

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

```
In [2]: from sklearn.linear_model import LogisticRegression
```

```
In [3]: df=pd.read_csv("Data.csv").dropna()
df
```

Out[3]:

|       | row_id | user_id | timestamp           | gate_id |
|-------|--------|---------|---------------------|---------|
| 0     | 0      | 18      | 2022-07-29 09:08:54 | 7       |
| 1     | 1      | 18      | 2022-07-29 09:09:54 | 9       |
| 2     | 2      | 18      | 2022-07-29 09:09:54 | 9       |
| 3     | 3      | 18      | 2022-07-29 09:10:06 | 5       |
| 4     | 4      | 18      | 2022-07-29 09:10:08 | 5       |
| ...   | ...    | ...     | ...                 | ...     |
| 37513 | 37513  | 6       | 2022-12-31 20:38:56 | 11      |
| 37514 | 37514  | 6       | 2022-12-31 20:39:22 | 6       |
| 37515 | 37515  | 6       | 2022-12-31 20:39:23 | 6       |
| 37516 | 37516  | 6       | 2022-12-31 20:39:31 | 9       |
| 37517 | 37517  | 6       | 2022-12-31 20:39:31 | 9       |

37518 rows × 4 columns

```
In [4]: df.dropna(inplace=True)
```

```
In [5]: df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
Int64Index: 37518 entries, 0 to 37517
Data columns (total 4 columns):
 #   Column      Non-Null Count  Dtype  
---  --          --          --      
 0   row_id      37518 non-null   int64  
 1   user_id     37518 non-null   int64  
 2   timestamp   37518 non-null   object 
 3   gate_id     37518 non-null   int64  
dtypes: int64(3), object(1)
memory usage: 1.4+ MB
```

```
In [6]: feature_matrix = df[['row_id','user_id']]
target_vector = df['gate_id']
```

```
In [7]: feature_matrix.shape
```

```
Out[7]: (37518, 2)
```

```
In [8]: target_vector.shape
```

```
Out[8]: (37518,)
```

```
In [9]: from sklearn.preprocessing import StandardScaler
```

```
In [10]: fs = StandardScaler().fit_transform(feature_matrix)
```

```
In [11]: logr = LogisticRegression()
logr.fit(fs,target_vector)
```

```
C:\ProgramData\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[11]: LogisticRegression()
```

```
In [12]: feature_matrix.shape
```

```
Out[12]: (37518, 2)
```

```
In [13]: target_vector.shape
```

```
Out[13]: (37518,)
```

```
In [14]: from sklearn.preprocessing import StandardScaler
```

```
In [15]: fs = StandardScaler().fit_transform(feature_matrix)
```

```
In [16]: logr = LogisticRegression()
logr.fit(fs,target_vector)
```

```
C:\ProgramData\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[16]: LogisticRegression()
```

```
In [17]: observation=df[['row_id','user_id']]
```

```
In [18]: prediction = logr.predict(observation)  
prediction
```

```
Out[18]: array([-1, -1, -1, ..., 16, 16, 16], dtype=int64)
```

```
In [19]: logr.classes_
```

```
Out[19]: array([-1, 0, 1, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 13, 14, 15, 16],  
dtype=int64)
```

```
In [20]: logr.predict_proba(observation)[0][1]
```

```
Out[20]: 1.7263815682078809e-09
```

```
In [21]: from sklearn.linear_model import Ridge,Lasso
```

```
In [22]: x = df[['row_id','user_id']]  
y = df['gate_id']
```

```
In [23]: 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)
```

```
In [24]: rr=Ridge(alpha=10)  
rr.fit(x_train,y_train)  
rr.score(x_test,y_test)  
rr.score(x_train,y_train)
```

```
Out[24]: 0.005545384866390224
```

```
In [25]: from sklearn.linear_model import LinearRegression  
lr= LinearRegression()  
lr.fit(x_train,y_train)
```

```
Out[25]: LinearRegression()
```

```
In [26]: lr.intercept_
```

```
Out[26]: 7.29500831419383
```

