[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	charcompace
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	
1	0	0	0	
2	0	0	0	
3	0	0	0	
4	0	0	0	
5	0	0	0	
6	0	0	0	
7	0	0	0	
8	0	0	0	
9	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  delimeter_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

   delimeter_path  delimeter_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

   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

[61]:
```
data.isnull().sum()
```

[61]:
```
Querylength                  0
domain_token_count           0
path_token_count             0
avgdomaintokenlen            0
longdomaintokenlen           0
avgpathtokenlen            280
tld                          0
charcompvowels               0
charcompace                  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
delimeter_Domain                       0
delimeter_path                         0
delimeter_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 = data.dropna()          #(subset=['Entropy_DirectoryName'])
```

[5]:
```
data_clean['URL_Type_obf_Type'].unique()
```

[5]:
```
array(['Defacement', 'benign', 'malware', 'phishing', 'spam'],
      dtype=object)
```

[6]:
```
data_clean.dtypes
```

[6]:
```
Querylength                  int64
domain_token_count           int64
path_token_count             int64
avgdomaintokenlen          float64
longdomaintokenlen           int64
avgpathtokenlen            float64
tld                          int64
charcompvowels               int64
charcompace                  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
delimeter_Domain                     int64
delimeter_path                       int64
delimeter_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
```

[62]:
```
# 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

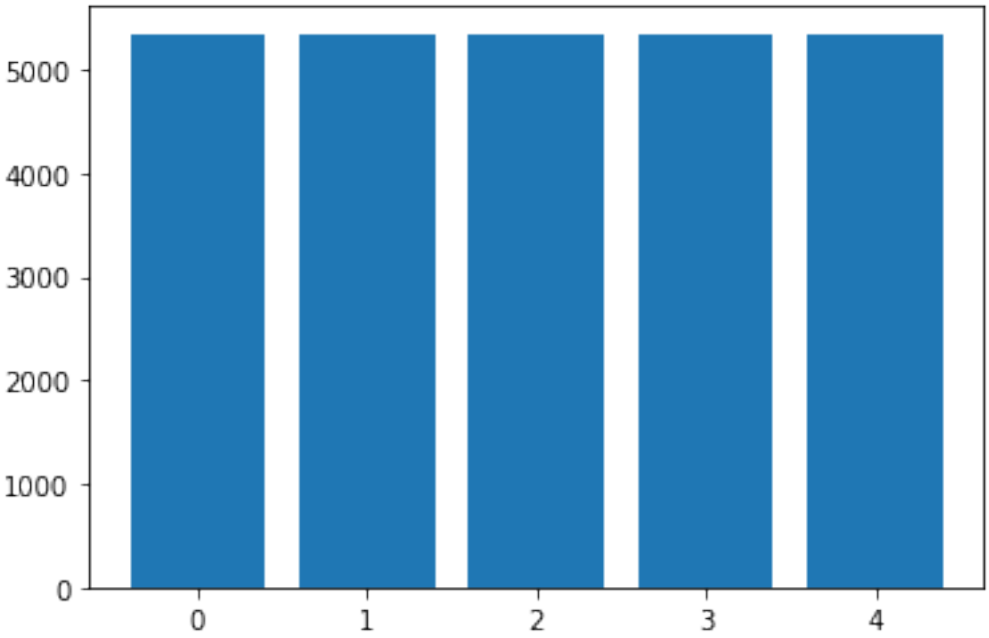
[63]:
```
from sklearn.preprocessing import LabelEncoder
from imblearn.over_sampling import SMOTE
from collections import Counter
```

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

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

[64]:
```
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=5342 (20.000%)
Class=1, n=5342 (20.000%)
Class=2, n=5342 (20.000%)
Class=3, n=5342 (20.000%)
Class=4, n=5342 (20.000%)
```

[64]: <BarContainer object of 5 artists>



Test and Training Split and Scaling the data

[66]:
```
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))
```

[66]:
```
StandardScaler()
```

