Phishing Detector using LR

The document has to specify the requirements for the project "Build a detector for Phishing websites (LR)." Apart from specifying the functional and non-functional requirements for the project, it also serves as an input for project scoping.

Importing the libraries

```
In [1]:
```

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import warnings
warnings.filterwarnings('ignore')
import seaborn as sns
```

Importing the dataset phishing.txt

```
In [2]:
data = pd.read_csv('phishing.txt',header =None)
```

Data understanding and Exploration

```
In [3]:
```

```
data.head()
```

Out[3]:

```
8 9 ... 21 22 23 24 25 26 27 28
                                                  30
      0
        1 -1 1 ...
                        1
                          -1
                                  0
                                                   -1
1 -1 -1 -1
                            1
                              -1
                                  1
                                     -1
                                                   -1
                    1
                        1
                          -1
                                   1
                                    -1
                                                  -1
1 -1 1 1 -1 1 ... -1
```

5 rows × 31 columns

```
In [4]:
```

```
data.shape
```

```
Out[4]:
```

(11055, 31)

Adding columns to the dataset

In [5]:

In [6]:

```
data.head()
```

Out[6]:

	having_IP_Address	URL_length	Shortining_Service	having_At_Symbol	double_slash_redire		
0	-1	1	1	1			
1	1	1	1	1			
2	1	0	1	1			
3	1	0	1	1			
4	1	0	-1	1			
5 rows × 31 columns							

In [7]:

data.columns

Out[7]:

```
Index(['having_IP_Address', 'URL_length', 'Shortining Service',
       'having At Symbol', 'double slash redirecting', 'Prefix suffi
x',
       'having_Sub_Domain', 'SSLfinal_State', 'Domain_registration_l
ength',
       'Favion', 'Port', 'HTTPS token', 'Request URL', 'URL of Ancho
r',
       'Links in tags', 'SFH', 'Submitting to email', 'Abnormal UR
L',
       'Redirect', 'on_movesover', 'RightClick', 'PopUpwindow', 'IFr
ame',
       'age of domain', 'DNSRecord', 'web_traffic', 'Page_Rank',
       'Google_Index', 'Links_pointing_to_page', 'Statistical_Repor
t',
       'Result'],
      dtype='object')
```

In [8]:

