PART 1

In [64]: import numpy as np
 import pandas as pd
 from sklearn.ensemble import AdaBoostClassifier
 from sklearn.model_selection import train_test_split
 from sklearn.tree import DecisionTreeClassifier
 from sklearn.metrics import accuracy_score
 from sklearn import tree
 import matplotlib.pyplot as plt

In [65]: df = pd.read_csv('DataSetForPhishingVSBenignUrl.csv')

In [66]: df

Out[66]:

	Querylength	domain_token_count	path_token_count	avgdomaintokenlen	longdomaintokenler
0	0	4	5	5.500000	14
1	0	4	5	5.500000	14
2	0	4	5	5.500000	14
3	0	4	12	5.500000	14
4	0	4	6	5.500000	14
36702	29	4	14	5.750000	12
36703	0	4	13	3.750000	8
36704	58	3	27	6.666666	16
36705	35	3	13	4.333334	Ę
36706	40	3	25	6.666666	16

36707 rows × 80 columns

In [67]: #Formatting Data Frame to contain only Phishing and Benign
df = df.loc[(df['URL_Type_obf_Type'] == 'phishing') | (df['URL_Type_obf_Type']

Out[67]:

	Querylength	domain_token_count	path_token_count	avgdomaintokenlen	longdomaintokenler
7930	0	2	12	5.500000	3
7931	0	3	12	5.000000	10
7932	2	2	11	4.000000	5
7933	0	2	7	4.500000	7
7934	19	2	10	6.000000	Ę
30004	0	2	3	8.000000	13
30005	0	3	0	9.000000	16
30006	0	3	2	6.666666	10
30007	0	2	3	8.000000	13
30008	0	2	3	9.000000	15

15367 rows × 80 columns

```
In [68]: df = df.dropna()
```

In [69]: df

Out[69]:

	Querylength	domain_token_count	path_token_count	avgdomaintokenlen	longdomaintokenler
7930	0	2	12	5.500000	8
7931	0	3	12	5.000000	10
7934	19	2	10	6.000000	ç
7935	0	2	10	5.500000	ç
7938	0	2	9	2.500000	3
					•••
29978	0	2	5	5.500000	٤
29986	0	2	5	3.500000	5
29992	0	2	4	7.000000	12
29996	0	3	5	4.666666	10
29999	0	2	4	8.000000	13

6723 rows × 80 columns

```
In [70]: X= df.iloc[0:,0:79]
In [71]: X
Out[71]:
                  Querylength domain_token_count path_token_count avgdomaintokenlen longdomaintokenlen
            7930
                          0
                                             2
                                                            12
                                                                        5.500000
                                                                                                8
                                                            12
                          0
                                             3
                                                                        5.000000
            7931
                                                                                               10
            7934
                          19
                                             2
                                                            10
                                                                        6.000000
                                                                                                9
                                             2
                                                                        5.500000
            7935
                          0
                                                            10
                                                                                                9
                                             2
                          0
                                                             9
                                                                        2.500000
            7938
                                                                                                3
                                             •••
                                                                                                ...
           29978
                          0
                                             2
                                                             5
                                                                        5.500000
                                                                                                8
                                             2
                                                             5
                                                                        3.500000
           29986
                           0
                                                                                                5
           29992
                          0
                                             2
                                                             4
                                                                        7.000000
                                                                                               12
           29996
                           0
                                             3
                                                             5
                                                                        4.666666
                                                                                               10
                                             2
           29999
                          0
                                                             4
                                                                        8.000000
                                                                                               13
In [72]:
          y = df.iloc[0:, 79]
In [73]: y
Out[73]: 7930
                       benign
          7931
                       benign
          7934
                       benign
          7935
                       benign
          7938
                       benign
          29978
                     phishing
          29986
                     phishing
          29992
                     phishing
          29996
                     phishing
          29999
                     phishing
          Name: URL Type obf Type, Length: 6723, dtype: object
In [74]: X_train, X_test, y_train, y_test = train_test_split( X, y, test_size = 0.3,
In [75]: X train.shape
Out[75]: (4706, 79)
In [76]: y_train.shape
Out[76]: (4706,)
```

```
In [77]: # Decision Tree classifier with criteria - Gini Index
         tree qini = DecisionTreeClassifier(criterion = "gini", random state = 50)
         tree_gini.fit(X_train, y_train)
Out[77]: DecisionTreeClassifier(random_state=50)
In [52]: tree_gini.score(X_train,y_train)
Out[52]: 1.0
In [78]: tree gini.score(X test, y test)
Out[78]: 0.9514129895884977
In [79]: # Create a decision tree object
         tree gini = DecisionTreeClassifier()
         # Fit on training data
         tree gini.fit(X train, y train)
Out[79]: DecisionTreeClassifier()
In [80]: #Create adaboost classifer object
         abc = AdaBoostClassifier(base estimator=tree gini)
         # Train Adaboost Classifer
         model = abc.fit(X train, y train)
         #Predict the response with test dataset
         y pred = model.predict(X test)
         /Users/soniarahman/opt/anaconda3/lib/python3.8/site-packages/sklearn/bas
         e.py:441: UserWarning: X does not have valid feature names, but AdaBoostC
         lassifier was fitted with feature names
           warnings.warn(
In [81]: print("Accuracy:", accuracy score(y test, y pred))
         Accuracy: 0.9489340604858701
In [82]: from sklearn.model selection import GridSearchCV
In [83]: | tree param = {'criterion':['gini', 'entropy'],
                      'max depth':[1,3,6,9,12,15,18]
         grid = GridSearchCV(DecisionTreeClassifier(), tree param, cv=5)
         grid.fit(X train, y train)
Out[83]: GridSearchCV(cv=5, estimator=DecisionTreeClassifier(),
                      param grid={'criterion': ['gini', 'entropy'],
                                   'max depth': [1, 3, 6, 9, 12, 15, 18]})
In [84]: print(grid.best score )
         0.9617491443127539
```

In [85]:

Accuracy: 0.9781854238968766

print(grid.best params)

/Users/soniarahman/opt/anaconda3/lib/python3.8/site-packages/sklearn/bas e.py:441: UserWarning: X does not have valid feature names, but AdaBoostC lassifier was fitted with feature names warnings.warn(

#

Using adaboost, the decision tree obtained an accuracy score of 0.9781854238968766 for this set, which is much higher than the previous week's score of 0.969261279127417. We get a higher score when we use adaboost. In both assignments, the gini index is greater than the entropy. The adaboost classifier and the set parameters resulted in 15 as the max depth with the best parameter and estimator.