In [27]:

```
coeff = pd.DataFrame(lr.coef_,x.columns,columns=['Co-efficient'])
coeff
```

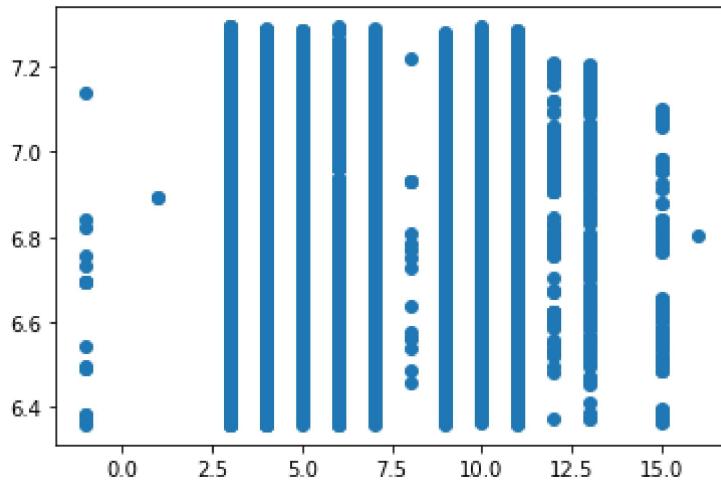
Out[27]:

|                | Co-efficient |
|----------------|--------------|
| <b>row_id</b>  | -0.000006    |
| <b>user_id</b> | -0.012988    |

In [28]:

```
prediction = lr.predict(x_test)
plt.scatter(y_test,prediction)
```

Out[28]: <matplotlib.collections.PathCollection at 0x1fa8049b9a0>



In [29]:

```
df['gate_id'].value_counts()
```

Out[29]:

|    |      |
|----|------|
| 4  | 8170 |
| 3  | 5351 |
| 10 | 4767 |
| 5  | 4619 |
| 11 | 4090 |
| 9  | 3390 |
| 7  | 3026 |
| 6  | 1800 |
| 13 | 1201 |
| 12 | 698  |
| 15 | 298  |
| -1 | 48   |
| 8  | 48   |
| 1  | 5    |
| 16 | 4    |
| 0  | 2    |
| 14 | 1    |

