

mk 02-09-2023

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

```
In [77]: 1 from sklearn.linear_model import LogisticRegression
        2 a=pd.read_csv(r"C:\USERS\user\Downloads\C3_bot_detection_data (1).csv")
        3 a
```

Out[77]:

	User ID	Username	Tweet	Retweet Count	Mention Count	Follower Count	Verified	Bot Label	
0	132131	flong	Station activity person against natural majori...	85	1	2353	False	1	A
1	289683	hinesstephanie	Authority research natural life material staff...	55	5	9617	True	0	Se
2	779715	roberttran	Manage whose quickly especially foot none to	6	2	4363	True	0	Ha

```
In [78]: 1 from sklearn.linear_model import LogisticRegression
```

In [79]:

1

a=a.head(10)

2

a

1	289683	hinesstephanie	Authority research natural life material staff...	55	5	9617	True	0	Sander
2	779715	roberttran	Manage whose quickly especially foot none to g...	6	2	4363	True	0	Harrisc
3	696168	pmason	Just cover eight opportunity strong policy which.	54	5	2242	True	1	Martinez
4	704441	noah87	Animal sign six data good or.	26	3	8438	False	1	Camacho

In [81]:

1

a.columns

Out[81]:

Index(['User ID', 'Username', 'Tweet', 'Retweet Count', 'Mention Count', 'Follower Count', 'Verified', 'Bot Label', 'Location', 'Created At', 'Hashtags'], dtype='object')

In [83]:

1

b=a[['User ID', 'Retweet Count', 'Mention Count', 'Follower Count']]

2

b

Out[83]:

	User ID	Retweet Count	Mention Count	Follower Count
0	132131	85	1	2353
1	289683	55	5	9617
2	779715	6	2	4363
3	696168	54	5	2242
4	704441	26	3	8438
5	570928	41	4	3792
6	734182	54	0	10
7	107312	64	0	1442
8	549888	25	2	836
9	117640	67	3	6523

```
In [84]: 1 c=b.iloc[:,0:11]
        2 d=b.iloc[:, -1]
```

```
In [85]: 1 c.shape
```

```
Out[85]: (10, 4)
```

```
In [86]: 1 d.shape
```

```
Out[86]: (10,)
```

```
In [ ]: 1
```

```
In [87]: 1 from sklearn.preprocessing import StandardScaler
        2 fs=StandardScaler().fit_transform(c)
        3 fs
```

```
Out[87]: array([[ -1.27874417,  1.67954984, -0.8588975 , -0.51903859],
                [ -0.67927374,  0.32870546,  1.43149584,  1.82479847],
                [  1.18525156, -1.87767368, -0.28629917,  0.12951765],
                [  0.86736314,  0.28367732,  1.43149584, -0.55485438],
                [  0.89884112, -0.97711076,  0.28629917,  1.44437668],
                [  0.39083682, -0.30168858,  0.8588975 , -0.05472395],
                [  1.01200281,  0.28367732, -1.43149584, -1.2750422 ],
                [-1.37317811,  0.73395878, -1.43149584, -0.81298621],
                [  0.31078161, -1.02213891, -0.28629917, -1.00852108],
                [-1.33388105,  0.86904321,  0.28629917,  0.8264736 ]])
```

```
In [88]: 1 logr=LogisticRegression()
        2 logr.fit(fs,d)
```

```
Out[88]: LogisticRegression()
```

```
In [129]: 1 e=[[2,5,77,8]]
```

```
In [130]: 1 prediction=logr.predict(e)
        2 prediction
```

```
Out[130]: array([2242], dtype=int64)
```

```
In [131]: 1 logr.classes_
```

```
Out[131]: array([ 10,  836, 1442, 2242, 2353, 3792, 4363, 6523, 8438, 9617],
                dtype=int64)
```

```
In [133]: 1 logr.predict_proba(e)[0][0]
```

```
Out[133]: 8.966898605558299e-48
```

```
In [134]: 1 import re
          2 from sklearn.datasets import load_digits
          3 import numpy as np
          4 import pandas as pd
          5 import matplotlib.pyplot as plt
          6 import seaborn as sns
```

```
In [135]: 1 from sklearn.linear_model import LogisticRegression
          2 from sklearn.model_selection import train_test_split
```

```
In [136]: 1 digits=load_digits()
          2 digits
```

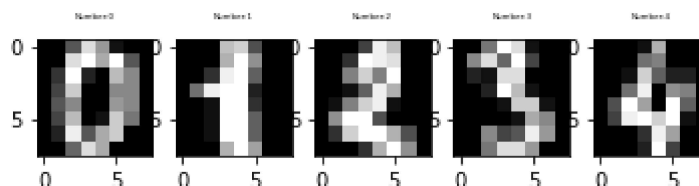
```
'pixel_0_6',
'pixel_0_7',
'pixel_1_0',
'pixel_1_1',
'pixel_1_2',
'pixel_1_3',
'pixel_1_4',
'pixel_1_5',
'pixel_1_6',
'pixel_1_7',
'pixel_2_0',
'pixel_2_1',
'pixel_2_2',
'pixel_2_3',
'pixel_2_4',
'pixel_2_5',
'pixel_2_6',
'pixel_2_7',
'pixel_3_0',
'pixel_3_1'
```

```
In [98]: 1 plt.figure(figsize=(20,4))
```

Out[98]: <Figure size 1440x288 with 0 Axes>

<Figure size 1440x288 with 0 Axes>

```
In [99]: 1 for index,(image,label) in enumerate(zip(digits.data[0:5],digits.target[0:
          2     plt.subplot(1,5,index+1)
          3     plt.imshow(np.reshape(image,(8
          4     ,8)),cmap=plt.cm.gray)
          5     plt.title('Number:%i\n'%label,fontsize=4)
```



```
In [100]: 1 x_train,x_test,y_train,y_test=train_test_split(digits.data,digits.target,t
```

```
In [101]: 1 print(x_train.shape)
2 print(x_test.shape)
3 print(y_train.shape)
4 print(y_test.shape)
```

```
(1257, 64)
(540, 64)
(1257,)
(540,)
```

```
In [102]: 1 logre=LogisticRegression(max_iter=10000)
2 logre.fit(x_train,y_train)
3
```

```
Out[102]: LogisticRegression(max_iter=10000)
```

```
In [103]: 1 print(logre.predict(x_test))
```

```
[7 4 5 4 5 6 7 6 4 6 7 5 7 7 7 9 4 7 6 5 8 0 7 1 6 2 7 4 7 7 6 7 3 1 2 9 7
 7 2 2 5 0 8 7 6 3 2 5 3 1 7 9 6 8 6 7 5 6 4 3 7 7 7 7 2 9 1 7 4 4 0 8 7 6
 0 6 1 3 5 5 2 9 4 0 2 3 3 4 9 4 8 1 4 2 8 8 0 9 1 6 1 4 9 0 7 9 1 3 3 0 7
 8 3 9 5 1 7 1 2 9 7 0 8 7 4 0 2 0 3 6 1 9 3 0 6 1 7 4 4 4 4 0 1 4 2 4 0 4
 4 6 7 3 6 8 6 8 8 6 1 0 9 9 6 8 2 9 5 2 5 0 7 3 9 6 1 5 0 9 6 9 1 7 2 4 0
 6 4 3 8 1 3 0 1 0 8 5 2 3 5 3 2 3 8 8 9 1 0 1 3 4 7 5 4 0 7 8 7 9 3 8 6 8
 7 3 5 2 7 1 4 4 5 3 0 6 4 5 9 3 5 1 1 8 9 8 7 3 0 8 4 1 9 4 8 4 3 7 6 7 9
 5 8 0 7 5 8 0 6 9 8 9 6 1 7 0 3 5 7 8 5 7 2 5 5 6 4 1 4 1 7 3 4 9 4 1 1 8
 9 6 5 9 2 8 3 5 9 6 1 4 2 8 0 7 7 8 3 3 8 8 9 9 0 8 8 4 4 8 6 5 2 6 6 7 0
 2 3 1 0 3 0 1 8 7 7 8 5 7 5 1 1 5 0 5 3 1 3 0 3 2 3 0 7 8 7 6 7 9 3 7 3 2
 1 4 6 7 7 8 8 7 5 8 4 5 0 4 9 3 2 9 8 1 7 4 6 5 4 0 4 8 0 5 0 1 5 4 6 9 8
 8 5 3 2 4 6 4 3 2 3 1 1 2 8 8 3 2 1 9 9 6 3 0 8 5 4 8 6 3 4 9 0 4 5 9 0 4
 5 6 2 8 1 2 5 6 2 8 9 2 7 0 5 3 4 8 3 8 1 1 0 1 4 5 1 4 1 9 0 1 6 0 3 3 5
 2 3 9 2 5 6 1 9 3 3 3 1 9 3 3 6 7 1 3 0 0 0 7 2 7 0 5 2 1 3 4 7 5 9 8 9 8
 2 3 1 2 6 6 8 8 9 2 4 3 4 7 1 3 7 1 4 0 3 4]
```

```
In [104]: 1 import numpy as np
2 import pandas as pd
3 import matplotlib.pyplot as plt
4 import seaborn as sns
```

```
In [105]: 1 df=pd.read_csv(r"C:\USERS\user\Downloads\C3_bot_detection_data (1).csv")
```

```
In [140]: 1 df['Mention Count'].value_counts()
```

```
Out[140]: 4    8441
5    8436
2    8347
1    8294
3    8252
0    8230
Name: Mention Count, dtype: int64
```

```
In [142]: 1 g1={"Mention Count":{"Mention Count":1,'b':2}}
          2 df=df.replace(g1)
          3 print(df)
```