```
Labels count in y: [5342 5342 5342 5342 5342]
Labels count in y_train: [4273 4274 4274 4273 4274]
Labels count in y_test: [1069 1068 1068 1069 1068]
```

TRAIN MODEL & CREATE PREDICTIONS USING SKLEARN - GINI CRITERION

[49]:
```
from sklearn.tree import DecisionTreeClassifier


clf_gini_d1 = DecisionTreeClassifier(criterion = 'gini' , max_depth = 1 , splitter =␣
↪'best')
clf_gini_d2 = DecisionTreeClassifier(criterion = 'gini' , max_depth = 2 , splitter =␣
↪'best')
clf_gini_d3 = DecisionTreeClassifier(criterion = 'gini' , max_depth = 3 , splitter =␣
↪'best')
clf_gini_d4 = DecisionTreeClassifier(criterion = 'gini' , max_depth = 4 , splitter =␣
↪'best')
clf_gini_d5 = DecisionTreeClassifier(criterion = 'gini' , max_depth = 5 , splitter =␣
↪'best')
clf_gini_d6 = DecisionTreeClassifier(criterion = 'gini' , max_depth = 6 , splitter =␣
↪'best')


#developing a model with gini

clf_gini_d1 = clf_gini_d1.fit(X_train_std,y_train)
clf_gini_d2 = clf_gini_d2.fit(X_train_std,y_train)
clf_gini_d3 = clf_gini_d3.fit(X_train_std,y_train)
clf_gini_d4 = clf_gini_d4.fit(X_train_std,y_train)
```

```
clf_gini_d5 = clf_gini_d5.fit(X_train_std,y_train)
clf_gini_d6 = clf_gini_d6.fit(X_train_std,y_train)
```

[50]:
```
pred_gini_d1 = clf_gini_d1.predict(X_test_std)
pred_gini_d2 = clf_gini_d2.predict(X_test_std)
pred_gini_d3 = clf_gini_d3.predict(X_test_std)
pred_gini_d4 = clf_gini_d4.predict(X_test_std)
pred_gini_d5 = clf_gini_d5.predict(X_test_std)
pred_gini_d6 = clf_gini_d6.predict(X_test_std)
```

TRAIN MODEL & CREATE PREDICTIONS USING SKLEARN - ENTROPY CRITERION

[51]:
```
clf_en_d1 = DecisionTreeClassifier(criterion = 'entropy' , max_depth = 1 , splitter =␣
↪'best')
clf_en_d2 = DecisionTreeClassifier(criterion = 'entropy' , max_depth = 2 , splitter =␣
↪'best')
clf_en_d3 = DecisionTreeClassifier(criterion = 'entropy' , max_depth = 3 , splitter =␣
↪'best')
clf_en_d4 = DecisionTreeClassifier(criterion = 'entropy' , max_depth = 4 , splitter =␣
↪'best')
clf_en_d5 = DecisionTreeClassifier(criterion = 'entropy' , max_depth = 5 , splitter =␣
↪'best')
clf_en_d6 = DecisionTreeClassifier(criterion = 'entropy' , max_depth = 6 , splitter =␣
↪'best')

clf_en_d1 = clf_en_d1.fit(X_train_std,y_train)
clf_en_d2 = clf_en_d2.fit(X_train_std,y_train)
clf_en_d3 = clf_en_d3.fit(X_train_std,y_train)
clf_en_d4 = clf_en_d4.fit(X_train_std,y_train)
clf_en_d5 = clf_en_d5.fit(X_train_std,y_train)
clf_en_d6 = clf_en_d6.fit(X_train_std,y_train)
```

[ ]:
```
pred_en_d1 = clf_en_d1.predict(X_test_std)
pred_en_d2 = clf_en_d2.predict(X_test_std)
pred_en_d3 = clf_en_d3.predict(X_test_std)
pred_en_d4 = clf_en_d4.predict(X_test_std)
pred_en_d5 = clf_en_d5.predict(X_test_std)
pred_en_d6 = clf_en_d6.predict(X_test_std)
```

Create GINICLASSFICATION REPORT USING SKLEARN

[67]:
```
from sklearn.metrics import classification_report

print('-------------------------------------CLASSICATION REPORT [Differnt␣
↪heights]------------------------')
print('')
print('')
print('ml_gini=1',classification_report(y_test,pred_gini_d1,target_names=['Defacement','benign',␣
↪'malware', 'phishing', 'spam']))

print('##########################################################################################
```