Name: gate\_id, dtype: int64

In [38]:

```
x=df[['row_id','user_id']]
y=df['gate_id']
```

```
In [39]: from sklearn.model_selection import train_test_split
x_train,x_test,y_train,y_test=train_test_split(x,y,train_size=0.70)
```

```
In [40]: from sklearn.ensemble import RandomForestClassifier
rfc = RandomForestClassifier()
rfc.fit(x_train,y_train)
```

```
Out[40]: RandomForestClassifier()
```

```
In [41]: parameters = {'max_depth':[1,2,3,4,5], 'min_samples_leaf':[5,10,15,20,25], 'n_estimators':
```

```
In [42]: from sklearn.model_selection import GridSearchCV
grid_search = GridSearchCV(estimator=rfc, param_grid= parameters, cv=2, scoring = "accuracy")
grid_search.fit(x_train,y_train)
```

```
Out[42]: GridSearchCV(cv=2, estimator=RandomForestClassifier(),
param_grid={'max_depth': [1, 2, 3, 4, 5],
'min_samples_leaf': [5, 10, 15, 20, 25],
'n_estimators': [10, 20, 30, 40, 50]},
scoring='accuracy')
```

```
In [43]: grid_search.best_score_
```

```
Out[43]: 0.22355494631025818
```

```
In [44]: rfc_best = grid_search.best_estimator_
```

```
In [46]: from sklearn.tree import plot_tree
plt.figure(figsize = (80,40))
plot_tree(rfc_best.estimators_[5],feature_names=x.columns,class_names = ['Yes','No','Ye
```

```
Out[46]: [Text(1915.2972972972973, 1993.2, 'row_id <= 16935.5\nngini = 0.871\nsamples = 16517\nvalue = [39, 1, 1, 3724, 5730, 3363, 1235, 2178, 37, 2368\nn3232, 2824, 501, 832, 197, 0]\nclass = Yes'),
Text(633.4054054054054, 1630.8000000000002, 'row_id <= 96.5\nngini = 0.876\nsamples = 7424\nvalue = [19, 1, 0, 1688, 2510, 1515, 640, 992, 8, 944, 1293\nn1164, 377, 588, 54, 0]\nclass = Yes'),
Text(241.2972972972973, 1268.4, 'user_id <= 27.5\nngini = 0.799\nsamples = 38\nvalue = [0, 0, 0, 5, 1, 15, 0, 9, 0, 20, 9, 3, 1, 0\nn1, 0]\nclass = No'),
Text(120.64864864864865, 906.0, 'gini = 0.744\nsamples = 23\nvalue = [0, 0, 0, 2, 1, 9, 0, 4, 0, 17, 4, 1, 1, 0\nn1, 0]\nclass = No'),
Text(361.94594594594594, 906.0, 'gini = 0.812\nsamples = 15\nvalue = [0, 0, 0, 3, 0, 6, 0, 5, 0, 3, 5, 2, 0, 0\nn0, 0]\nclass = No'),
Text(1025.5135135135135, 1268.4, 'user_id <= 52.5\nngini = 0.876\nsamples = 7386\nvalue = [19, 1, 0, 1683, 2509, 1500, 640, 983, 8, 924, 1284\nn1161, 376, 588, 53, 0]\nclass = Yes'),
Text(603.2432432432432, 906.0, 'user_id <= 19.5\nngini = 0.878\nsamples = 6252\nvalue = [14, 1, 0, 1178, 2075, 1416, 472, 839, 8, 869, 1081\nn994, 361, 555, 53, 0]\nclass = Yes'),
Text(361.94594594594594, 543.5999999999999, 'user_id <= 16.5\nngini = 0.879\nsamples = 3077\nvalue = [4, 0, 0, 495, 1028, 741, 182, 406, 1, 354, 512\nn462, 298, 419, 0, 0]\nclass = Yes'),
```

