CSE445 Machine Learning

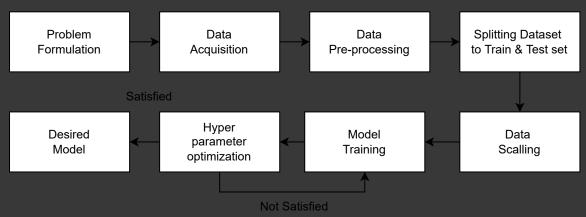
Online Class Preference Prediction Using Machine Learning Approach

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We are proposing a machine-learning model to predict preference of online class among Bangladeshi students. Our goal is to create an efficient machine-learning model to predict if a student prefers online class or not by using some common available features such as age, gender, level of study, preferred device, results, knowledge and class performance development during online class, internet availability, location of joining, difficulties faced, etc.

Methodology

The major objective of this work is to develop a machine-learning model that will aid to predict if a student likes online classes or not. The approach adopted in this work is outlined in Fig. 1



The model to be developed to predict the response for the training data will be developed using the decision tree technique. It is one of the most popular and straightforward machine learning algorithms for categorization problems. Since supervised learning approach is to be used in this work and the model has to predict a target class that is categorized into "Yes" and "No", the decision tree algorithm will be useful to create a training model that can predict the target class by learning some decision rules inferred from training data.

Importing Libraries

```
# importing libraries
import pandas as pd # data processing
import numpy as np # linear algebra
import matplotlib.pyplot as plt # visualization
%matplotlib inline
import seaborn as sns
# increases the size of sns plots
sns.set(rc={'figure.figsize':(8,6)})

from sklearn.model_selection import train_test_split, KFold, cross_val_score
from sklearn import tree
from sklearn.tree import DecisionTreeClassifier, export_graphviz
from sklearn.metrics import accuracy_score, confusion_matrix, r2_score, roc_curve, auc, classification_report
```

```
import warnings
warnings.filterwarnings('ignore')
```

→ Data Acquisition

mount google drive

from google.colab import drive
drive.mount('/content/drive')

Dataset is collected from Kaggle. The dataset is created based on an online survey on Bangladeshi students and it contains 17 features such as age, level of study, devices used, result, knowledge and class performance in online class, have interest, internet availability, institute type, happy with online class etc.

```
# raw data in panda dataframe
df = pd.read_csv('/content/drive/MyDrive/CSE 445 Project/Online Survey Data on Education Bd.csv')
print('Data Frame Shape: \n{}'.format(df.shape))
df.columns = df.columns.str.replace('Used smartphone/computer/laptop previously before online class?',
'Used Electronic Devices?')
# shows five instances of the dataframe
print('First few instances of the dataset: ')
df.head()
```

Data Frame Shape: (8783, 17) First few instances of the dataset:

	Level of study?	Age?	Used Electronic Devices??	Result increased after online education (comparatively)?	Knowledge increased after online education (comparatively)?	Happy with online education?	Education Institute Area?	Have Internet availability?	Broadband / Mobile Internet?	online	hours of study after online education?	Class performance increased in online education?	Institute Type		Gender	any issue with online class?	Preferred device for an online course
0	Upto HSC	20.0	Yes	No	Yes	No	Urban	No	Broadband	4	3	No	Public	Rural	Male	Yes	Mobile
1	Hons or Grater	25.0	No	No	No	No	Urban	No	Mobile Internet	4	4	No	Public	Rural	Male	Yes	Mobile
2	Hons or Grater	25.0	Yes	Yes	Yes	Yes	Rural	No	Mobile Internet	5	2	Yes	Public	Rural	Female	Yes	Computer
3	Upto HSC	21.0	Yes	Yes	No	Yes	Urban	Yes	Mobile Internet	5	3	No	Private	Urban	Male	Yes	Mobile
4	Hons or	22.0	Yes	No	No	No	Rural	No	Mobile Internet	4	2	Yes	Public	Urban	Male	No	Mobile

Total

Faced

columns of the dataset
df.columns

Grater

Index(['Level of study?', 'Age?', 'Used Electronic Devices??',

```
'Result increased after online education (comparatively)?',
            'Knowledge increased after online education (comparatively)?',
            'Happy with online education?', 'Education Institute Area?', 'Have Internet availability?', 'Broadband / Mobile Internet?',
            'Total hours of study before online education?',
            'Total hours of study after online education?',
            'Class performance increased in online education?', 'Institute Type',
            'Current location (During Study) ?', 'Gender',
            'Faced any issue with online class?',
            'Preferred device for an online course'],
           dtype='object')
# investigating all the elements whithin each Feature
for column in df:
  unique_vals = df[column].unique()
  nr_values = len(unique_vals)
  if nr_values < 10:</pre>
    print('The number of values for feature {} :{} -- {}'.format(column, nr_values,unique_vals))
  else:
    print('The number of values for feature {} :{}'.format(column, nr_values))
    The number of values for feature Level of study? :2 -- ['Upto HSC' 'Hons or Grater']
     The number of values for feature Age? :12
    The number of values for feature Used Electronic Devices?? :3 -- ['Yes' 'No' nan]
     The number of values for feature Result increased after online education (comparatively)? :3 -- ['No' 'Yes' nan]
     The number of values for feature Knowledge increased after online education (comparatively)? :2 -- ['Yes' 'No']
    The number of values for feature Happy with online education? :2 -- ['No' 'Yes']
     The number of values for feature Education Institute Area? :3 -- ['Urban' 'Rural' nan]
    The number of values for feature Have Internet availability? :2 -- ['No' 'Yes']
     The number of values for feature Broadband / Mobile Internet? :2 -- ['Broadband' 'Mobile Internet']
     The number of values for feature Total hours of study before online education? :4 -- [4 5 3 6]
    The number of values for feature Total hours of study after online education? :3 -- [3 4 2]
     The number of values for feature Class performance increased in online education? :2 -- ['No' 'Yes']
     The number of values for feature Institute Type :3 -- ['Public' 'Private' nan]
    The number of values for feature Current location (During Study) ? :3 -- ['Rural' 'Urban' nan]
     The number of values for feature Gender :3 -- ['Male' 'Female' nan]
    The number of values for feature Faced any issue with online class? :3 -- ['Yes' 'No' nan]
    The number of values for feature Preferred device for an online course :2 -- ['Mobile' 'Computer']
# checking for the null values
```

checking for the null values df.isnull().sum()

```
Level of study?
                                                               445
Age?
Used Electronic Devices??
                                                               188
Result increased after online education (comparatively)?
                                                               323
Knowledge increased after online education (comparatively)?
                                                                 0
Happy with online education?
                                                                 0
Education Institute Area?
                                                               529
Have Internet availability?
                                                                 0
Broadband / Mobile Internet?
                                                                 0
Total hours of study before online education?
                                                                 0
Total hours of study after online education?
                                                                 0
Class performance increased in online education?
                                                                 0
Institute Type
                                                               726
Current location (During Study) ?
                                                               726
Gender
                                                               676
                                                               701
Faced any issue with online class?
```

Preferred device for an online course dtype: int64

'Faced any issue with online class?',

Data Preprocessing

For some entries in the collection, multiple columns have null values. The null values are removed. Correlation Matrix is also plotted to see the relationship among attributes.

Removing Null Values

```
Removing null values to make a clean dataset
```

```
# removing rows containing null values and creating a demo dataset
new_df = df.dropna()
print('New Data Frame Shape: ', new_df.shape)
    New Data Frame Shape: (5715, 17)
# checking null values in new data frame
new_df.isnull().sum()
    Level of study?
                                                                      0
     Age?
     Used Electronic Devices??
                                                                      0
     Result increased after online education (comparatively)?
     Knowledge increased after online education (comparatively)?
     Happy with online education?
     Education Institute Area?
     Have Internet availability?
    Broadband / Mobile Internet?
     Total hours of study before online education?
     Total hours of study after online education?
     Class performance increased in online education?
     Institute Type
     Current location (During Study) ?
     Gender
     Faced any issue with online class?
                                                                      0
     Preferred device for an online course
                                                                      0
     dtype: int64
# exporting new dataframe as csv
new_df.to_csv('/content/drive/MyDrive/CSE 445 Project/Online Education Filtered.csv')
# attributes of new dataframe
new_df.columns
     Index(['Level of study?', 'Age?', 'Used Electronic Devices??',
            'Result increased after online education (comparatively)?',
            'Knowledge increased after online education (comparatively)?',
            'Happy with online education?', 'Education Institute Area?', 'Have Internet availability?', 'Broadband / Mobile Internet?',
            'Total hours of study before online education?',
            'Total hours of study after online education?',
            'Class performance increased in online education?', 'Institute Type',
            'Current location (During Study) ?', 'Gender',
```

```
'Preferred device for an online course'], dtype='object')
```

▼ Dataset Encoding

Encoding the dataset to make it suitable for machine learning algorithms

Data columns (total 17 columns):

Column

```
# data types
new_df.dtypes
    Level of study?
                                                                      object
     Age?
                                                                      float64
     Used Electronic Devices??
                                                                      object
     Result increased after online education (comparatively)?
                                                                      object
     Knowledge increased after online education (comparatively)?
                                                                      object
     Happy with online education?
                                                                      object
     Education Institute Area?
                                                                      object
     Have Internet availability?
                                                                      object
     Broadband / Mobile Internet?
                                                                       object
     Total hours of study before online education?
                                                                       int64
     Total hours of study after online education?
                                                                       int64
     Class performance increased in online education?
                                                                      object
     Institute Type
                                                                      object
     Current location (During Study) ?
                                                                      object
     Gender
                                                                      object
    Faced any issue with online class?
                                                                      object
    Preferred device for an online course
                                                                      object
     dtype: object
# Find out all the features with type object
objectList = new_df.select_dtypes(include = "object").columns
print (objectList)
     Index(['Level of study?', 'Used Electronic Devices??',
            'Result increased after online education (comparatively)?',
            'Knowledge increased after online education (comparatively)?',
            'Happy with online education?', 'Education Institute Area?', 'Have Internet availability?', 'Broadband / Mobile Internet?',
            'Class performance increased in online education?', 'Institute Type',
            'Current location (During Study) ?', 'Gender',
            'Faced any issue with online class?',
            'Preferred device for an online course'],
           dtype='object')
#Label Encoding for object to numeric conversion
from sklearn.preprocessing import LabelEncoder
encoder = LabelEncoder()
for obj in objectList:
    new_df[obj] = encoder.fit_transform(new_df[obj].astype(str))
print (new_df.info())
     <class 'pandas.core.frame.DataFrame'>
     Int64Index: 5715 entries, 0 to 8781
```

Non-Null Count Dtype

```
0 Level of study?
                                                                     5715 non-null int64
                                                                     5715 non-null float64
       1 Age?
          Used Electronic Devices??
                                                                     5715 non-null int64
       3 Result increased after online education (comparatively)?
                                                                     5715 non-null int64
           Knowledge increased after online education (comparatively)? 5715 non-null int64
                                                                     5715 non-null int64
          Happy with online education?
                                                                     5715 non-null int64
       6 Education Institute Area?
          Have Internet availability?
                                                                     5715 non-null int64
       8 Broadband / Mobile Internet?
                                                                     5715 non-null int64
       9 Total hours of study before online education?
                                                                     5715 non-null int64
       10 Total hours of study after online education?
                                                                     5715 non-null int64
       11 Class performance increased in online education?
                                                                     5715 non-null int64
       12 Institute Type
                                                                     5715 non-null int64
       13 Current location (During Study) ?
                                                                     5715 non-null int64
                                                                     5715 non-null int64
       14 Gender
       15 Faced any issue with online class?
                                                                     5715 non-null int64
       16 Preferred device for an online course
                                                                     5715 non-null int64
      dtypes: float64(1), int64(16)
      memory usage: 803.7 KB
      None
 # exporting new dataframe as csv
 new_df.to_csv('/content/drive/MyDrive/CSE 445 Project/Online Education Cleanded Dataset.csv')

→ Splitting Dataset
  Splitting the dataset in a 70:30 ratio. 70% for training & 30% for testing
 # separating attributes and target
 attribute = new_df.drop(columns = ['Happy with online education?'])
 target = new_df['Happy with online education?']
 print('Attribute Shape: ', attribute.shape)
 print('Target Shape: ', target.shape)
      Attribute Shape: (5715, 16)
      Target Shape: (5715,)
  target.value_counts()
      0 3677
      1
          2038
      Name: Happy with online education?, dtype: int64
 # first few instances of attribute
 attribute.columns = attribute.columns.str.replace('Used smartphone/computer/laptop previously before online class?',
                                                       'Used Electronic Devices?')
 attribute.head()
```

```
Faced
                                                                                                           Total hours
                                                                                                                                               Class
                                                                                                                                                                                           Preferred
                              Result increased
                                                          Knowledge
                                                                                                                          Total hours
                                                                                                                                                                  Current
                                                                                                                                                                                      any
                                                                                                                                                                                              device
                                                                                                              of study
    Level
                                                                     Education
                                                                                               Broadband
                                                                                                                                         performance
                                  after online
                                                    increased after
                                                                               Have Internet
                                                                                                                            of study
                                                                                                                                                      Institute location
                                                                                                                                                                                    issue
       of Age? Electronic
                                                                     Institute
                                                                                                / Mobile
                                                                                                                before
                                                                                                                                        increased in
                                                                                                                                                                                              for an
                                                                               availability?
                                     education
                                                   online education
                                                                                                                        after online
                                                                                                                                                           Type
                                                                                                                                                                  (During
                                                                                                                                                                                     with
                                                                        Area?
                                                                                              Internet?
                                                                                                                online
                                                                                                                                              online
                                                                                                                                                                                              online
   study?
                 Devices??
                              (comparatively)?
                                                   (comparatively)?
                                                                                                                          education?
                                                                                                                                                                 Study) ?
                                                                                                                                                                                   online
                                                                                                            education?
                                                                                                                                          education?
                                                                                                                                                                                              course
                                                                                                                                                                                   class?
0
       1 20.0
                         1
                                             0
                                                                 1
                                                                                           0
                                                                                                      0
                                                                                                                     4
                                                                                                                                   3
                                                                                                                                                   0
                                                                                                                                                                        0
                                                                                                                                                                                       1
                                                                                                                                                                                                   1
                         0
                                             0
                                                                 0
                                                                                           0
       0 25.0
                                                                                                                                   4
                                                                                                                                                   0
                                                                                                                                                                        0
                                                                                                                                                                                                   0
       0 25.0
                                                                            0
                                                                                            0
                                                                                                                                                                        0
                                                                                                                                                                                0
```

```
# first few instances of target
target.head()
```

```
0 0
1 0
2 1
3 1
```

Name: Happy with online education?, dtype: int64

```
# train test splitting
X_train, X_test, y_train, y_test = train_test_split(attribute, target, train_size = 0.7, test_size = 0.3, random_state = 0)
```

```
print('For training: ')
print('Attribute Shape: ', X_train.shape)
print('Target Shape: ', y_train.shape)

print('\nFor testing: ')
print('Attribute Shape: ', X_test.shape)
print('Target Shape: ', y_test.shape)

For training:
   Attribute Shape: (4000, 16)
```

For testing:

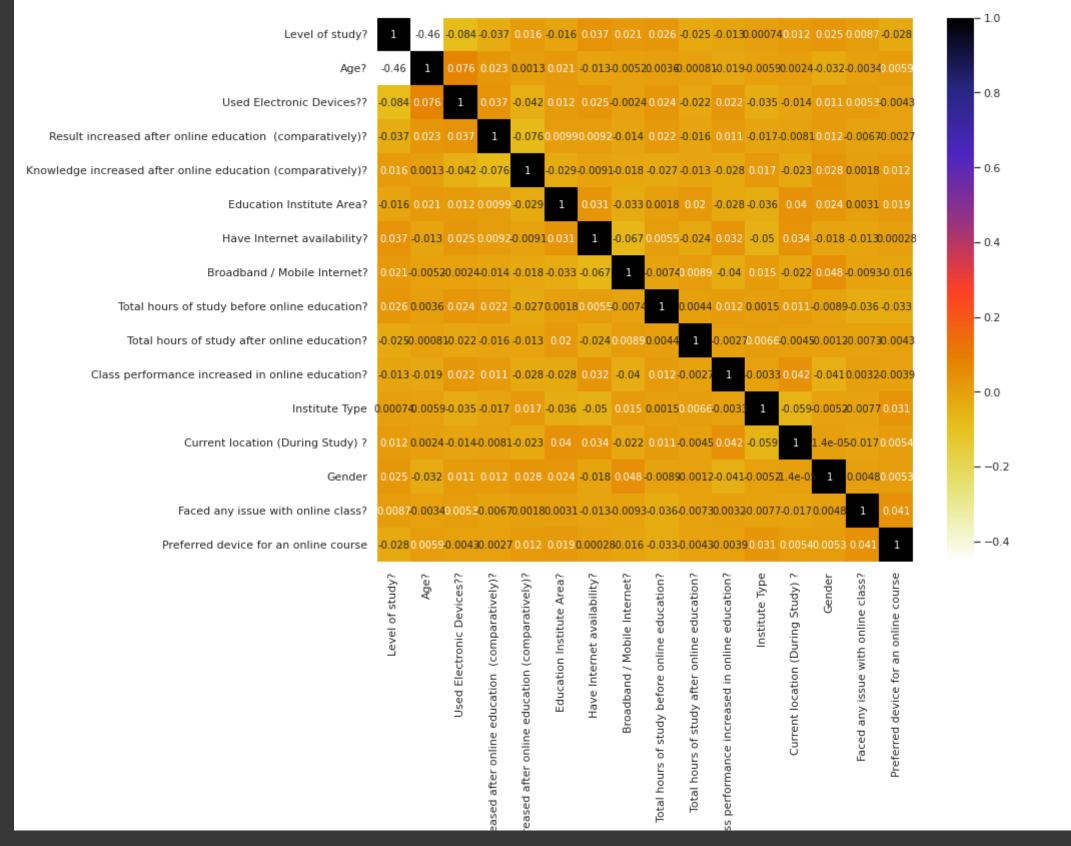
Attribute Shape: (1715, 16)
Target Shape: (1715,)

Target Shape: (4000,)

→ Correlation of Features

Finding the correlation among the features to see how they are connected. Main purpose is to find duplicate features

```
# using pearson correlation
plt.figure(figsize=(12, 10))
correlation = X_train.corr()
sns.heatmap(correlation, annot=True, cmap=plt.cm.CMRmap_r)
plt.show()
```



- Decision Tree

Initially building a decision tree model with a max depth 5, later we will build a random forest classification model with hyper parameter tuing

```
# Decision Tree Model
dtree = DecisionTreeClassifier(max_depth = 5, random_state = 1)
dtree.fit(X_train, y_train)

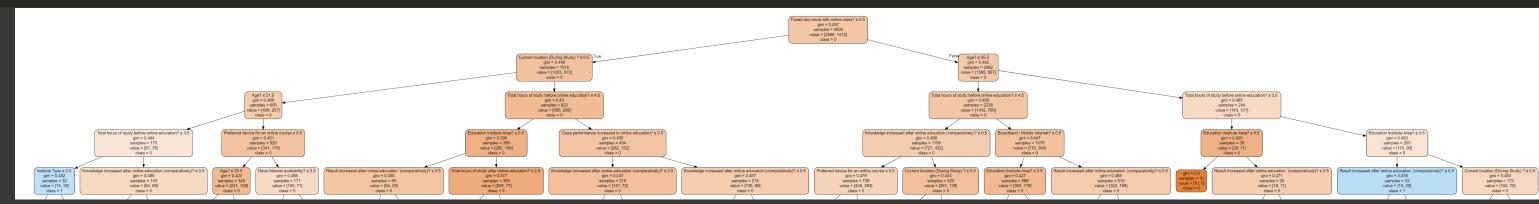
DecisionTreeClassifier(max_depth=5, random_state=1)
```

Graph available in: https://dreampuf.github.io/GraphvizOnline
import graphviz

```
dot_data = tree.export_graphviz(dtree, out_file='/content/drive/MyDrive/CSE 445 Project/Decision Tree.dot',
    feature_names = new_df.drop('Happy with online education?', axis=1).columns,
    class_names = new_df['Happy with online education?'].unique().astype(str),
    filled=True, rounded=True,
    special_characters=True)

graph = graphviz.Source(dot_data)
# Decision Tree garageted from Crophviz
```

```
# Decision Tree generated from Graphviz
from IPython.display import Image
Image(filename='/content/drive/MyDrive/CSE 445 Project/Decision Tree.png')
```



▼ Feature Importance

```
# Finding importance of each feature

for i, column in enumerate(new_df.drop('Happy with online education?', axis=1)):
    print('Importance of feature {}:. {:.3f}'.format(column, dtree.feature_importances_[i]))
    feature_imp = pd.DataFrame({'Variable': [column], 'Feature Importance Score': [dtree.feature_importances_[i]]})

try:
    final_feature_imp = pd.concat([final_feature_imp, feature_imp], ignore_index = True)
    except:
        final_feature_imp = feature_imp

# Ordering the data
final_feature_imp = final_feature_imp.sort_values('Feature Importance Score', ascending = False).reset_index()
final_feature_imp
```

```
Importance of feature Level of study?:, 0.000
Importance of feature Age?:, 0.108
Importance of feature Used Electronic Devices??:, 0.000
Importance of feature Result increased after online education (comparatively)?:, 0.142
Importance of feature Knowledge increased after online education (comparatively)?:, 0.104
Importance of feature Education Institute Area?:, 0.151
Importance of feature Have Internet availability?:, 0.043
Importance of feature Broadband / Mobile Internet?:, 0.037
Importance of feature Total hours of study before online education?:, 0.142
Importance of feature Total hours of study after online education?:, 0.013
Importance of feature Class performance increased in online education?:, 0.008
Importance of feature Institute Type:, 0.015
Importance of feature Current location (During Study) ?:, 0.114
Importance of feature Gender:, 0.000
Importance of feature Faced any issue with online class?:, 0.018
Importance of feature Preferred device for an online course:, 0.103
     index
                                             Variable Feature Importance Score
0
        5
                                Education Institute Area?
                                                                         0.151486
                Total hours of study before online education?
                                                                         0.142285
 2
             Result increased after online education (comp...
                                                                         0.142143
 3
       12
                          Current location (During Study)?
                                                                         0.114371
                                                                         0.107682
 5
           Knowledge increased after online education (co...
                                                                         0.104196
       15
                      Preferred device for an online course
                                                                         0.102895
 6
```

Have Internet availability?

print("Training Accuracy is: ", dtree.score(X_train, y_train))

def plot_confusion_matrix(cm, classes=None, title='Confusion matrix'):

→ Result From Decision Tree

Confusion Matrix

Confusion Matrix function

if classes is not None:

Training Accuracy Of Decision Tree

```
# Test Accuracy Of Decision Tree
print("Testing Accuracy is: ", dtree.score(X_test, y_test))

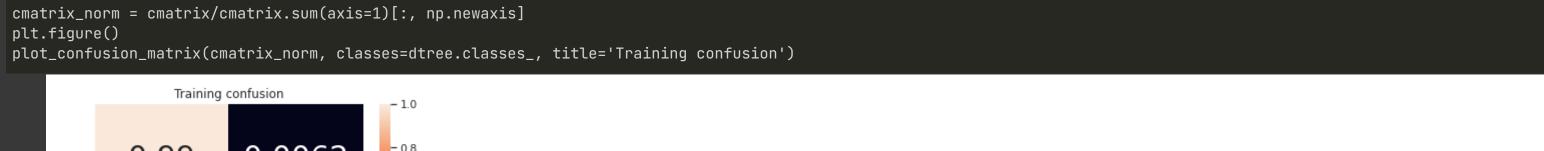
Training Accuracy is: 0.65125
Testing Accuracy is: 0.6297376093294461

# after applying k fold cross validation
kfold_validation = KFold(n_splits = 10)
results = cross_val_score(dtree, attribute, target, cv = kfold_validation)
print(results)
print ('\nResults = ', np.mean(results), '+/-', np.std(results))

[0.6520979  0.62587413  0.63286713  0.63461538  0.64160839  0.6357268  0.62521891  0.63397548  0.67250438  0.6234676 ]

Results = 0.6377956106940345 +/- 0.014080679908422033
```

0.043115





```
# Calculating False Positives (FP), False Negatives (FN), True Positives(TP), True Negatices (TN)
FP = cmatrix.sum(axis=0) - np.diag(cmatrix)
FN = cmatrix.sum(axis=1) - np.diag(cmatrix)
TP = np.diag(cmatrix)
TN = cmatrix.sum() - (FP + FN + TP)

# precision or positive predictive value precision = TP / (TP + FP)
print('Precision per class: ', precision)

# sensitivity, recall or true predictive rate recall = TP / (TP + FN)
print('Recall per class: ', recall)

# false positive rate
```

```
fpr = FP / (FP + TN)
print('False positive rate per class: ', fpr)
# false negative rate
fnr = FN / (TP + FN)
print('False negative rate per class: ', fnr)
# classification error
c_{error} = (FP + FN) / (TP + FP + FN + TN)
print('The classification error of each class: ' ,c_error)
# overall accuracy
accuracy = (TP + TN) / (TP + FP + FN + TN)
print('The accuracy of each class: ' ,accuracy)
# Averages
print('\nAverage Recall : ' ,recall.sum()/2)
print('Average Precision : ' ,precision.sum()/2)
print('Average Miss Rate : ' ,fnr.sum()/2)
print('Average Classification error : ' ,c_error.sum()/2)
print('Average accuracy : ' ,accuracy.sum()/2)
    Precision per class: [0.65097444 0.67346939]
    Recall per class: [0.99381762 0.0233711 ]
    False positive rate per class: [0.9766289 0.00618238]
    False negative rate per class: [0.00618238 0.9766289
    The classification error of each class: [0.34875 0.34875]
    The accuracy of each class: [0.65125 0.65125]
     Average Recall: 0.5085943622997404
    Average Precision : 0.662221912303266
    Average Miss Rate : 0.49140563770025963
     Average Classification error: 0.34875
    Average accuracy: 0.65125
```

▼ Tuning Decision Tree

Tuning the decision tree and applying cross validation technique to see if we can find a better result

```
print('Tunned Decision Tree Parameters {}'.format(tunned_tree_cv.best_params_))
  print('Best score: {}'.format(tunned_tree_cv.best_score_))
       Tunned Decision Tree Parameters {'max_depth': 1, 'criterion': 'gini'}
       So far the model accuracy is not good. Lets try random forest algorithm to see if we can find a better model with better accuracy
  We will also perform some hyper parameter tuning to get a better model

    Random Forest

  A random forest is a meta estimator that fits a number of decision tree classifiers on various sub-samples of the dataset and uses averaging to
  improve the predictive accuracy and control over-fitting. Takes the average of many Decision Trees via bagging.
  n_estmators : number of trees in a forest
  max_depth : the maximum depth of the tree
  max_features : maximum number of features to consider when looking for the best split
  min_samples_split : minimum number of samples required to split an internal node
  min_samples_leaf : minimum number of samples required to be at a leaf node
  from sklearn.ensemble import RandomForestClassifier
  forest = RandomForestClassifier(n_estimators=300, criterion='entropy')
  forest.fit(X_train, y_train)
  prediction_test = forest.predict(X=X_test)
  # Training Accuracy Of Random Forest
  print("Training Accuracy : ", forest.score(X_train, y_train))
```

Training Accuracy : 0.98975 Testing Accuracy : 0.6011661807580175

Test Accuracy Of Random Forest

The model overfitted since we did not define any max_depth

print("Testing Accuracy : ", forest.score(X_test, y_test))

Randomized Search CV

Random Search. Define a search space as a bounded domain of hyperparameter values and randomly sample points in that domain

```
# Number of trees in random forest
n_estimators = [int(x) for x in np.linspace(start = 20, stop = 300, num = 10)]
# Number of features to consider at every split
max_features = ['auto', 'sqrt','log2']
# Maximum number of levels in tree
max_depth = [int(x) for x in np.linspace(5, 100,5)]
# Minimum number of samples required to split a node
```

```
min_samples_split = [2, 3, 5, 7, 9, 10, 11, 14]
# Minimum number of samples required at each leaf node
min_samples_leaf = [1, 2, 4, 6, 7, 8]
# Create the random grid
random_grid = {'n_estimators': n_estimators,
  'max_features': max_features,
  'max_depth': max_depth,
  'min_samples_split': min_samples_split,
  'min_samples_leaf': min_samples_leaf,
  'criterion':['entropy','gini']
print(random_grid)
rand_forest = RandomForestClassifier()
rand_forest_randomcv = RandomizedSearchCV(estimator=rand_forest,param_distributions=random_grid,
                                            n_iter=100,cv=10,verbose=2, random_state=100,n_jobs=-1)
# fit the randomized model
rand_forest_randomcv.fit(X_train,y_train)
    Fitting 10 folds for each of 100 candidates, totalling 1000 fits
    RandomizedSearchCV(cv=10, estimator=RandomForestClassifier(), n_iter=100,
                       n_jobs=-1,
                       param_distributions={'criterion': ['entropy', 'gini'],
                                            'max_depth': [5, 28, 52, 76, 100],
                                            'max_features': ['auto', 'sqrt',
                                                            'log2'],
                                            'min_samples_leaf': [1, 2, 4, 6, 7, 8],
                                            'min_samples_split': [2, 3, 5, 7, 9, 10,
                                                                 11, 14],
                                            'n_estimators': [20, 51, 82, 113, 144,
                                                            175, 206, 237, 268,
                                                            300]},
                       random_state=100, verbose=2)
# best parameters
rand_forest_randomcv.best_params_
    {'criterion': 'gini',
      'max_depth': 100,
      'max_features': 'sqrt',
      'min_samples_leaf': 8,
      'min_samples_split': 10,
      'n_estimators': 206}
# best estimator
rand_forest_randomcv.best_estimator_
    RandomForestClassifier(max_depth=100, max_features='sqrt', min_samples_leaf=8,
                          min_samples_split=10, n_estimators=206)
best_random_grid = rand_forest_randomcv.best_estimator_
y_pred=best_random_grid.predict(X_test)
print(confusion_matrix(y_test,y_pred))
```

```
print("Accuracy Score {}".format(accuracy_score(y_test,y_pred)))
 print("Classification report: \n{}".format(classification_report(y_test,y_pred)))
      [[1089
              2]]
       「 624
      Accuracy Score 0.6361516034985423
      Classification report:
                   precision recall f1-score support
                                                    1089
                        0.64
                                 1.00
                                          0.78
                        1.00
                                 0.00
                                          0.01
                                                    626
                                          0.64
                                                   1715
          accuracy
         macro avg
                       0.82
                                 0.50
                                          0.39
                                                    1715
                                                   1715
                       0.77
                                          0.50
      weighted avg
                                 0.64

    Hyperparameter Tuning

 from itertools import product
 n_{estimators} = [1, 2, 4, 8, 16, 32, 64, 100, 200, 300, 500]
 max_features = ['auto', 'sqrt', 'log2']
 max_depths = [None, 1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 13, 15]
 train_results = []
  test_results = []
 # to iterate through all possible combinations
  for feature, depth in product(max_features, max_depths):
    for estimator in n_estimators:
      tunned_forest = RandomForestClassifier(n_estimators=estimator,
                                      criterion='entropy',
                                     max_features=feature,
                                      max_depth=depth,
                                     n_jobs=1,
                                     random_state=30)
```

tunned_forest.fit(X_train, y_train)

train_results.append(roc_auc)

test_results.append(roc_auc)

prediction_train = tunned_forest.predict(X=X_train)

prediction_test = tunned_forest.predict(X=X_test)

Checking classification accuracy of each tree

print('For n_estimators : ' ,estimator)

roc_auc = auc(false_positive_rate, true_positive_rate)

roc_auc = auc(false_positive_rate, true_positive_rate)

false_positive_rate, true_positive_rate, thresholds = roc_curve(y_train, prediction_train)

false_positive_rate, true_positive_rate, thresholds = roc_curve(y_test, prediction_test)

.format(feature, depth, accuracy_score(y_train, prediction_train)))

.format(feature, depth, accuracy_score(y_test, prediction_test)))

print('Classification accuracy on Train set with max_features = {} and max_depth = {}: Accuracy: = {}'

print('Classification accuracy on test set with max_features = {} and max_depth = {}: Accuracy: = {}'

```
print()
# Generating confusion matrix
c_matrix = confusion_matrix(y_test, prediction_test)
c_matrix_norm = c_matrix/c_matrix.sum(axis=1)[:, np.newaxis]
#plt.figure()
#plot_confusion_matrix(c_matrix_norm, classes=tunned_forest.classes_,
     title='Classification accuracy on test set with max_features = {} and max_depth = {}: Accuracy = {}'
                        .format(feature, depth, accuracy_score(y_test, prediction_test)))
For n_estimators : 1
Classification accuracy on Train set with max_features = auto and max_depth = None: Accuracy: = 0.83175
Classification accuracy on test set with max_features = auto and max_depth = None: Accuracy: = 0.5317784256559767
For n_estimators : 2
Classification accuracy on Train set with max_features = auto and max_depth = None: Accuracy: = 0.833
Classification accuracy on test set with max_features = auto and max_depth = None: Accuracy: = 0.5790087463556851
For n_estimators : 4
Classification accuracy on Train set with max_features = auto and max_depth = None: Accuracy: = 0.8965
Classification accuracy on test set with max_features = auto and max_depth = None: Accuracy: = 0.5865889212827988
For n_estimators : 8
Classification accuracy on Train set with max_features = auto and max_depth = None: Accuracy: = 0.95225
Classification accuracy on test set with max_features = auto and max_depth = None: Accuracy: = 0.5830903790087464
For n_estimators : 16
Classification accuracy on Train set with max_features = auto and max_depth = None: Accuracy: = 0.98
Classification accuracy on test set with max_features = auto and max_depth = None: Accuracy: = 0.5819241982507288
For n_estimators : 32
Classification accuracy on Train set with max_features = auto and max_depth = None: Accuracy: = 0.9885
Classification accuracy on test set with max_features = auto and max_depth = None: Accuracy: = 0.5906705539358601
For n_estimators : 64
Classification accuracy on Train set with max_features = auto and max_depth = None: Accuracy: = 0.98975
Classification accuracy on test set with max_features = auto and max_depth = None: Accuracy: = 0.5924198250728863
For n_estimators : 100
Classification accuracy on Train set with max_features = auto and max_depth = None: Accuracy: = 0.98975
Classification accuracy on test set with max_features = auto and max_depth = None: Accuracy: = 0.5935860058309038
For n_estimators : 200
Classification accuracy on Train set with max_features = auto and max_depth = None: Accuracy: = 0.98975
Classification accuracy on test set with max_features = auto and max_depth = None: Accuracy: = 0.597667638483965
For n_estimators : 300
Classification accuracy on Train set with max_features = auto and max_depth = None: Accuracy: = 0.98975
Classification accuracy on test set with max_features = auto and max_depth = None: Accuracy: = 0.5930029154518951
For n_estimators : 500
Classification accuracy on Train set with max_features = auto and max_depth = None: Accuracy: = 0.98975
Classification accuracy on test set with max_features = auto and max_depth = None: Accuracy: = 0.5982507288629738
For n_estimators : 1
Classification accuracy on Train set with max_features = auto and max_depth = 1: Accuracy: = 0.647
Classification accuracy on test set with max_features = auto and max_depth = 1: Accuracy: = 0.6349854227405248
For n_estimators : 2
Classification accuracy on Train set with max_features = auto and max_depth = 1: Accuracy: = 0.647
Classification accuracy on test set with max_features = auto and max_depth = 1: Accuracy: = 0.6349854227405248
For n_estimators : 4
```

```
Classification accuracy on Train set with max_features = auto and max_depth = 1: Accuracy: = 0.647
Classification accuracy on test set with max_features = auto and max_depth = 1: Accuracy: = 0.6349854227405248

For n_estimators : 8
Classification accuracy on Train set with max_features = auto and max_depth = 1: Accuracy: = 0.647
```

Random Forest

```
Building a RandomForest model based on best parameters
```

Training Accuracy: 0.72825

Testing Accuracy: 0.6361516034985423

print(classification_report(y_test, prediction_test))

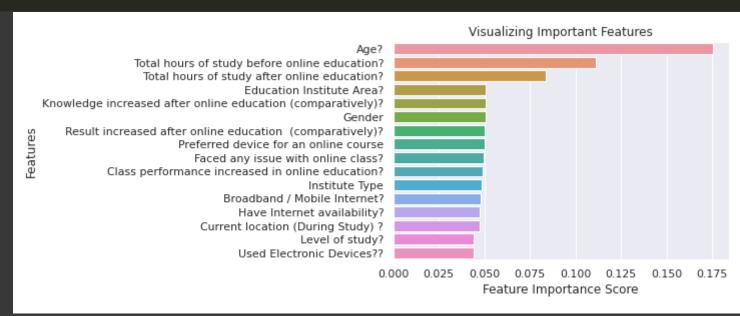
	precision	recall	f1-score	support
0 1	0.64 0.57	0.99 0.01	0.78 0.03	1089 626
accuracy macro avg weighted avg	0.60 0.61	0.50 0.64	0.64 0.40 0.50	1715 1715 1715

feature scores
feature_scores = pd.Series(rand_forest.feature_importances_, index=X_train.columns).sort_values(ascending=False)
feature_scores

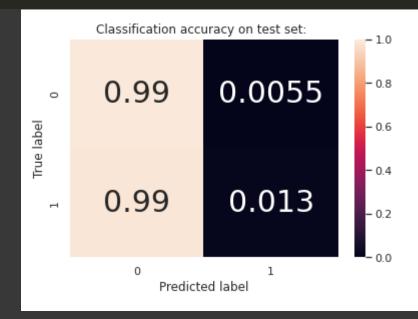
Age? 0.175185 Total hours of study before online education? 0.111110 Total hours of study after online education? 0.083414 Education Institute Area? 0.050592 Knowledge increased after online education (comparatively)? 0.050544 Gender 0.050532 Result increased after online education (comparatively)? 0.050390 Preferred device for an online course 0.050249 Faced any issue with online class? 0.049787 Class performance increased in online education? 0.049142 Institute Type 0.048337 Broadband / Mobile Internet? 0.047752 Have Internet availability? 0.047577 0.047188 Current location (During Study) ?

```
Level of study? 0.044355
Used Electronic Devices?? 0.043845
dtype: float64
```

```
# seaborn bar plot
sns.barplot(x=feature_scores, y=feature_scores.index)
plt.xlabel('Feature Importance Score')
plt.ylabel('Features')
plt.title("Visualizing Important Features")
plt.show()
```



▼ Confusion Matrix



```
c_matrix_rand
```

array([[1083, 6],

[618, 8]])

>