```
print('')
print('')
print('ml_gini=2',classification_report(y_test,pred_gini_d2,target_names=['Defacement','benign',␣
 ↪'malware', 'phishing', 'spam']))

print('##########################################################################################')
print('')
print('')
print('ml_gini=3',classification_report(y_test,pred_gini_d3,target_names=['Defacement','benign',␣
 ↪'malware', 'phishing', 'spam']))


print('##########################################################################################')
print('')
print('')
print('ml_gini=4',classification_report(y_test,pred_gini_d4,target_names=['Defacement','benign',␣
 ↪'malware', 'phishing', 'spam']))


print('##########################################################################################')
print('')
print('')
print('ml_gini=5',classification_report(y_test,pred_gini_d5,target_names=['Defacement','benign',␣
 ↪'malware', 'phishing', 'spam']))


print('##########################################################################################')
print('')
print('')
print('ml_gini=6',classification_report(y_test,pred_gini_d6,target_names=['Defacement','benign',␣
 ↪'malware', 'phishing', 'spam']))
```

```
-------------------------------------CLASSICATION REPORT [Differnt
heights]------------------------


ml_gini=1               precision    recall  f1-score   support

  Defacement       0.00      0.00      0.00      1069
      benign       0.46      0.93      0.61      1068
     malware       0.00      0.00      0.00      1068
    phishing       0.00      0.00      0.00      1069
        spam       0.34      1.00      0.50      1068

    accuracy                           0.39      5342
   macro avg       0.16      0.39      0.22      5342
weighted avg       0.16      0.39      0.22      5342

##########################################################################################
##########
```

```
ml_gini=2              precision    recall  f1-score   support

  Defacement       0.34      0.57      0.43      1069
      benign       0.62      0.93      0.75      1068
     malware       0.00      0.00      0.00      1068
    phishing       0.56      0.30      0.39      1069
        spam       0.65      0.83      0.73      1068

    accuracy                           0.53      5342
   macro avg       0.43      0.53      0.46      5342
weighted avg       0.43      0.53      0.46      5342

##########################################################################################
##########


ml_gini=3              precision    recall  f1-score   support

  Defacement       0.68      0.69      0.68      1069
      benign       0.77      0.84      0.80      1068
     malware       0.83      0.21      0.33      1068
    phishing       0.47      0.75      0.58      1069
        spam       0.80      0.82      0.81      1068

    accuracy                           0.66      5342
   macro avg       0.71      0.66      0.64      5342
weighted avg       0.71      0.66      0.64      5342

##########################################################################################
##########


ml_gini=4              precision    recall  f1-score   support

  Defacement       0.84      0.67      0.74      1069
      benign       0.84      0.77      0.81      1068
     malware       0.67      0.62      0.64      1068
    phishing       0.53      0.75      0.62      1069
        spam       0.86      0.82      0.84      1068

    accuracy                           0.73      5342
   macro avg       0.75      0.73      0.73      5342
weighted avg       0.75      0.73      0.73      5342

##########################################################################################
##########


ml_gini=5              precision    recall  f1-score   support

  Defacement       0.69      0.88      0.77      1069
      benign       0.85      0.86      0.85      1068
```

```
     malware       0.75      0.65      0.70      1068
    phishing       0.70      0.70      0.70      1069
        spam       0.91      0.76      0.83      1068

    accuracy                           0.77      5342
   macro avg       0.78      0.77      0.77      5342
weighted avg       0.78      0.77      0.77      5342

##########################################################################################
##########


ml_gini=6              precision    recall  f1-score   support

  Defacement       0.76      0.88      0.82      1069
      benign       0.86      0.90      0.88      1068
     malware       0.89      0.75      0.82      1068
    phishing       0.76      0.71      0.73      1069
        spam       0.91      0.92      0.91      1068

    accuracy                           0.83      5342
   macro avg       0.84      0.83      0.83      5342
weighted avg       0.84      0.83      0.83      5342

/Users/samipsinghal/opt/anaconda3/lib/python3.7/site-
packages/sklearn/metrics/_classification.py:1308: UndefinedMetricWarning: Precision and
F-score are ill-defined and being set to 0.0 in labels with no predicted samples. Use
`zero_division` parameter to control this behavior.
  _warn_prf(average, modifier, msg_start, len(result))
/Users/samipsinghal/opt/anaconda3/lib/python3.7/site-
packages/sklearn/metrics/_classification.py:1308: UndefinedMetricWarning: Precision and
F-score are ill-defined and being set to 0.0 in labels with no predicted samples. Use
`zero_division` parameter to control this behavior.
  _warn_prf(average, modifier, msg_start, len(result))
/Users/samipsinghal/opt/anaconda3/lib/python3.7/site-
packages/sklearn/metrics/_classification.py:1308: UndefinedMetricWarning: Precision and
F-score are ill-defined and being set to 0.0 in labels with no predicted samples. Use
`zero_division` parameter to control this behavior.
  _warn_prf(average, modifier, msg_start, len(result))
/Users/samipsinghal/opt/anaconda3/lib/python3.7/site-
packages/sklearn/metrics/_classification.py:1308: UndefinedMetricWarning: Precision and
F-score are ill-defined and being set to 0.0 in labels with no predicted samples. Use
`zero_division` parameter to control this behavior.
  _warn_prf(average, modifier, msg_start, len(result))
/Users/samipsinghal/opt/anaconda3/lib/python3.7/site-
packages/sklearn/metrics/_classification.py:1308: UndefinedMetricWarning: Precision and
F-score are ill-defined and being set to 0.0 in labels with no predicted samples. Use
`zero_division` parameter to control this behavior.
  _warn_prf(average, modifier, msg_start, len(result))
/Users/samipsinghal/opt/anaconda3/lib/python3.7/site-
packages/sklearn/metrics/_classification.py:1308: UndefinedMetricWarning: Precision and
```

```
F-score are ill-defined and being set to 0.0 in labels with no predicted samples. Use
`zero_division` parameter to control this behavior.
  _warn_prf(average, modifier, msg_start, len(result))
```

Create ENTROPY CLASSFICATION REPORT USING SKLEARN

[68]:
```
from sklearn.metrics import classification_report

print('-------------------------------------ENTROPY CLASSICATION REPORT [Differnt␣
 ↪heights]------------------------')
print('')
print('')
print('ml_en=1',classification_report(y_test,pred_en_d1,target_names=['Defacement','benign',␣
 ↪'malware', 'phishing', 'spam']))

print('#############################################################################################
print('')
print('')
print('ml_en=2',classification_report(y_test,pred_gini_d2,target_names=['Defacement','benign',␣
 ↪'malware', 'phishing', 'spam']))

print('#############################################################################################
print('')
print('')
print('ml_en=3',classification_report(y_test,pred_gini_d3,target_names=['Defacement','benign',␣
 ↪'malware', 'phishing', 'spam']))


print('#############################################################################################
print('')
print('')
print('ml_en=4',classification_report(y_test,pred_gini_d4,target_names=['Defacement','benign',␣
 ↪'malware', 'phishing', 'spam']))


print('#############################################################################################
print('')
print('')
print('ml_en=5',classification_report(y_test,pred_gini_d5,target_names=['Defacement','benign',␣
 ↪'malware', 'phishing', 'spam']))


print('#############################################################################################
print('')
print('')
print('ml_en=6',classification_report(y_test,pred_gini_d6,target_names=['Defacement','benign',␣
 ↪'malware', 'phishing', 'spam']))
```

```
-------------------------------------ENTROPY CLASSICATION REPORT [Differnt
heights]------------------------
```

```
ml_en=1              precision    recall  f1-score   support

  Defacement       0.00      0.00      0.00      1069
      benign       0.46      0.93      0.61      1068
     malware       0.00      0.00      0.00      1068
    phishing       0.00      0.00      0.00      1069
        spam       0.34      1.00      0.50      1068

    accuracy                           0.39      5342
   macro avg       0.16      0.39      0.22      5342
weighted avg       0.16      0.39      0.22      5342

##########################################################################################
##########


ml_en=2              precision    recall  f1-score   support

  Defacement       0.34      0.57      0.43      1069
      benign       0.62      0.93      0.75      1068
     malware       0.00      0.00      0.00      1068
    phishing       0.56      0.30      0.39      1069
        spam       0.65      0.83      0.73      1068

    accuracy                           0.53      5342
   macro avg       0.43      0.53      0.46      5342
weighted avg       0.43      0.53      0.46      5342

##########################################################################################
##########


ml_en=3              precision    recall  f1-score   support

  Defacement       0.68      0.69      0.68      1069
      benign       0.77      0.84      0.80      1068
     malware       0.83      0.21      0.33      1068
    phishing       0.47      0.75      0.58      1069
        spam       0.80      0.82      0.81      1068

    accuracy                           0.66      5342
   macro avg       0.71      0.66      0.64      5342
weighted avg       0.71      0.66      0.64      5342

##########################################################################################
##########


ml_en=4              precision    recall  f1-score   support

  Defacement       0.84      0.67      0.74      1069
      benign       0.84      0.77      0.81      1068
```

```
     malware       0.67      0.62      0.64      1068
    phishing       0.53      0.75      0.62      1069
        spam       0.86      0.82      0.84      1068

    accuracy                           0.73      5342
   macro avg       0.75      0.73      0.73      5342
weighted avg       0.75      0.73      0.73      5342

##########################################################################################
##########


ml_en=5              precision    recall  f1-score   support

  Defacement       0.69      0.88      0.77      1069
      benign       0.85      0.86      0.85      1068
     malware       0.75      0.65      0.70      1068
    phishing       0.70      0.70      0.70      1069
        spam       0.91      0.76      0.83      1068

    accuracy                           0.77      5342
   macro avg       0.78      0.77      0.77      5342
weighted avg       0.78      0.77      0.77      5342

##########################################################################################
##########


ml_en=6              precision    recall  f1-score   support

  Defacement       0.76      0.88      0.82      1069
      benign       0.86      0.90      0.88      1068
     malware       0.89      0.75      0.82      1068
    phishing       0.76      0.71      0.73      1069
        spam       0.91      0.92      0.91      1068

    accuracy                           0.83      5342
   macro avg       0.84      0.83      0.83      5342
weighted avg       0.84      0.83      0.83      5342


/Users/samipsinghal/opt/anaconda3/lib/python3.7/site-
packages/sklearn/metrics/_classification.py:1308: UndefinedMetricWarning: Precision and
F-score are ill-defined and being set to 0.0 in labels with no predicted samples. Use
`zero_division` parameter to control this behavior.
  _warn_prf(average, modifier, msg_start, len(result))
/Users/samipsinghal/opt/anaconda3/lib/python3.7/site-
packages/sklearn/metrics/_classification.py:1308: UndefinedMetricWarning: Precision and
F-score are ill-defined and being set to 0.0 in labels with no predicted samples. Use
`zero_division` parameter to control this behavior.
  _warn_prf(average, modifier, msg_start, len(result))
/Users/samipsinghal/opt/anaconda3/lib/python3.7/site-
```