	User ID	Username \
0	132131	flong
1	289683	hinesstephanie
2	779715	robertttran
3	696168	pmason
4	704441	noah87
...
49995	491196	uberg
49996	739297	jessicamunoz
49997	674475	lynnccunningham
49998	167081	richardthompson
49999	311204	daniel29

	Tweet	Retweet Count \
0	Station activity person against natural majori...	85
1	Authority research natural life material staff...	55
2	Manage whose quickly especially foot none to g...	6
3	Just cover eight opportunity strong policy which.	54
4	Animal sign six data good or.	26
...
49995	Want but put card direction know miss former h...	64
49996	Provide whole maybe agree church respond most ...	18
49997	Bring different everyone international capital...	43
49998	Than about single generation itself seek sell ...	45
49999	Here morning class various room human true bec...	91

	Mention Count	Follower Count	Verified	Bot Label	Location
\					
0	1	2353	False	1	Adkinston
1	5	9617	True	0	Sanderston
2	2	4363	True	0	Harrisonfurt
3	5	2242	True	1	Martinezberg
4	3	8438	False	1	Camachoville
...
49995	0	9911	True	1	Lake Kimberlyburgh
49996	5	9900	False	1	Greenbury
49997	3	6313	True	1	Deborahfort
49998	1	6343	False	0	Stephenside
49999	4	4006	False	0	Novakberg

	Created At	Hashtags
0	2020-05-11 15:29:50	NaN
1	2022-11-26 05:18:10	both live
2	2022-08-08 03:16:54	phone ahead
3	2021-08-14 22:27:05	ever quickly new I
4	2020-04-13 21:24:21	foreign mention
...
49995	2023-04-20 11:06:26	teach quality ten education any
49996	2022-10-18 03:57:35	add walk among believe
49997	2020-07-08 03:54:08	onto admit artist first
49998	2022-03-22 12:13:44	star
49999	2022-12-03 06:11:07	home

[50000 rows x 11 columns]

```
In [143]: 1 from sklearn.model_selection import train_test_split
          2 x_train,x_test,y_train,y_test=train_test_split(x,y,train_size=0.70)
```

```
In [144]: 1 from sklearn.ensemble import RandomForestClassifier
```

```
In [145]: 1 rfc=RandomForestClassifier()
          2 rfc.fit(x_train,y_train)
```

Out[145]: RandomForestClassifier()

```
In [146]: 1 parameters={'max_depth':[1,2,3,4,5],
          2               'min_samples_leaf':[5,10,15,20,25],
          3               'n_estimators':[10,20,30,40,50]}
```

```
In [147]: 1 from sklearn.model_selection import GridSearchCV
```

```
In [148]: 1 grid_search=GridSearchCV(estimator=rfc,param_grid=parameters,cv=2,scoring=
          2 grid_search.fit(x_train,y_train)
```

Out[148]: 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 [149]: 1 grid_search.best_score_
```

Out[149]: 0.9467213114754098

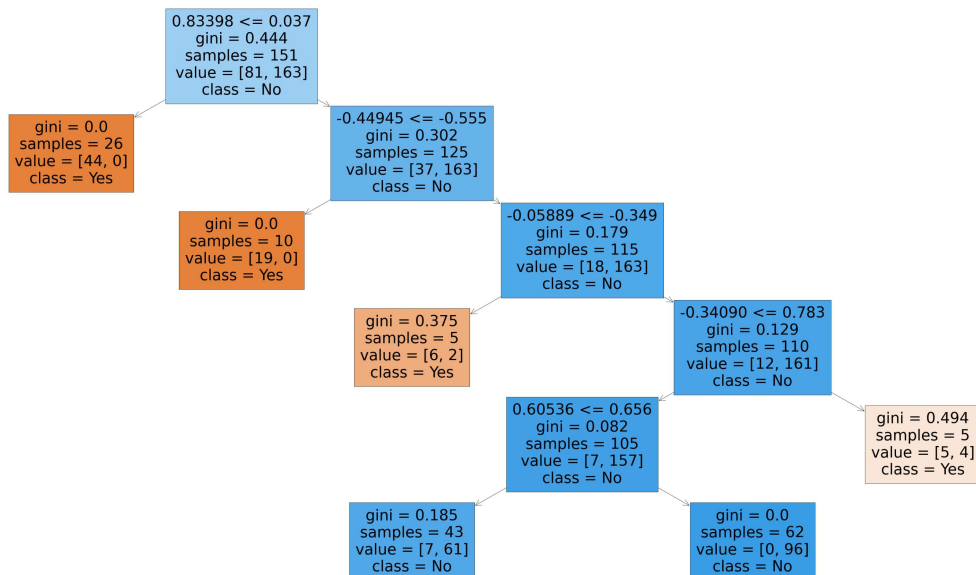
```
In [150]: 1 rfc_best=grid_search.best_estimator_
```

```
In [151]: 1 from sklearn.tree import plot_tree
```



```
In [152]: 1 plt.figure(figsize=(80,40))
          2 plot_tree(rfc_best.estimators_[5],feature_names=x.columns,class_names=['Ye
          3
```

```
Out[152]: [Text(1275.4285714285713, 1993.2, '0.83398 <= 0.037\ngini = 0.444\nsamples =
151\nvalue = [81, 163]\nclass = No'),
Text(637.7142857142857, 1630.8000000000002, 'gini = 0.0\nsamples = 26\nvalue
= [44, 0]\nclass = Yes'),
Text(1913.1428571428569, 1630.8000000000002, '-0.44945 <= -0.555\ngini = 0.3
02\nsamples = 125\nvalue = [37, 163]\nclass = No'),
Text(1275.4285714285713, 1268.4, 'gini = 0.0\nsamples = 10\nvalue = [19, 0]
\nclass = Yes'),
Text(2550.8571428571427, 1268.4, '-0.05889 <= -0.349\ngini = 0.179\nsamples
= 115\nvalue = [18, 163]\nclass = No'),
Text(1913.1428571428569, 906.0, 'gini = 0.375\nsamples = 5\nvalue = [6, 2]\n
class = Yes'),
Text(3188.5714285714284, 906.0, '-0.34090 <= 0.783\ngini = 0.129\nsamples =
110\nvalue = [12, 161]\nclass = No'),
Text(2550.8571428571427, 543.5999999999999, '0.60536 <= 0.656\ngini = 0.082
\nsamples = 105\nvalue = [7, 157]\nclass = No'),
Text(1913.1428571428569, 181.19999999999982, 'gini = 0.185\nsamples = 43\nva
lue = [7, 61]\nclass = No'),
Text(3188.5714285714284, 181.19999999999982, 'gini = 0.0\nsamples = 62\nvalu
e = [0, 96]\nclass = No'),
Text(3826.2857142857138, 543.5999999999999, 'gini = 0.494\nsamples = 5\nvalu
e = [5, 4]\nclass = Yes')]
```



```
In [ ]: 1
```