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Text(241.2972972972973, 181.19999999999982, 'gini = 0.86\nsamples = 2232\nvalue = [4, 0, 0, 376, 828, 647, 152, 330, 1, 279, 411\n389, 47, 119, 0, 0]\nnclass = Yes'),
Text(482.5945945945946, 181.19999999999982, 'gini = 0.86\nsamples = 845\nvalue = [0, 0, 0, 119, 200, 94, 30, 76, 0, 75, 101, 73\n251, 300, 0, 0]\nnclass = No'),
Text(844.5405405405405, 543.59999999999999, 'row_id <= 6864.0\nngini = 0.873\nsamples = 3175\nvalue = [10, 1, 0, 683, 1047, 675, 290, 433, 7, 515, 569\n532, 63, 136, 53, 0]\nnclass = Yes'),
Text(723.8918918918919, 181.19999999999982, 'gini = 0.875\nsamples = 1255\nvalue = [2, 0, 0, 313, 410, 236, 142, 166, 0, 145, 205\n224, 47, 64, 28, 0]\nnclass = Yes'),
Text(965.1891891891892, 181.19999999999982, 'gini = 0.87\nsamples = 1920\nvalue = [8, 1, 0, 370, 637, 439, 148, 267, 7, 370, 364\n308, 16, 72, 25, 0]\nnclass = Yes'),
Text(1447.7837837837837, 906.0, 'row_id <= 16422.0\nngini = 0.826\nsamples = 1134\nvalue = [5, 0, 0, 505, 434, 84, 168, 144, 0, 55, 203, 167\n15, 33, 0, 0]\nnclass = No'),
Text(1327.135135135135, 543.59999999999999, 'row_id <= 231.0\nngini = 0.827\nsamples = 1105\nvalue = [5, 0, 0, 497, 406, 84, 164, 144, 0, 55, 197, 163\n15, 33, 0, 0]\nnclass = No'),
Text(1206.4864864864865, 181.19999999999982, 'gini = 0.544\nsamples = 16\nvalue = [0, 0, 0, 15, 2, 0, 1, 3, 0, 0, 1, 1, 0\n0, 0]\nnclass = No'),
Text(1447.7837837837837, 181.19999999999982, 'gini = 0.829\nsamples = 1089\nvalue = [5, 0, 0, 482, 404, 84, 163, 141, 0, 55, 196, 162\n15, 33, 0, 0]\nnclass = No'),
Text(1568.4324324324325, 543.59999999999999, 'gini = 0.634\nsamples = 29\nvalue = [0, 0, 0, 8, 28, 0, 4, 0, 0, 0, 6, 4, 0, 0\n0, 0]\nnclass = Yes'),
Text(3197.189189189189, 1630.8000000000002, 'user_id <= 16.0\nngini = 0.865\nsamples = 9093\nvalue = [20, 0, 1, 2036, 3220, 1848, 595, 1186, 29, 1424\n1939, 1660, 124, 244, 143, 0]\nnclass = Yes'),
Text(2533.6216216216217, 1268.4, 'user_id <= 11.5\nngini = 0.853\nsamples = 3173\nvalue = [4, 0, 0, 384, 1161, 900, 149, 386, 4, 570, 726\n602, 26, 75, 26, 0]\nnclass = Yes'),
Text(2051.027027027027, 906.0, 'row_id <= 17565.5\nngini = 0.863\nsamples = 2038\nvalue = [4, 0, 0, 322, 705, 497, 134, 253, 0, 306, 503\n361, 26, 74, 26, 0]\nnclass = Yes'),
Text(1809.7297297297296, 543.59999999999999, 'row_id <= 17428.5\nngini = 0.843\nsamples = 50\nvalue = [0, 0, 0, 16, 6, 1, 8, 9, 0, 17, 9, 7, 0, 1\n0, 0]\nnclass = No'),
Text(1689.081081081, 181.19999999999982, 'gini = 0.835\nsamples = 35\nvalue = [0, 0, 0, 8, 4, 0, 6, 5, 0, 15, 8, 7, 0, 1\n0, 0]\nnclass = No'),
Text(1930.3783783783783, 181.19999999999982, 'gini = 0.765\nsamples = 15\nvalue = [0, 0, 0, 8, 2, 1, 2, 4, 0, 2, 1, 0, 0, 0\n0, 0]\nnclass = No'),
Text(2292.324324324324, 543.59999999999999, 'row_id <= 18138.0\nngini = 0.861\nsamples = 1988\nvalue = [4, 0, 0, 306, 699, 496, 126, 244, 0, 289, 494\n354, 26, 73, 26, 0]\nnclass = Yes'),
Text(2171.675675675676, 181.19999999999982, 'gini = 0.766\nsamples = 58\nvalue = [0, 0, 0, 8, 37, 11, 3, 8, 0, 2, 7, 7, 2, 1\n0, 0]\nnclass = Yes'),
Text(2412.972972972973, 181.19999999999982, 'gini = 0.863\nsamples = 1930\nvalue = [4, 0, 0, 298, 662, 485, 123, 236, 0, 287, 487\n347, 24, 72, 26, 0]\nnclass = Yes'),
Text(3016.2162162162163, 906.0, 'user_id <= 13.0\nngini = 0.825\nsamples = 1135\nvalue = [0, 0, 0, 62, 456, 403, 15, 133, 4, 264, 223, 241\n0, 1, 0, 0]\nnclass = Yes'),
Text(2774.9189189189187, 543.59999999999999, 'row_id <= 27965.5\nngini = 0.833\nsamples = 466\nvalue = [0, 0, 0, 52, 195, 143, 4, 52, 4, 96, 89, 110\n0, 1, 0, 0]\nnclass = Yes'),
Text(2654.27027027027, 181.19999999999982, 'gini = 0.83\nsamples = 273\nvalue = [0, 0, 0, 39, 116, 73, 2, 20, 0, 67, 60, 60, 0\n0, 0, 0]\nnclass = Yes'),
Text(2895.5675675675675, 181.19999999999982, 'gini = 0.827\nsamples = 193\nvalue = [0, 0, 0, 13, 79, 70, 2, 32, 4, 29, 29, 50, 0\n1, 0, 0]\nnclass = Yes'),
Text(3257.5135135135133, 543.59999999999999, 'row_id <= 29325.0\nngini = 0.815\nsamples = 669\nvalue = [0, 0, 0, 10, 261, 260, 11, 81, 0, 168, 134, 131\n0, 0, 0]\nnclass = Yes'),
Text(3136.864864864865, 181.19999999999982, 'gini = 0.807\nsamples = 458\nvalue = [0, 0, 0, 1, 192, 155, 10, 41, 0, 154, 85, 85\n0, 0, 0]\nnclass = Yes'),
Text(3378.162162162162, 181.19999999999982, 'gini = 0.8\nsamples = 211\nvalue = [0, 0, 0, 9, 69, 105, 1, 40, 0, 14, 49, 46, 0\n0, 0, 0]\nnclass = No'),
Text(3860.7567567567567, 1268.4, 'row_id <= 16983.0\nngini = 0.865\nsamples = 5920\nvalue = [16, 0, 1, 1652, 2059, 948, 446, 800, 25, 854, 1213\n1058, 98, 169, 117, 0]\nnclass = Yes'),
Text(3740.108108108108, 906.0, 'gini = 0.365\nsamples = 18\nvalue = [0, 0, 0, 0, 25, 0, 0, 0, 3, 0, 4, 0, 0\n0, 0]\nnclass = Yes'),
Text(3981.4054054054054, 906.0, 'row_id <= 25235.0\nngini = 0.865\nsamples = 5902\nvalue = [16, 0, 1, 1652, 2034, 948, 446, 800, 25, 851, 1213\n1054, 98, 169, 117, 0]\nnclass = Yes'),

