [108... train_score=grid_search.predict_proba(x_train)[:,1]  
real = y_train
```

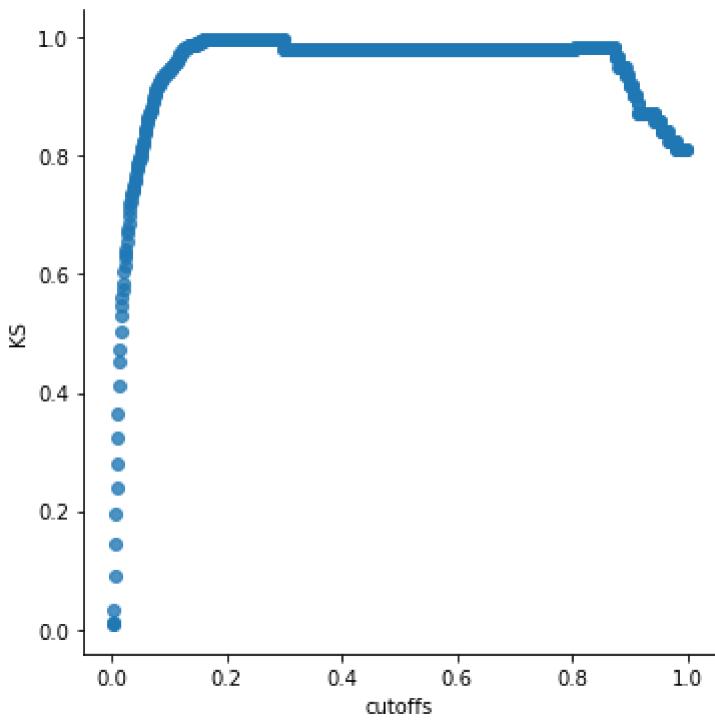
```
In [109... cutoffs = np.linspace(.001,0.999, 999)
```

```
In [110... import seaborn as sns
```

```
In [111... KS=[]
```

```
In [112... for cutoff in cutoffs:  
    predicted=(train_score>cutoff).astype(int)  
    TP=((real==1)&(predicted==1)).sum()  
    FP=((real==0)&(predicted==1)).sum()  
    TN=((real==0)&(predicted==0)).sum()  
    FN=((real==1)&(predicted==0)).sum()  
  
    ks=(TP/(TP+FN))-(FP/(TN+FP))  
    KS.append(ks)  
  
temp=pd.DataFrame({'cutoffs':cutoffs,'KS':KS})  
sns.lmplot(x='cutoffs',y='KS',data=temp,fit_reg=False)
```

```
Out[112... <seaborn.axisgrid.FacetGrid at 0x2660e28f3d0>
```



In [113...]

ks

```
Out[113... 0.8095238095238095
```

In [114...]

```
test_hard_classes=(test_prediction>cutoffs[KS==max(KS)][0]).astype(int)
```

In [115...]

test hard classes


```
In [117]: pd.DataFrame(test_hard_classes).to_csv("mysubmissionelogistic.csv", index=False)
```

In []: