In [1]:

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
```

Logistic Regression

In [2]:

from sklearn.linear_model import LogisticRegression

```
In [3]:
```

df=pd.read_csv(r"C:\Users\user\Downloads\C3_bot_detection_data.csv")
df

Out[3]:

	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	
1	289683	hinesstephanie	Authority research natural life material staff	55	5	9617	True	0	S
2	779715	roberttran	Manage whose quickly especially foot none to g	6	2	4363	True	0	Н
3	696168	pmason	Just cover eight opportunity strong policy which.	54	5	2242	True	1	Ма
4	704441	noah87	Animal sign six data good or.	26	3	8438	False	1	Caı
49995	491196	uberg	Want but put card direction know miss former h	64	0	9911	True	1	Kiml
49996	739297	jessicamunoz	Provide whole maybe agree church respond most	18	5	9900	False	1	(
49997	674475	lynncunningham	Bring different everyone international capital	43	3	6313	True	1	D
49998	167081	richardthompson	Than about single generation itself seek sell	45	1	6343	False	0	St
49999	311204	daniel29	Here morning class various room human true bec	91	4	4006	False	0	1

50000 rows × 11 columns

```
In [4]:
df.columns
Out[4]:
Index(['User ID', 'Username', 'Tweet', 'Retweet Count', 'Mention Count',
       'Follower Count', 'Verified', 'Bot Label', 'Location', 'Created A
t',
       'Hashtags'],
      dtype='object')
In [5]:
df.info()
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 50000 entries, 0 to 49999
Data columns (total 11 columns):
#
                     Non-Null Count Dtype
     Column
     User ID
                     50000 non-null int64
 0
 1
     Username
                     50000 non-null
                                     object
                     50000 non-null
 2
     Tweet
                                     object
 3
     Retweet Count
                     50000 non-null
                                     int64
 4
     Mention Count
                     50000 non-null int64
 5
     Follower Count 50000 non-null int64
 6
     Verified
                     50000 non-null
                                     bool
 7
     Bot Label
                     50000 non-null int64
 8
     Location
                     50000 non-null
                                     object
9
     Created At
                     50000 non-null
                                     object
 10 Hashtags
                     41659 non-null
                                     object
dtypes: bool(1), int64(5), object(5)
memory usage: 3.9+ MB
In [6]:
feature_matrix=df[['Verified','User ID','Retweet Count','Mention Count','Follower Count',
target vector = df[['Verified']]
In [7]:
feature_matrix.shape
Out[7]:
(50000, 6)
In [8]:
```

(50000, 1)

Out[8]:

target vector.shape

```
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\utils\validation.py:63:
DataConversionWarning: A column-vector y was passed when a 1d array was ex
pected. Please change the shape of y to (n_samples, ), for example using r
avel().
  return f(*args, **kwargs)
Out[11]:
LogisticRegression()
In [12]:
observation=[[1.4,2,3,4,5,6]]
# Take Random Values
In [13]:
prediction = logr.predict(observation)
print(prediction)
[ True]
In [14]:
logr.classes_
Out[14]:
array([False, True])
In [15]:
logr.predict_proba(observation)[0][0]
Out[15]:
5.222004585858642e-06
In [16]:
logr.predict_proba(observation)[0][1]
Out[16]:
0.9999947779954141
```

Logistic Regression-2

In [17]:

```
import re
from sklearn.datasets import load_digits
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.linear_model import LogisticRegression
from sklearn.model_selection import train_test_split
```

In [18]:

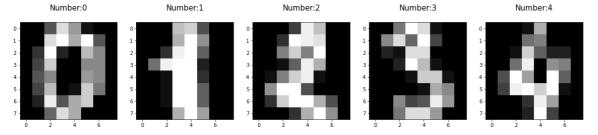
```
digits = load_digits()
digits
```

```
[ 0., 4., 16., ..., 16., 6., 0.],
[ 0., 8., 16., ..., 16., 8., 0.],
[ 0., 1., 8., ..., 12., 1., 0.]]]),
```

'DESCR': ".. _digits_dataset:\n\nOptical recognition of handwritten dig its dataset\n-----\n\n**Dat a Set Characteristics:**\n\n :Number of Instances: 1797\n of Attributes: 64\n :Attribute Information: 8x8 image of integer pixe ls in the range 0..16.\n :Missing Attribute Values: None\n r: E. Alpaydin (alpaydin '@' boun.edu.tr)\n :Date: July; 1998\n\nThis is a copy of the test set of the UCI ML hand-written digits datasets\nht tps://archive.ics.uci.edu/ml/datasets/Optical+Recognition+of+Handwritten +Digits\n\nThe data set contains images of hand-written digits: 10 class es where\neach class refers to a digit.\n\nPreprocessing programs made a vailable by NIST were used to extract\nnormalized bitmaps of handwritten digits from a preprinted form. From a\ntotal of 43 people, 30 contribute d to the training set and different 13\nto the test set. 32x32 bitmaps a re divided into nonoverlapping blocks of\n4x4 and the number of on pixel s are counted in each block. This generates\nan input matrix of 8x8 wher e each element is an integer in the range\n0..16. This reduces dimension

In [19]:

```
plt.figure(figsize=(20,4))
for index,(image,label) in enumerate(zip(digits.data[0:5],digits.target[0:5])):
    plt.subplot(1,5,index+1)
    plt.imshow(np.reshape(image,(8,8)),cmap=plt.cm.gray)
    plt.title('Number:%i\n'%label,fontsize=15)
```



```
In [20]:
```

```
x_train,x_test,y_train,y_test = train_test_split(digits.data,digits.target,test_size=0.30
```

In [21]:

```
print(x_train.shape)
print(x_test.shape)
print(y_train.shape)
print(y_test.shape)
```

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

In [22]:

```
logre=LogisticRegression(max_iter=10000)
logre.fit(x_train,y_train)
```

Out[22]:

LogisticRegression(max_iter=10000)

In [23]:

```
print(logre.predict(x_test))
```

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

In [24]:

```
print(logre.score(x_test,y_test))
```

0.9685185185185186