

```
[2]: import numpy as np
import pandas as pd
from IPython.core.interactiveshell import InteractiveShell

InteractiveShell.ast_node_interactivity = "all"

pd.set_option("display.max_columns", 100)
pd.set_option("display.max_rows", 100)

data = pd.read_csv('DataSetForPhishingVSBenignUrl.csv', header=0)

data.head(10)
```

```
[2]: Querylength  domain_token_count  path_token_count  avgdomaintokenlen  \
0           0           4           5           5.5
1           0           4           5           5.5
2           0           4           5           5.5
3           0           4          12           5.5
4           0           4           6           5.5
5           0           4           8           5.5
6           0           4           5           5.5
7           0           4           7           5.5
8           0           4           6           5.5
9           0           4           5           5.5

      longdomaintokenlen  avgpathtokenlen  tld  charcompvowels  charcompacc  \
0           14           4.400000      4           8           3
1           14           6.000000      4          12           4
2           14           5.800000      4          12           5
3           14           5.500000      4          32          16
4           14           7.333334      4          18          11
5           14           6.500000      4          22          10
6           14           7.800000      4          17          10
7           14           6.285714      4          16           9
8           14           6.500000      4          16          10
9           14           3.600000      4           7           3

      ldl_url  ldl_domain  ldl_path  ldl_filename  ldl_getArg  dld_url  \
0           0           0           0           0           0           0
1           0           0           0           0           0           0
2           0           0           0           0           0           0
3           0           0           0           0           0           0
4           0           0           0           0           0           0
5           0           0           0           0           0           0
6           0           0           0           0           0           0
7           0           0           0           0           0           0
8           0           0           0           0           0           0
9           0           0           0           0           0           0

      dld_domain  dld_path  dld_filename  dld_getArg  urlLen  domainlength  \
```

0	0	0	0	0	58	25
1	0	0	0	0	66	25
2	0	0	0	0	65	25
3	0	0	0	0	109	25
4	0	0	0	0	81	25
5	0	0	0	0	91	25
6	0	0	0	0	75	25
7	0	0	0	0	82	25
8	0	0	0	0	76	25
9	0	0	0	0	54	25

	pathLength	subDirLen	fileNameLen	this.fileExtLen	ArgLen	pathurlRatio	\
0	26	26	13	1	2	0.448276	
1	34	34	2	2	2	0.515151	
2	33	33	2	2	2	0.507692	
3	77	77	2	2	2	0.706422	
4	49	49	2	2	2	0.604938	
5	59	59	2	2	2	0.648352	
6	43	43	2	2	2	0.573333	
7	50	50	2	2	2	0.609756	
8	44	44	2	2	2	0.578947	
9	22	22	9	1	2	0.407407	

	ArgUrlRatio	argDomanRatio	domainUrlRatio	pathDomainRatio	argPathRatio	\
0	0.034483	0.08	0.431034	1.04	0.0769231	
1	0.030303	0.08	0.378788	1.36	0.0588235	
2	0.030769	0.08	0.384615	1.32	0.0606061	
3	0.018349	0.08	0.229358	3.08	0.025974	
4	0.024691	0.08	0.308642	1.96	0.0408163	
5	0.021978	0.08	0.274725	2.36	0.0338983	
6	0.026667	0.08	0.333333	1.72	0.0465116	
7	0.024390	0.08	0.304878	2.00	0.04	
8	0.026316	0.08	0.328947	1.76	0.0454545	
9	0.037037	0.08	0.462963	0.88	0.0909091	

	executable	isPortEighty	NumberofDotsinURL	ISIpAddressInDomainName	\
0	0	-1	5	-1	
1	0	-1	4	-1	
2	0	-1	4	-1	
3	0	-1	4	-1	
4	0	-1	4	-1	
5	0	-1	4	-1	
6	0	-1	4	-1	
7	0	-1	4	-1	
8	0	-1	4	-1	
9	0	-1	5	-1	

	CharacterContinuityRate	LongestVariableValue	URL_DigitCount	\
0	0.6	-1	1	
1	0.6	-1	0	
2	0.6	-1	0	

3	0.6	-1	0
4	0.6	-1	0
5	0.6	-1	0
6	0.6	-1	0
7	0.6	-1	8
8	0.6	-1	0
9	0.6	-1	1

	host_DigitCount	Directory_DigitCount	File_name_DigitCount	\
0	0	0	0	0
1	0	0	0	0
2	0	0	0	0
3	0	0	0	0
4	0	0	0	0
5	0	0	0	0
6	0	0	0	0
7	0	0	0	0
8	0	0	0	0
9	0	0	0	0

	Extension_DigitCount	Query_DigitCount	URL_Letter_Count	\
0	1	-1	47	
1	0	-1	56	
2	0	-1	55	
3	0	-1	92	
4	0	-1	70	
5	0	-1	78	
6	0	-1	65	
7	8	-1	62	
8	0	-1	65	
9	1	-1	43	

	host_letter_count	Directory_LetterCount	Filename_LetterCount	\
0	22	8	13	
1	22	8	13	
2	22	8	13	
3	22	8	13	
4	22	8	13	
5	22	8	13	
6	22	8	13	
7	22	8	13	
8	22	8	13	
9	22	8	9	

	Extension_LetterCount	Query_LetterCount	LongestPathTokenLength	\
0	0	-1	13	
1	9	-1	13	
2	8	-1	13	
3	45	-1	52	
4	23	-1	24	
5	31	-1	34	

6	18	-1	18
7	15	-1	25
8	18	-1	19
9	0	-1	9

	Domain_LongestWordLength	Path_LongestWordLength	\
0	14	13	
1	14	13	
2	14	13	
3	14	13	
4	14	13	
5	14	15	
6	14	18	
7	14	13	
8	14	13	
9	14	9	

	sub-Directory_LongestWordLength	Arguments_LongestWordLength	\
0	5	-1	
1	5	-1	
2	5	-1	
3	13	-1	
4	13	-1	
5	13	-1	
6	5	-1	
7	13	-1	
8	13	-1	
9	5	-1	

	URL_sensitiveWord	URLQueries_variable	spcharUrl	delimiter_Domain	\
0	0	0	3	0	
1	0	0	4	0	
2	0	0	4	0	
3	0	0	4	0	
4	0	0	4	0	
5	0	0	4	0	
6	0	0	4	0	
7	0	0	4	0	
8	0	0	4	0	
9	0	0	3	0	

	delimiter_path	delimiter_Count	NumberRate_URL	NumberRate_Domain	\
0	2	-1	0.017241	0.0	
1	1	-1	0.000000	0.0	
2	1	-1	0.000000	0.0	
3	8	-1	0.000000	0.0	
4	2	-1	0.000000	0.0	
5	4	-1	0.000000	0.0	
6	1	-1	0.000000	0.0	
7	3	-1	0.097561	0.0	
8	2	-1	0.000000	0.0	

9	2	-1	0.018519	0.0
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	NumberRate_DirectoryName	NumberRate_FileName	NumberRate_Extension	\
0	0.0	0.066667	1.0	
1	0.0	0.000000	NaN	
2	0.0	0.000000	NaN	
3	0.0	0.000000	NaN	
4	0.0	0.000000	NaN	
5	0.0	0.000000	NaN	
6	0.0	0.000000	NaN	
7	0.0	0.320000	NaN	
8	0.0	0.000000	NaN	
9	0.0	0.090909	1.0	

	NumberRate_AfterPath	SymbolCount_URL	SymbolCount_Domain	\
0	-1.0	8	3	
1	-1.0	8	3	
2	-1.0	8	3	
3	-1.0	8	3	
4	-1.0	8	3	
5	-1.0	8	3	
6	-1.0	8	3	
7	-1.0	8	3	
8	-1.0	8	3	
9	-1.0	8	3	

	SymbolCount_Directoryname	SymbolCount_FileName	SymbolCount_Extension	\
0	2	1	0	
1	3	0	0	
2	3	0	0	
3	3	0	0	
4	3	0	0	
5	3	0	0	
6	3	0	0	
7	3	0	0	
8	3	0	0	
9	2	1	0	

	SymbolCount_Afterpath	Entropy_URL	Entropy_Domain	Entropy_DirectoryName	\
0	-1	0.726298	0.784493	0.894886	
1	-1	0.688635	0.784493	0.814725	
2	-1	0.695049	0.784493	0.814725	
3	-1	0.640130	0.784493	0.814725	
4	-1	0.681307	0.784493	0.814725	
5	-1	0.666676	0.784493	0.814725	
6	-1	0.682440	0.784493	0.814725	
7	-1	0.709396	0.784493	0.814725	
8	-1	0.678242	0.784493	0.814725	
9	-1	0.740950	0.784493	0.894886	

Entropy_Filename	Entropy_Extension	Entropy_Afterpath	URL_Type	obf_Type
------------------	-------------------	-------------------	----------	----------

0	0.850608	NaN	-1.0	Defacement
1	0.859793	0.0	-1.0	Defacement
2	0.801880	0.0	-1.0	Defacement
3	0.663210	0.0	-1.0	Defacement
4	0.804526	0.0	-1.0	Defacement
5	0.755658	0.0	-1.0	Defacement
6	0.766719	0.0	-1.0	Defacement
7	0.797498	0.0	-1.0	Defacement
8	0.732258	0.0	-1.0	Defacement
9	0.894886	NaN	-1.0	Defacement

### Handle columns with Nulls in the current data set

```
[3]: data.isnull().sum()
```

```
[3]: Querylength          0
      domain_token_count   0
      path_token_count     0
      avgdomaintokenlen    0
      longdomaintokenlen   0
      avgpathtokenlen      280
      tld                  0
      charcompvowels       0
      charcomppace        0
      ldl_url              0
      ldl_domain           0
      ldl_path             0
      ldl_filename         0
      ldl_getArg           0
      dld_url              0
      dld_domain           0
      dld_path             0
      dld_filename         0
      dld_getArg           0
      urlLen               0
      domainlength         0
      pathLength           0
      subDirLen            0
      fileNameLen          0
      this.fileExtLen      0
      ArgLen               0
      pathurlRatio         0
      ArgUrlRatio          0
      argDomanRatio        0
      domainUrlRatio       0
      pathDomainRatio      0
      argPathRatio         0
      executable           0
      isPortEighty         0
      NumberofDotsinURL    0
      ISIpAddressInDomainName 0
```

CharacterContinuityRate	0
LongestVariableValue	0
URL_DigitCount	0
host_DigitCount	0
Directory_DigitCount	0
File_name_DigitCount	0
Extension_DigitCount	0
Query_DigitCount	0
URL_Letter_Count	0
host_letter_count	0
Directory_LetterCount	0
Filename_LetterCount	0
Extension_LetterCount	0
Query_LetterCount	0
LongestPathTokenLength	0
Domain_LongestWordLength	0
Path_LongestWordLength	0
sub-Directory_LongestWordLength	0
Arguments_LongestWordLength	0
URL_sensitiveWord	0
URLQueries_variable	0
spcharUrl	0
delimiter_Domain	0
delimiter_path	0
delimiter_Count	0
NumberRate_URL	0
NumberRate_Domain	0
NumberRate_DirectoryName	10
NumberRate_FileName	10
NumberRate_Extension	10130
NumberRate_AfterPath	3
SymbolCount_URL	0
SymbolCount_Domain	0
SymbolCount_Directoryname	0
SymbolCount_FileName	0
SymbolCount_Extension	0
SymbolCount_Afterpath	0
Entropy_URL	0
Entropy_Domain	0
Entropy_DirectoryName	8468
Entropy_Filename	236
Entropy_Extension	40
Entropy_Afterpath	6
URL_Type_obf_Type	0
dtype: int64	

```
[4]: data_clean_na = data.dropna()            #(subset=['Entropy_DirectoryName'])
```

```
[16]: lst = ['phishing' , 'benign']

data_clean = data_clean_na[data_clean_na.URL_Type_obf_Type.isin(lst) ]
```

```
data_clean['URL_Type_obf_Type'].unique()
```

```
[16]: array(['benign', 'phishing'], dtype=object)
```

```
[17]: data_clean.dtypes
```

```
[17]: Querylength                int64
      domain_token_count       int64
      path_token_count         int64
      avgdomaintokenlen        float64
      longdomaintokenlen       int64
      avgpathtokenlen          float64
      tld                      int64
      charcompvowels           int64
      charcomppace             int64
      ldl_url                  int64
      ldl_domain               int64
      ldl_path                 int64
      ldl_filename             int64
      ldl_getArg               int64
      dld_url                  int64
      dld_domain               int64
      dld_path                 int64
      dld_filename             int64
      dld_getArg               int64
      urlLen                   int64
      domainlength             int64
      pathLength               int64
      subDirLen                int64
      fileNameLen              int64
      this.fileExtLen          int64
      ArgLen                   int64
      pathurlRatio             float64
      ArgUrlRatio              float64
      argDomanRatio            float64
      domainUrlRatio           float64
      pathDomainRatio          float64
      argPathRatio             object
      executable               int64
      isPortEighty             int64
      NumberofDotsinURL        int64
      ISIpAddressInDomainName   int64
      CharacterContinuityRate   float64
      LongestVariableValue      int64
      URL_DigitCount           int64
      host_DigitCount           int64
      Directory_DigitCount      int64
      File_name_DigitCount      int64
      Extension_DigitCount      int64
      Query_DigitCount          int64
      URL_Letter_Count          int64
```



host_letter_count	int64
Directory_LetterCount	int64
Filename_LetterCount	int64
Extension_LetterCount	int64
Query_LetterCount	int64
LongestPathTokenLength	int64
Domain_LongestWordLength	int64
Path_LongestWordLength	int64
sub-Directory_LongestWordLength	int64
Arguments_LongestWordLength	int64
URL_sensitiveWord	int64
URLQueries_variable	int64
spcharUrl	int64
delimiter_Domain	int64
delimiter_path	int64
delimiter_Count	int64
NumberRate_URL	float64
NumberRate_Domain	float64
NumberRate_DirectoryName	float64
NumberRate_FileName	float64
NumberRate_Extension	float64
NumberRate_AfterPath	float64
SymbolCount_URL	int64
SymbolCount_Domain	int64
SymbolCount_Directoryname	int64
SymbolCount_FileName	int64
SymbolCount_Extension	int64
SymbolCount_Afterpath	int64
Entropy_URL	float64
Entropy_Domain	float64
Entropy_DirectoryName	float64
Entropy_Filename	float64
Entropy_Extension	float64
Entropy_Afterpath	float64
URL_Type_obf_Type	object
dtype:	object

```
[18]: # split into input and output elements
```

```
data_numpy = data_clean.values
X, y = data_numpy[:, :-1], data_numpy[:, -1]

#X = data_clean.drop('URL_Type_obf_Type',axis = 'columns')
#y = data_clean.URL_Type_obf_Type
```

## SMOT Oversampling the data to removed skewed classes

```
[19]: from sklearn.preprocessing import LabelEncoder
      from imblearn.over_sampling import SMOTE
      from collections import Counter
```

```
y = LabelEncoder().fit_transform(y)

# transform the dataset
oversample = SMOTE()
X, y = oversample.fit_resample(X, y)
# summarize distribution
```

```
[20]: from matplotlib import pyplot

counter = Counter(y)
for k,v in counter.items():
    per = v / len(y) * 100
    print('Class=%d, n=%d (%.3f%%)' % (k, v, per))
# plot the distribution
pyplot.bar(counter.keys(), counter.values())
pyplot.show()
```

Class=0, n=4014 (50.000%)

Class=1, n=4014 (50.000%)

```
[20]: <BarContainer object of 2 artists>
```

<Figure size 640x480 with 1 Axes>

## Test and Training Split and Scaling the data

```
[21]: from sklearn.model_selection import train_test_split
      from sklearn.preprocessing import StandardScaler

X_train, X_test, y_train, y_test = train_test_split(
    X, y, test_size=0.2, random_state=1, stratify=y)

sc = StandardScaler()
sc.fit(X_train)

X_train_std = sc.transform(X_train)
X_test_std = sc.transform(X_test)

print('Labels count in y:', np.bincount(y))
print('Labels count in y_train:', np.bincount(y_train))

print('Labels count in y_test:', np.bincount(y_test))
```

```
[21]: StandardScaler()
```

```
Labels count in y: [4014 4014]
```

```
Labels count in y_train: [3211 3211]
```

```
Labels count in y_test: [803 803]
```

## TRAIN MODEL & CREATE PREDICTIONS USING SKLEARN - GINI CRITERION

```
[73]: from sklearn.tree import DecisionTreeClassifier
      from sklearn.ensemble import AdaBoostClassifier
      from sklearn import metrics

      #developing a model with gini , tree and adboost

      clf_gini_d1 = DecisionTreeClassifier(criterion = 'gini' , max_depth = 1 , splitter =_
      ↪'best')
      clf_gini_d3 = DecisionTreeClassifier(criterion = 'gini' , max_depth = 3 , splitter =_
      ↪'best')
      clf_gini_d6 = DecisionTreeClassifier(criterion = 'gini' , max_depth = 6 , splitter =_
      ↪'best')
      clf_gini_d9 = DecisionTreeClassifier(criterion = 'gini' , max_depth = 9 , splitter =_
      ↪'best')
      clf_gini_d12 = DecisionTreeClassifier(criterion = 'gini' , max_depth = 12 , splitter =_
      ↪'best')
      clf_gini_d15 = DecisionTreeClassifier(criterion = 'gini' , max_depth = 15 , splitter =_
      ↪'best')
      clf_gini_d18 = DecisionTreeClassifier(criterion = 'gini' , max_depth = 18 , splitter =_
      ↪'best')

      clf_g_d1 = AdaBoostClassifier(base_estimator = clf_gini_d1 ,n_estimators=50,_
      ↪random_state=0)
      clf_g_d3 = AdaBoostClassifier(base_estimator = clf_gini_d3 ,n_estimators=50,_
      ↪random_state=0)
      clf_g_d6 = AdaBoostClassifier(base_estimator = clf_gini_d6 ,n_estimators=50,_
      ↪random_state=0)
      clf_g_d9 = AdaBoostClassifier(base_estimator = clf_gini_d9 ,n_estimators=50,_
      ↪random_state=0)
      clf_g_d12 = AdaBoostClassifier(base_estimator = clf_gini_d12 ,n_estimators=50,_
      ↪random_state=0)
      clf_g_d15 = AdaBoostClassifier(base_estimator = clf_gini_d15 ,n_estimators=50,_
      ↪random_state=0)
      clf_g_d18 = AdaBoostClassifier(base_estimator = clf_gini_d18 ,n_estimators=50,_
      ↪random_state=0)

      clf_ada_g_1 = clf_g_d1.fit(X_train_std,y_train)
```

```

clf_ada_g_3 = clf_g_d3.fit(X_train_std,y_train)
clf_ada_g_6 = clf_g_d6.fit(X_train_std,y_train)
clf_ada_g_9 = clf_g_d9.fit(X_train_std,y_train)
clf_ada_g_12 = clf_g_d12.fit(X_train_std,y_train)
clf_ada_g_15 = clf_g_d15.fit(X_train_std,y_train)
clf_ada_g_18 = clf_g_d18.fit(X_train_std,y_train)

```

```

y_pred_g_1 = clf_ada_g_1.predict(X_test_std)
y_pred_g_3 = clf_ada_g_3.predict(X_test_std)
y_pred_g_6 = clf_ada_g_6.predict(X_test_std)
y_pred_g_9 = clf_ada_g_9.predict(X_test_std)
y_pred_g_12 = clf_ada_g_12.predict(X_test_std)
y_pred_g_15 = clf_ada_g_15.predict(X_test_std)
y_pred_g_18 = clf_ada_g_18.predict(X_test_std)

```

```

[74]: print("Accuracy Gini 1:",metrics.accuracy_score(y_test, y_pred_g_1))
      print("Accuracy Gini 3:",metrics.accuracy_score(y_test, y_pred_g_3))
      print("Accuracy Gini 6:",metrics.accuracy_score(y_test, y_pred_g_6))
      print("Accuracy Gini 9:",metrics.accuracy_score(y_test, y_pred_g_9))
      print("Accuracy Gini 12:",metrics.accuracy_score(y_test, y_pred_g_12))
      print("Accuracy Gini 15:",metrics.accuracy_score(y_test, y_pred_g_15))
      print("Accuracy Gini 18:",metrics.accuracy_score(y_test, y_pred_g_18))

```

```

Accuracy Gini 1: 0.9645080946450809
Accuracy Gini 3: 0.975093399750934
Accuracy Gini 6: 0.9782067247820673
Accuracy Gini 9: 0.9788293897882939
Accuracy Gini 12: 0.9825653798256538
Accuracy Gini 15: 0.9813200498132005
Accuracy Gini 18: 0.9838107098381071

```

## TRAIN MODEL & CREATE PREDICTIONS USING SKLEARN - ENTROPY

```

[76]: clf_entropy_d1 = DecisionTreeClassifier(criterion = 'entropy' , max_depth = 1 ,
      ↳splitter = 'best')
      clf_entropy_d3 = DecisionTreeClassifier(criterion = 'entropy' , max_depth = 3 ,
      ↳splitter = 'best')
      clf_entropy_d6 = DecisionTreeClassifier(criterion = 'entropy' , max_depth = 6 ,
      ↳splitter = 'best')
      clf_entropy_d9 = DecisionTreeClassifier(criterion = 'entropy' , max_depth = 9 ,
      ↳splitter = 'best')
      clf_entropy_d12 = DecisionTreeClassifier(criterion = 'entropy' , max_depth = 12 ,
      ↳splitter = 'best')
      clf_entropy_d15 = DecisionTreeClassifier(criterion = 'entropy' , max_depth = 15 ,
      ↳splitter = 'best')
      clf_entropy_d18 = DecisionTreeClassifier(criterion = 'entropy' , max_depth = 18 ,
      ↳splitter = 'best')

```

```

clf_e_d1 = AdaBoostClassifier(base_estimator = clf_entropy_d1 ,n_estimators=50,
    ↳random_state=0)
clf_e_d3 = AdaBoostClassifier(base_estimator = clf_entropy_d3 ,n_estimators=50,
    ↳random_state=0)
clf_e_d6 = AdaBoostClassifier(base_estimator = clf_entropy_d6 ,n_estimators=50,
    ↳random_state=0)
clf_e_d9 = AdaBoostClassifier(base_estimator = clf_entropy_d9 ,n_estimators=50,
    ↳random_state=0)
clf_e_d12 = AdaBoostClassifier(base_estimator = clf_entropy_d12 ,n_estimators=50,
    ↳random_state=0)
clf_e_d15 = AdaBoostClassifier(base_estimator = clf_entropy_d15 ,n_estimators=50,
    ↳random_state=0)
clf_e_d18 = AdaBoostClassifier(base_estimator = clf_entropy_d18 ,n_estimators=50,
    ↳random_state=0)

clf_ada_e_1 = clf_e_d1.fit(X_train_std,y_train)
clf_ada_e_3 = clf_e_d3.fit(X_train_std,y_train)
clf_ada_e_6 = clf_e_d6.fit(X_train_std,y_train)
clf_ada_e_9 = clf_e_d9.fit(X_train_std,y_train)
clf_ada_e_12 = clf_e_d12.fit(X_train_std,y_train)
clf_ada_e_15 = clf_e_d15.fit(X_train_std,y_train)
clf_ada_e_18 = clf_e_d18.fit(X_train_std,y_train)

y_pred_e_1 = clf_ada_e_1.predict(X_test_std)
y_pred_e_3 = clf_ada_e_3.predict(X_test_std)
y_pred_e_6 = clf_ada_e_6.predict(X_test_std)
y_pred_e_9 = clf_ada_e_9.predict(X_test_std)
y_pred_e_12 = clf_ada_e_12.predict(X_test_std)
y_pred_e_15 = clf_ada_e_15.predict(X_test_std)
y_pred_e_18 = clf_ada_e_18.predict(X_test_std)

```

```

[77]: print("Accuracy Entropy 1:",metrics.accuracy_score(y_test, y_pred_e_1))
print("Accuracy Entropy 3:",metrics.accuracy_score(y_test, y_pred_e_3))
print("Accuracy Entropy 6:",metrics.accuracy_score(y_test, y_pred_e_6))
print("Accuracy Entropy 9:",metrics.accuracy_score(y_test, y_pred_e_9))
print("Accuracy Entropy 12:",metrics.accuracy_score(y_test, y_pred_e_12))
print("Accuracy Entropy 15:",metrics.accuracy_score(y_test, y_pred_e_15))
print("Accuracy Entropy 18:",metrics.accuracy_score(y_test, y_pred_e_18))

```

```

Accuracy Entropy 1: 0.9676214196762142
Accuracy Entropy 3: 0.9800747198007472
Accuracy Entropy 6: 0.9794520547945206
Accuracy Entropy 9: 0.9825653798256538
Accuracy Entropy 12: 0.9819427148194272
Accuracy Entropy 15: 0.9694894146948941
Accuracy Entropy 18: 0.9694894146948941

```

**ANSWER :** The accuracy using Ensemble is much higher at a greater performance as compared

to using Decision Tree independently. The max accuracy reached with Ensemble was around 98% however around 88% with only using decision tree without ensemble.