```
data.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 11055 entries, 0 to 11054
Data columns (total 31 columns):
having IP Address
                               11055 non-null int64
URL length
                               11055 non-null int64
Shortining Service
                               11055 non-null int64
having At Symbol
                               11055 non-null int64
double slash redirecting
                               11055 non-null int64
Prefix suffix
                               11055 non-null int64
having Sub Domain
                               11055 non-null int64
                               11055 non-null int64
SSLfinal State
Domain registration length
                               11055 non-null int64
Favion
                               11055 non-null int64
Port.
                               11055 non-null int64
HTTPS token
                               11055 non-null int64
Request URL
                               11055 non-null int64
URL of Anchor
                               11055 non-null int64
Links in tags
                               11055 non-null int64
SFH
                               11055 non-null int64
Submitting_to_email
                               11055 non-null int64
                               11055 non-null int64
Abnormal URL
Redirect
                               11055 non-null int64
on movesover
                               11055 non-null int64
                               11055 non-null int64
RightClick
PopUpwindow
                               11055 non-null int64
IFrame
                               11055 non-null int64
age of domain
                               11055 non-null int64
DNSRecord
                               11055 non-null int64
                               11055 non-null int64
web traffic
Page Rank
                               11055 non-null int64
Google Index
                               11055 non-null int64
                               11055 non-null int64
Links pointing to page
Statistical Report
                               11055 non-null int64
Result
                               11055 non-null int64
dtypes: int64(31)
memory usage: 2.6 MB
```

In [9]:

data.describe()

Out[9]:

	having_IP_Address	URL_length	Shortining_Service	having_At_Symbol	double_slash_re
count	11055.000000	11055.000000	11055.000000	11055.000000	110
mean	0.313795	-0.633198	0.738761	0.700588	
std	0.949534	0.766095	0.673998	0.713598	
min	-1.000000	-1.000000	-1.000000	-1.000000	
25%	-1.000000	-1.000000	1.000000	1.000000	
50%	1.000000	-1.000000	1.000000	1.000000	
75%	1.000000	-1.000000	1.000000	1.000000	
max	1.000000	1.000000	1.000000	1.000000	

8 rows × 31 columns

In [10]:

data.corr()

Out[10]:

	having_IP_Address	URL_length	Shortining_Service	having_At_Syml
having_IP_Address	1.000000	-0.052411	0.403461	0.1586
URL_length	-0.052411	1.000000	-0.097881	-0.0751
Shortining_Service	0.403461	-0.097881	1.000000	0.1044
having_At_Symbol	0.158699	-0.075108	0.104447	1.0000
double_slash_redirecting	0.397389	-0.081247	0.842796	0.0869
Prefix_suffix	-0.005257	0.055247	-0.080471	-0.0117
having_Sub_Domain	-0.080745	0.003997	-0.041916	-0.0589
SSLfinal_State	0.071414	0.048754	-0.061426	0.0312
Domain_registration_length	-0.022739	-0.221892	0.060923	0.0155
Favion	0.087025	-0.042497	0.006101	0.3048
Port	0.060979	0.000323	0.002201	0.3648
HTTPS_token	0.363534	-0.089383	0.757838	0.1045
Request_URL	0.029773	0.246348	-0.037235	0.0279
URL_of_Anchor	0.099847	-0.023396	0.000561	0.0579
Links_in_tags	0.006212	0.052869	-0.133379	-0.0708
SFH	-0.010962	0.414196	-0.022723	-0.0086
Submitting_to_email	0.077989	-0.014457	0.049328	0.3701
Abnormal_URL	0.336549	-0.106761	0.739290	0.2039
Redirect	-0.321181	0.046832	-0.534530	-0.0281
on_movesover	0.084059	-0.045103	0.062383	0.2796
RightClick	0.042881	-0.013613	0.038118	0.2195
PopUpwindow	0.096882	-0.049381	0.036616	0.2908
IFrame	0.054694	-0.013838	0.016581	0.2844
age_of_domain	-0.010446	0.179426	-0.052596	-0.0054
DNSRecord	-0.050733	-0.040823	0.436064	-0.0478
web_traffic	0.002922	0.008993	-0.047074	0.0329
Page_Rank	-0.091774	0.183518	0.014591	-0.0647
Google_Index	0.029153	0.002902	0.155844	0.0370
Links_pointing_to_page	-0.339065	-0.022987	-0.198410	-0.0060
Statistical_Report	-0.019103	-0.067153	0.085461	-0.0803
Result	0.094160	0.057430	-0.067966	0.0529

31 rows × 31 columns

```
In [11]:
sns.heatmap(data.corr(),cmap ='viridis',linecolor = 'black')
Out[11]:
<matplotlib.axes. subplots.AxesSubplot at 0x1a24eda390>
            having_IP_Address
                                                                                                        0.9
            Shortining Service
     double_slash_redirecting
          having_Sub_Domain
                                                                                                        0.6
 Domain_registration_length
                             Port
                  Request_URL
                                                                                                        0.3
                  Links_in_tags
          Submitting_to_email
                        Redirect
                                                                                                        0.0
                      RightClick
                         IFrame
                    DNSRecord
                     Page Rank
      Links pointing to page
                                      Shortining_Service
                                              having Sub Domain
                                                              Links in tags
                                                                                          Links_pointing_to_page
                                   having IP Address
                                           double slash redirecting
                                                  Domain registration length
                                                          Request URL
                                                                  Submitting to email
                                                                          RightClick
                                                                                  DNSRecord
                                                                                      Page_Rank
```

Splitting the dataset into independent variables and dependent variables

```
In [12]:

X = data.iloc[:,0:30].values

In [13]:

X.shape

Out[13]:
(11055, 30)

In [14]:

y = data.loc[:,['Result']].values

In [15]:

y.shape

Out[15]:
(11055, 1)
```

Spitting the dataset into training and testing dataset with ratio 70:30

70% of the dataset is goes for training 30% of the dataset is goes for testing

```
In [16]:

# Importing the train_test_split
from sklearn.model_selection import train_test_split
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size = 0.30, random_stat e = 1)
```

In [17]:

```
print(X_train.shape)
print(X_test.shape)
print(y_train.shape)
print(y_test.shape)

(7738, 30)
(3317, 30)
(7738, 1)
```

Logistic Regression Model

In [18]:

(3317, 1)

```
# Importing the classifier from linear model
from sklearn.linear_model import LogisticRegression
classifier = LogisticRegression()
```

In [19]:

```
classifier.get_params().keys()
```

Out[19]:

```
dict_keys(['C', 'class_weight', 'dual', 'fit_intercept', 'intercept_
scaling', 'max_iter', 'multi_class', 'n_jobs', 'penalty', 'random_st
ate', 'solver', 'tol', 'verbose', 'warm_start'])
```

In [20]:

```
# applying grid search to find best performing parameters
from sklearn.model selection import GridSearchCV
parameters = [\{'C':[0.001,0.01,0.1,1,10,100,1000],
                'penalty':['11','12']
              }]
grid search = GridSearchCV(classifier,parameters,cv=5,n jobs =-1)
grid search.fit(X train,y train)
# Printing best parameters
print('Best Accuracy =',(grid search.best score ))
print('Best parameters = ',(grid search.best params ))
Best Accuracy = 0.9280175756009305
Best parameters = {'C': 0.1, 'penalty': '11'}
/anaconda3/lib/python3.6/site-packages/sklearn/utils/validation.py:7
61: DataConversionWarning: A column-vector y was passed when a 1d ar
ray was expected. Please change the shape of y to (n samples, ), for
example using ravel().
  y = column or 1d(y, warn=True)
In [21]:
# Instantiate the classifier Logistic Regression
classifier = LogisticRegression(C=0.1,penalty ='11')
In [22]:
# Fitting the classifier or model on training dataset to train
classifier.fit(X train,y train)
/anaconda3/lib/python3.6/site-packages/sklearn/utils/validation.py:7
61: DataConversionWarning: A column-vector y was passed when a 1d ar
ray was expected. Please change the shape of y to (n samples, ), for
example using ravel().
  y = column or 1d(y, warn=True)
Out[22]:
LogisticRegression(C=0.1, class weight=None, dual=False, fit interce
pt=True,
          intercept scaling=1, max iter=100, multi class='warn',
          n jobs=None, penalty='l1', random state=None, solver='war
n',
          tol=0.0001, verbose=0, warm_start=False)
In [23]:
# Predicting the values on test dataset
y pred = classifier.predict(X test)
```

```
In [24]:
```

```
# Confusion matrix for the LR classifier
from sklearn.metrics import confusion matrix
cm = confusion matrix(y test,y pred)
cm
Out[24]:
array([[1338, 152],
       [ 91, 1736]])
In [25]:
TP = cm[0][0]
print('True Postive = ',TP)
FP = cm[0][1]
print('False Postive = ',FP)
FN = cm[1][0]
print('False Negative = ',FN)
TN = cm[1][1]
print('True Negative = ',TN)
True Postive = 1338
False Postive = 152
False Negative = 91
True Negative = 1736
In [26]:
from sklearn.metrics import accuracy score
accuracy = accuracy_score(y_test,y_pred)
print ('The Accuracy of the LR model : ',round(accuracy*100,ndigits =2),'%')
The Accuracy of the LR model: 92.67 %
```

Random Forest Classification Model

```
In [89]:
```

```
from sklearn.ensemble import RandomForestClassifier
classifier = RandomForestClassifier(n estimators =700,
                                   criterion ='entropy',
                                   max features ='sqrt',
                                   random state=0)
```