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Text(3740.108108108108, 543.5999999999999, 'row_id <= 17090.5\nngini = 0.871\nnsamples = 2368\nvalue = [2, 0, 0, 676, 741, 369, 212, 327, 6, 306, 533\nn432, 58, 110, 40, 0]\nclass = Yes'),  
Text(3619.459459459459, 181.1999999999982, 'gini = 0.825\nnsamples = 38\nvalue = [0, 0, 0, 14, 3, 10, 1, 11, 0, 16, 14, 0, 0\nn0, 2, 0]\nclass = No'),  
Text(3860.7567567567567, 181.1999999999982, 'gini = 0.87\nnsamples = 2330\nvalue = [2, 0, 0, 662, 738, 359, 211, 316, 6, 290, 519\nn432, 58, 110, 38, 0]\nclass = Yes'),  
Text(4222.7027027027025, 543.5999999999999, 'user_id <= 34.5\nngini = 0.86\nnsamples = 3534\nvalue = [14, 0, 1, 976, 1293, 579, 234, 473, 19, 545, 680\nn622, 40, 59, 77, 0]\nclass = Yes'),  
Text(4102.054054054054, 181.1999999999982, 'gini = 0.858\nnsamples = 1383\nvalue = [5, 0, 1, 481, 452, 170, 117, 211, 10, 173, 262\nn255, 15, 18, 29, 0]\nclass = No'),  
Text(4343.351351351352, 181.1999999999982, 'gini = 0.858\nnsamples = 2151\nvalue = [9, 0, 0, 495, 841, 409, 117, 262, 9, 372, 418\nn367, 25, 41, 48, 0]\nclass = Yes')
```