```
packages/sklearn/metrics/_classification.py:1308: UndefinedMetricWarning: Precision and
F-score are ill-defined and being set to 0.0 in labels with no predicted samples. Use
`zero_division` parameter to control this behavior.
  _warn_prf(average, modifier, msg_start, len(result))
/Users/samipsinghal/opt/anaconda3/lib/python3.7/site-
packages/sklearn/metrics/_classification.py:1308: UndefinedMetricWarning: Precision and
F-score are ill-defined and being set to 0.0 in labels with no predicted samples. Use
`zero_division` parameter to control this behavior.
  _warn_prf(average, modifier, msg_start, len(result))
/Users/samipsinghal/opt/anaconda3/lib/python3.7/site-
packages/sklearn/metrics/_classification.py:1308: UndefinedMetricWarning: Precision and
F-score are ill-defined and being set to 0.0 in labels with no predicted samples. Use
`zero_division` parameter to control this behavior.
  _warn_prf(average, modifier, msg_start, len(result))
/Users/samipsinghal/opt/anaconda3/lib/python3.7/site-
packages/sklearn/metrics/_classification.py:1308: UndefinedMetricWarning: Precision and
F-score are ill-defined and being set to 0.0 in labels with no predicted samples. Use
`zero_division` parameter to control this behavior.
  _warn_prf(average, modifier, msg_start, len(result))
```

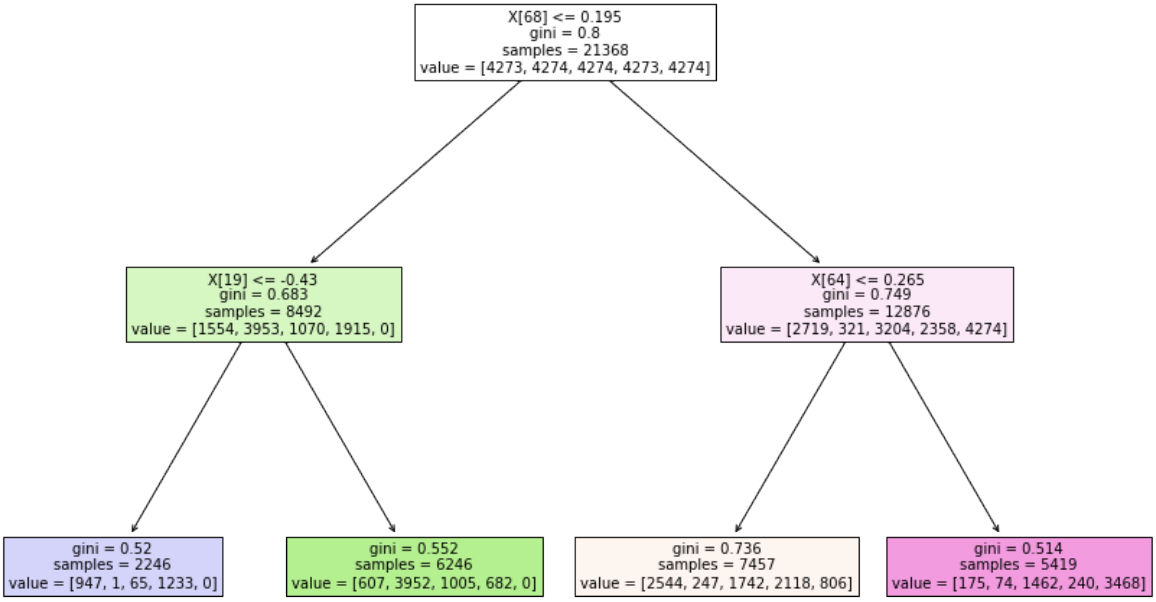
ANSWER : As the depth of the tree increases , the accuracy of prediction increases

VISUALIZING A TREE [for numpy , depth =2)

[69]:
```
import matplotlib.pyplot as plt
from sklearn import tree
%matplotlib inline

plt.figure(figsize=(15,10))
tree.plot_tree(clf_gini_d2,filled='true')
```

[69]:
```
<Figure size 1080x720 with 0 Axes>
```

[69]:
```
[Text(418.5, 453.0, 'X[68] <= 0.195\ngini = 0.8\nsamples = 21368\nvalue = [4273, 4274,
4274, 4273, 4274]'),
 Text(209.25, 271.8, 'X[19] <= -0.43\ngini = 0.683\nsamples = 8492\nvalue = [1554, 3953,
1070, 1915, 0]'),
 Text(104.625, 90.59999999999997, 'gini = 0.52\nsamples = 2246\nvalue = [947, 1, 65,
1233, 0]'),
 Text(313.875, 90.59999999999997, 'gini = 0.552\nsamples = 6246\nvalue = [607, 3952,
1005, 682, 0]'),
 Text(627.75, 271.8, 'X[64] <= 0.265\ngini = 0.749\nsamples = 12876\nvalue = [2719, 321,
3204, 2358, 4274]'),
 Text(523.125, 90.59999999999997, 'gini = 0.736\nsamples = 7457\nvalue = [2544, 247,
1742, 2118, 806]'),
 Text(732.375, 90.59999999999997, 'gini = 0.514\nsamples = 5419\nvalue = [175, 74, 1462,
240, 3468]')]
```


X[68] <= 0.195
gini = 0.8
samples = 21368
value = [4273, 4274, 4274, 4273, 4274]
X[19] <= -0.43
gini = 0.683
samples = 8492
value = [1554, 3953, 1070, 1915, 0]
X[64] <= 0.265
gini = 0.749
samples = 12876
value = [2719, 321, 3204, 2358, 4274]
gini = 0.52
samples = 2246
value = [947, 1, 65, 1233, 0]
gini = 0.552
samples = 6246
value = [607, 3952, 1005, 682, 0]
gini = 0.736
samples = 7457
value = [2544, 247, 1742, 2118, 806]
gini = 0.514
samples = 5419
value = [175, 74, 1462, 240, 3468]

[ ]:

[ ]: