# MLBA\_Assignment\_2\_Readme Created by Group 66

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#### Run the code:

Usage: python <filename.py> --files <full testdata filename.csv> <full traindata filename>

Example: python mlba\_cancer\_pred.py --files cancer\_dataset/kaggle\_train.csv cancer\_dataset/kaggle\_test.csv

**Aim:** Classify high and low risk cancer patients.

Provided data: test, train datasets, and sample submission file

**Methodology:** We have tried using various machine learning techniques like random forest, gradient boost etc, on the given train dataset.

The maximum performance was obtained for the random forest classifier, which has given us an accuracy of around ~81.6 percent. We have also used grid searches to find the best hyperparameters and avoid overfitting.

## Steps:

Data importing and preprocessing

We had the kaggle\_train.csv. We used the pandas library to read the CSV file.

```
9 #Downloading the given packages
10 '''import subprocess
packages_to_install = ['pandas', 'numpy','sklearn','imbalanced-learn']
12 for package in packages_to_install:
     subprocess.check_call(['pip', 'install', package])'''
13
15 #importing the necessary packages
import argparse
import pandas as pd
18 import numpy as np
from sklearn.model_selection import train_test_split
from sklearn.ensemble import RandomForestClassifier
21 from sklearn.metrics import roc_auc score
from sklearn.preprocessing import StandardScaler from sklearn.feature_selection import RFE
24 import csv
25 #Parsing the data from command line
parser=argparse.ArgumentParser()
parser.add_argument("--files",nargs='+',help="Train and test files")
28 args = parser.parse args()
29 args=args.f
30 train_csv=args[0]
31 test_csv=args[1]
32 #reading the training csv file
33 train_data = pd.read_csv(train_csv)
```

# • Test Train Split

We then applied the test train split function to split the dataset into a test and train set. We have taken the value of the hyperparameter as 100 and random state as 42. We are splitting the dataset into the ratio 80:20

```
30
31 #Taking in the labels from the train_data
32 labels = train_data["Labels"]
33
34
35 scaler = StandardScaler()
36 #Applying scaling to the training dataset
37 X_train = scaler.fit_transform(X_train)
```

# Creating the model

Recursive feature elimination(RFE):

RFE is popular because it is easy to configure and use and because it is effective at selecting those features (columns) in a training dataset that are more or most relevant in predicting the target variable. The RFE technique would start with the original number of features and would select features by recursively eliminating less important features.

### Final Output

We applied the rf classifier model to the test dataset and have predicted the probability of the person having cancer . We saved the file in the output folder in the attached folder. We consider the AUC ROC score and hence we have found out the probability of a particular row belonging to cancer or not.

```
65
66 #Predicting the final output probability
67 final_pred =rf_classifier.predict_proba(test_data)[:, 1]
68
69 #Writing the output predictions onto the file ./output_rf.csv
70 with open('./output_rf.csv', mode='w', newline='') as file:
71 writer = csv.writer(file)
72 writer.writerow(['ID', 'Labels'])
73 for i in range(0,len(test_ID)):
74 writer.writerow([test_ID[i],final_pred[i]])
75
```

# **Directory structure:**