```
In [90]:
classifier.fit(X train,y train)
/anaconda3/lib/python3.6/site-packages/ipykernel launcher.py:1: Data
ConversionWarning: A column-vector y was passed when a 1d array was
expected. Please change the shape of y to (n samples,), for example
using ravel().
  """Entry point for launching an IPython kernel.
Out[90]:
RandomForestClassifier(bootstrap=True, class weight=None, criterion
='entropy',
            max depth=None, max features='sqrt', max leaf nodes=Non
e,
            min impurity decrease=0.0, min impurity split=None,
            min samples leaf=1, min samples split=2,
            min weight fraction leaf=0.0, n estimators=700, n jobs=N
one,
            oob score=False, random state=0, verbose=0, warm start=F
alse)
In [91]:
y pred = classifier.predict(X test)
In [92]:
from sklearn.metrics import confusion matrix
confusion_matrix(y_test,y_pred)
Out[92]:
array([[ 996, 494],
         3, 1824]])
       [
In [93]:
from sklearn.metrics import accuracy score
accuracy = accuracy score(y test,y pred)
accuracy
Out[93]:
0.8501658124811576
In [ ]:
```

XGBoost Model

```
In [32]:
```

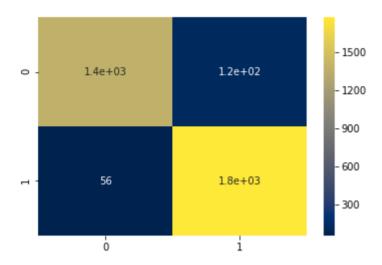
```
%time
# Importing the XGBoost model from scikit learn ensemble
from xgboost import XGBClassifier
classifier = XGBClassifier(n estimator =1000)
classifier.fit(X train,y train)
CPU times: user 4 \mus, sys: 1 \mus, total: 5 \mus
Wall time: 7.87 \mu s
/anaconda3/lib/python3.6/site-packages/sklearn/preprocessing/label.p
y:219: DataConversionWarning: A column-vector y was passed when a 1d
array was expected. Please change the shape of y to (n samples, ), f
or example using ravel().
  y = column or 1d(y, warn=True)
/anaconda3/lib/python3.6/site-packages/sklearn/preprocessing/label.p
y:252: DataConversionWarning: A column-vector y was passed when a 1d
array was expected. Please change the shape of y to (n samples, ), f
or example using ravel().
  y = column or 1d(y, warn=True)
Out[32]:
XGBClassifier(base score=0.5, booster='gbtree', colsample bylevel=1,
       colsample bytree=1, gamma=0, learning rate=0.1, max delta ste
p=0,
       max depth=3, min child weight=1, missing=None, n estimator=10
00.
       n estimators=100, n jobs=1, nthread=None,
       objective='binary:logistic', random state=0, reg alpha=0,
       reg lambda=1, scale pos weight=1, seed=None, silent=True,
       subsample=1)
In [33]:
# Predicting the values on independent variables testing dataset
y pred = classifier.predict(X test)
In [34]:
# Confusion matrix for evaluation to get the accuracy of the XGBoost model
from sklearn.metrics import confusion matrix
cm = confusion_matrix(y_test,y_pred)
cm
Out[34]:
array([[1370, 120],
       [ 56, 1771]])
```

In [35]:

```
sns.heatmap(cm,annot = True,cmap="cividis")
```

Out[35]:

<matplotlib.axes. subplots.AxesSubplot at 0x1a27074780>



In [36]:

```
# Accuracy of the XGBoost model is base on Actual values and predicting values b
y the model
from sklearn.metrics import accuracy_score
model_accuracy = accuracy_score(y_test,y_pred)
model_accuracy
```

Out[36]:

0.9469400060295448

The Champion Model out of LR,RF and XGBoost is Random Forest of accuracy 96%

Exercies 2

Train with only two input parameters - parameter Prefix_Suffix and 13 URL_of_Anchor. Check accuracy using the test data and compare the accuracy with the previous value.

```
In [37]:
```

```
data.head()
```

Out[37]:

	having_IP_Address	URL_length	Shortining_Service	having_At_Symbol	double_slash_redirect
0	-1	1	1	1	_
1	1	1	1	1	
2	1	0	1	1	
3	1	0	1	1	
4	1	0	-1	1	

5 rows × 31 columns

In [38]:

```
data.columns
```

```
Out[38]:
```

```
Index(['having_IP_Address', 'URL_length', 'Shortining_Service',
       'having At Symbol', 'double slash redirecting', 'Prefix suffi
x',
       'having Sub Domain', 'SSLfinal State', 'Domain registration 1
ength',
       'Favion', 'Port', 'HTTPS_token', 'Request_URL', 'URL_of_Ancho
r',
       'Links in tags', 'SFH', 'Submitting to email', 'Abnormal UR
L',
       'Redirect', 'on movesover', 'RightClick', 'PopUpwindow', 'IFr
ame',
       'age_of_domain', 'DNSRecord', 'web_traffic', 'Page_Rank',
       'Google Index', 'Links pointing to page', 'Statistical Repor
t',
       'Result'],
      dtype='object')
```

In [39]:

```
X1 = data.loc[:,['Prefix_suffix','URL_of_Anchor']].values
```

In [40]:

```
y1 = data.loc[:,['Result']].values
```

In [41]:

```
X = pd.DataFrame(X1)
```

In [42]:

```
y = pd.DataFrame(y)
```

Data understanding and Exploration

```
In [43]:
```

```
X.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 11055 entries, 0 to 11054
Data columns (total 2 columns):
0    11055 non-null int64
1    11055 non-null int64
dtypes: int64(2)
memory usage: 172.8 KB
```

In [44]:

```
X.describe().transpose()
```

Out[44]:

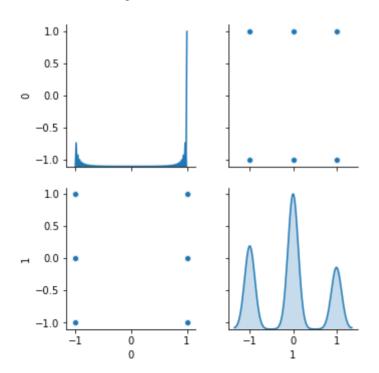
	count	mean	std	min	25%	50%	75%	max
0	11055.0	-0.734962	0.678139	-1.0	-1.0	-1.0	-1.0	1.0
1	11055.0	-0.076526	0.715138	-1.0	-1.0	0.0	0.0	1.0

In [45]:

```
sns.pairplot(X,diag_kind = 'kde')
```

Out[45]:

<seaborn.axisgrid.PairGrid at 0x1a27b662e8>

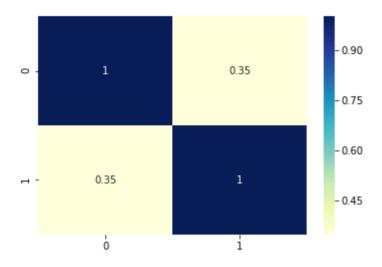


```
In [46]:
```

```
sns.heatmap(X.corr(),cmap = 'YlGnBu',annot =True)
```

Out[46]:

<matplotlib.axes._subplots.AxesSubplot at 0x1a27e20390>



In []:

```
In [47]:
```

```
y1 = y.values
```

Spitting the dataset into training and testing dataset with ratio 70:30

```
In [48]:
```

```
# Importing the train_test_split
from sklearn.model_selection import train_test_split
X_train, X_test, y_train, y_test = train_test_split(X1, y1, test_size =0.30, random_st ate =1)
```

In [49]:

```
print(X_train.shape)
print(X_test.shape)
print(y_train.shape)
print(y_test.shape)
```

```
(7738, 2)
```

(3317, 2)

(7738, 1)

(3317, 1)

Logistic Regression Model

```
In [66]:
```

```
# Importing the classifier from linear model
from sklearn.linear_model import LogisticRegression
```

In [67]:

```
# Instantiate the classifier Logistic Regression
classifier = LogisticRegression()
classifier.fit(X_train,y_train)
```

/anaconda3/lib/python3.6/site-packages/sklearn/utils/validation.py:7 61: DataConversionWarning: A column-vector y was passed when a 1d ar ray was expected. Please change the shape of y to (n_samples,), for example using ravel().

```
y = column_or_1d(y, warn=True)
```

Out[67]:

In [68]:

```
# Predicting the values on independent variables testing dataset
y_pred = classifier.predict(X_test)
```

In [69]:

```
# Confusion matrix for evaluation to get the accuracy of the XGBoost model
from sklearn.metrics import confusion_matrix
cm = confusion_matrix(y_test,y_pred)
cm
```

Out[69]:

In [70]:

```
TP = cm[0][0]
print('True Postive = ',TP)
FP = cm[0][1]
print('False Postive = ',FP)
FN = cm[1][0]
print('False Negative = ',FN)
TN = cm[1][1]
print('True Negative = ',TN)
```

```
True Postive = 996
False Postive = 494
False Negative = 3
True Negative = 1824
```

In [73]:

```
# Accuracy of the XGBoost model is base on Actual values and predicting values b
y the model
from sklearn.metrics import accuracy_score
model_accuracy = accuracy_score(y_test,y_pred)
model_accuracy
```

Out[73]:

0.8501658124811576

Random Forest Classification Model

```
In [72]:
```

In [74]:

```
classifier.fit(X_train,y_train)
```

/anaconda3/lib/python3.6/site-packages/ipykernel_launcher.py:1: Data ConversionWarning: A column-vector y was passed when a 1d array was expected. Please change the shape of y to (n_samples,), for example using ravel().

"""Entry point for launching an IPython kernel.

Out[74]:

In [75]:

```
# Predicting the values on independent variables testing dataset
y_pred = classifier.predict(X_test)
```

```
In [76]:
```

```
# Confusion matrix for evaluation to get the accuracy of the XGBoost model
from sklearn.metrics import confusion matrix
cm = confusion_matrix(y_test,y_pred)
cm
Out[76]:
array([[ 996, 494],
         3, 1824]])
       [
In [77]:
TP = cm[0][0]
print('True Postive = ',TP)
FP = cm[0][1]
print('False Postive = ',FP)
FN = cm[1][0]
print('False Negative = ',FN)
TN = cm[1][1]
print('True Negative = ',TN)
True Postive = 996
False Postive = 494
False Negative = 3
True Negative = 1824
In [94]:
# Accuracy of the XGBoost model is base on Actual values and predicting values b
y the model
from sklearn.metrics import accuracy_score
model_accuracy = accuracy_score(y_test,y_pred)
model accuracy
```

Out[94]:

0.8501658124811576

XGBoost Model

```
In [79]:
```

```
%time
# Importing the XGBoost model from scikit learn ensemble
from xgboost import XGBClassifier
classifier = XGBClassifier(n estimator =1000)
classifier.fit(X train,y train)
CPU times: user 4 \mus, sys: 1 \mus, total: 5 \mus
Wall time: 10 \mus
/anaconda3/lib/python3.6/site-packages/sklearn/preprocessing/label.p
y:219: DataConversionWarning: A column-vector y was passed when a 1d
array was expected. Please change the shape of y to (n samples, ), f
or example using ravel().
  y = column or 1d(y, warn=True)
/anaconda3/lib/python3.6/site-packages/sklearn/preprocessing/label.p
y:252: DataConversionWarning: A column-vector y was passed when a 1d
array was expected. Please change the shape of y to (n samples, ), f
or example using ravel().
  y = column or 1d(y, warn=True)
Out[79]:
XGBClassifier(base score=0.5, booster='gbtree', colsample bylevel=1,
       colsample bytree=1, gamma=0, learning rate=0.1, max delta ste
p=0,
       max depth=3, min child weight=1, missing=None, n estimator=10
00,
       n estimators=100, n jobs=1, nthread=None,
       objective='binary:logistic', random state=0, reg alpha=0,
       reg lambda=1, scale pos weight=1, seed=None, silent=True,
       subsample=1)
In [80]:
# Predicting the values on independent variables testing dataset
y pred = classifier.predict(X test)
In [81]:
# Confusion matrix for evaluation to get the accuracy of the XGBoost model
from sklearn.metrics import confusion matrix
cm = confusion matrix(y test,y pred)
cm
Out[81]:
array([[ 996, 494],
           3, 1824]])
       [
```

In [96]: # Accuracy of the XGBoost model is base on Actual values and predicting values b y the model from sklearn.metrics import accuracy_score accuracy = accuracy_score(y_test,y_pred) accuracy Out[96]: 0.8501658124811576 In []: In []: In []: