Healthcare -Capstone Project Done by Mr.Arun Chougale

1. Data Exploration part1:

```
In [1]: import pandas as pd
         import numpy as np
         import seaborn as sns
         import matplotlib.pyplot as plt
         %matplotlib inline
In [2]: dataset=pd.read_csv('health care diabetes.csv')
In [3]: dataset.head()
Out[3]:
             Pregnancies Glucose BloodPressure SkinThickness Insulin BMI DiabetesPedigreeFunction Age Outcome
                                                                      33.6
                                                                                              0.627
                              85
                                             66
                                                           29
                                                                    0 26.6
                                                                                              0.351
          2
                      8
                             183
                                             64
                                                            0
                                                                    0 23.3
                                                                                              0.672
          3
                      1
                              89
                                             66
                                                           23
                                                                   94 28.1
                                                                                              0.167
                                                                                                     21
                                                                                                                0
                      0
                             137
                                             40
                                                           35
                                                                  168 43.1
                                                                                              2.288
                                                                                                     33
In [4]: dataset.shape
Out[4]: (768, 9)
In [5]: dataset['Outcome'].value_counts()
Out[5]: 0
               500
               268
         Name: Outcome, dtype: int64
In [6]: dataset.describe()
Out[6]:
                               Glucose BloodPressure SkinThickness
                 Pregnancies
                                                                         Insulin
                                                                                      BMI DiabetesPedigreeFunction
                                                                                                                          Age
                                                                                                                                 Outcome
          count
                 768.000000 768.000000
                                            768.000000
                                                          768.000000 768.000000
                                                                                768.000000
                                                                                                         768.000000 768.000000
                                                                                                                               768.000000
          mean
                    3.845052 120.894531
                                             69.105469
                                                           20.536458
                                                                      79.799479
                                                                                 31.992578
                                                                                                           0.471876
                                                                                                                     33.240885
                                                                                                                                 0.348958
                                             19.355807
                                                                                                                     11.760232
            std
                    3.369578
                              31.972618
                                                           15.952218 115.244002
                                                                                  7.884160
                                                                                                           0.331329
                                                                                                                                 0.476951
                    0.000000
                                             0.000000
                                                            0.000000
                                                                                                           0.078000
                                                                                                                     21 000000
                               0.000000
                                                                       0.000000
                                                                                  0.000000
                                                                                                                                 0.000000
            min
           25%
                    1.000000
                              99.000000
                                             62.000000
                                                            0.000000
                                                                                 27.300000
                                                                                                           0.243750
                                                                                                                     24.000000
                                                                                                                                 0.000000
                                                                       0.000000
           50%
                                            72.000000
                                                                                                                     29.000000
                    3.000000 117.000000
                                                           23.000000
                                                                      30.500000
                                                                                 32.000000
                                                                                                           0.372500
                                                                                                                                 0.000000
           75%
                    6.000000 140.250000
                                            80.000000
                                                           32.000000 127.250000
                                                                                 36.600000
                                                                                                           0.626250
                                                                                                                     41.000000
                                                                                                                                 1.000000
                   17.000000 199.000000
                                            122.000000
                                                           99.000000 846.000000
                                                                                 67.100000
                                                                                                           2.420000
                                                                                                                     81.000000
                                                                                                                                 1.000000
           max
```

Findings

Data has no negative values

We didnt get any null values here but need to check by using histograms

```
In [8]: plt.hist(dataset['Glucose'])
Out[8]: (array([ 5., 0., 4., 32., 156., 211., 163., 95., 56., 46.]),
         array([ 0., 19.9, 39.8, 59.7, 79.6, 99.5, 119.4, 139.3, 159.2, 179.1, 199. ]),
         <BarContainer object of 10 artists>)
         200
         175
         150
         125
         100
          75
          50
          25
                   25
                              75
                                   100
                                       125
                                             150
```

By this graph we can understood in between 19.9 to 39.8 '0' values observed and those are looking like missing values and need to treat them

Also here as per the details we can say mode value is average of 100 and 125, so we will replace this by 112.5

```
In [9]: dataset['Glucose']=dataset['Glucose'].replace(0.,112.5)
In [10]: dataset['Glucose']
Out[10]: 0
                148.0
                 85.0
         2
                183.0
         3
                 89.0
         4
                137.0
         763
                101.0
                122.0
         765
                121.0
         766
                126.0
                 93.0
         767
         Name: Glucose, Length: 768, dtype: float64
In [11]: plt.hist(dataset['Glucose'])
Out[11]: (array([ 4., 19., 87., 149., 166., 125., 88., 54., 44., 32.]),
          array([ 44., 59.5, 75., 90.5, 106., 121.5, 137., 152.5, 168., 183.5, 199.]),
          <BarContainer object of 10 artists>)
          160
          140
          120
           80
           60
           40
```

80 100 120 140 160 180

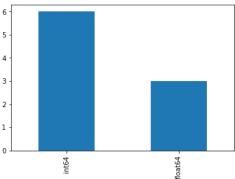
40

```
In [12]: plt.hist(dataset['BloodPressure'])
<BarContainer object of 10 artists>)
         250
         200
         150
         100
          50
                                               120
In [13]:
        plt.hist(dataset[['BloodPressure','Insulin','BMI','SkinThickness']])
Out[13]: (array([[656., 112.,
                            0.,
                                 0.,
                                                                 1.],
                [487., 155., 70., 30.,
                                       8.,
                                            9.,
                                                 5.,
                                                      1.,
                                                            2.,
                                      0.,
                                           0.,
                                                 0.,
                                                           0.,
               [768., 0., 0., 0.,
                                                      0.,
                                                                 0.],
                                     0.,
                                           0.,
                                                0.,
                                                      0., 0.,
                                                                0.]]),
               [767., 1., 0., 0.,
         array([ 0., 84.6, 169.2, 253.8, 338.4, 423., 507.6, 592.2, 676.8,
               761.4, 846. ]),
         <a list of 4 BarContainer objects>)
         800
         700
         600
         500
         400
         300
         200
                                              800
                             400
                                     600
In [14]: dataset['BloodPressure'].replace(0,69)
Out[14]: 0
              72
              66
        2
              64
        3
              66
        4
              40
        763
              76
        764
              70
              72
        765
        766
              60
        767
              70
        Name: BloodPressure, Length: 768, dtype: int64
```

```
In [15]: dataset['BloodPressure'].value_counts()
Out[15]: 70
              52
        78
              45
              45
        68
        72
              44
        64
              43
        80
              40
        76
              39
              37
        60
        0
              35
        62
        66
              30
        82
              30
        88
              25
        84
              23
              22
        86
              21
        58
50
              21
              13
              12
        52
              11
        54
              11
        75
92
               8
               8
        65
               7
        85
               6
        94
               6
        48
               5
        44
        100
               3
        106
               3
3
        98
        110
               2
        55
        108
        104
               2
        46
        30
        122
        95
               1
        102
        61
        24
        38
               1
        40
               1
        Name: BloodPressure, dtype: int64
In [16]: plt.hist(dataset['BloodPressure'])
250
         200
         150
         100
         50
```

```
In [17]: dataset['Glucose'].value_counts().head(20)
Out[17]: 99.0
                       100.0
                                             17
                       111.0
                                             14
                       129.0
                                            14
                       125.0
                                             14
                       106.0
                                             14
                       112.0
                                             13
                       108.0
                                            13
                       95.0
                                            13
                       105.0
                                            13
                       102.0
                                            13
                       122.0
                                            12
                       109.0
                                            12
                       117.0
                                            11
                       124.0
                                             11
                       90.0
                                            11
                       107.0
                                            11
                       128.0
                                            11
                       120.0
                                            11
                       119.0
                                             11
                       Name: Glucose, dtype: int64
In [18]: dataset.info()
                       <class 'pandas.core.frame.DataFrame'>
                       RangeIndex: 768 entries, 0 to 767
                      Data columns (total 9 columns):
                                                                                                  Non-Null Count
                        #
                                   Column
                                                                                                                                         Dtype
                         0
                                   Pregnancies
                                                                                                  768 non-null
                                                                                                                                         int64
                         1
                                   Glucose
                                                                                                  768 non-null
                                                                                                                                         float64
                                   BloodPressure
                                                                                                  768 non-null
                                                                                                                                         int64
                                  SkinThickness
                                                                                                  768 non-null
                                                                                                                                         int64
                         4
                                   Insulin
                                                                                                  768 non-null
                                                                                                                                         int64
                                   BMI
                                                                                                  768 non-null
                                                                                                                                         float64
                         6
                                   DiabetesPedigreeFunction
                                                                                                  768 non-null
                                                                                                                                         float64
                                                                                                  768 non-null
                                                                                                                                         int64
                                   Age
                                                                                                                                         int64
                        8
                                  Outcome
                                                                                                  768 non-null
                       dtypes: float64(3), int64(6)
                       memory usage: 54.1 KB
In [19]: dataset.dtypes
Out[19]: Pregnancies
                                                                                                int64
                       Glucose
                                                                                           float64
                      BloodPressure
                                                                                                int64
                       SkinThickness
                                                                                                int64
                       Insulin
                                                                                                int64
                                                                                           float64
                      DiabetesPedigreeFunction
                                                                                           float64
                                                                                                int64
                       Age
                      Outcome
                                                                                                int64
                       dtype: object
In [20]: sns.countplot('Glucose',data=dataset)
                       \hbox{$C:\Users\Admin\anaconda3\lib\site-packages\seaborn\_decorators.py:36: Future\Warning: Pass the following variable as a keyword and a substitution of the packages of the following variable as a keyword and the packages of the following variable as a keyword and the packages of the
                       rg: x. From version 0.12, the only valid positional argument will be `data`, and passing other arguments without an explicit ke
                       yword will result in an error or misinterpretation.
                           warnings.warn(
Out[20]: <AxesSubplot:xlabel='Glucose', ylabel='count'>
                             16
                             14
                             12
                             10
```

Glucose



Findings

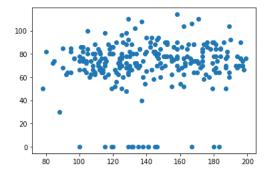
1)Here I observed there only two types of attributes, one is int64 and another is float64 and how they are distributed.

2)6 attributes are int64 and 3 attributes are float64.

Data Exploration part-2

```
In [23]: data1=dataset[dataset['Outcome']==1]
    data0=dataset[dataset['Outcome']==0]
In [24]: plt.scatter(data1['Glucose'],data1['BloodPressure'])
```

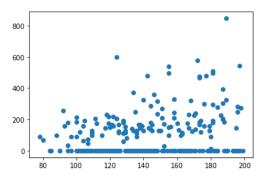
Out[24]: <matplotlib.collections.PathCollection at 0x29434182250>



Findings-

The person who has blood-pressure above 60 ,may have diabeties too.

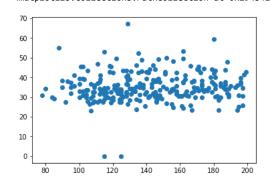
In [25]: plt.scatter(data1['Glucose'],data1['Insulin'])
Out[25]: cmatplotlib.collections.PathCollection at 0x294341e3610>



Findings-

The person who have insulin lower side ,they may have diabeties.

In [26]: plt.scatter(data1['Glucose'],data1['BMI'])
Out[26]: <matplotlib.collections.PathCollection at 0x29434250880>

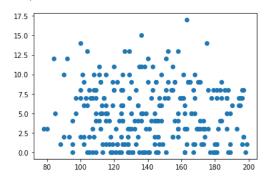


Findings-

The person whose BMI is above 30, may be have some diabetic condition with him/her.

In [27]: plt.scatter(data1['Glucose'],data1['Pregnancies'])

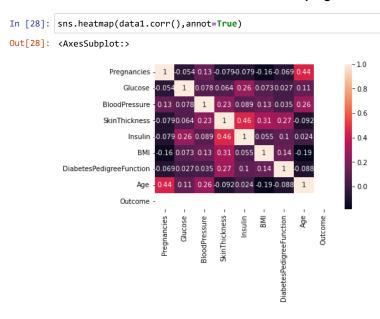
Out[27]: <matplotlib.collections.PathCollection at 0x29435287cd0>



Findings-

Normally in the pregnancies, glocose level increases.

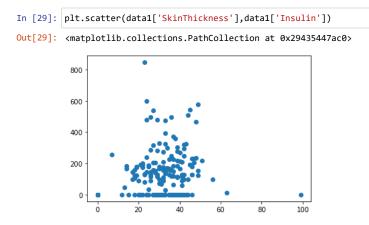
But there is no relation between number of pregnancies and diabeties.



###Findings-

Skinthikness and Insuline have more correlation.

so, need to observe its realtion by scatter plot



Finding-

Theperson whose skinthckness between 20 and 40 ,have less insuline in his body.

```
In [30]: data1.shape
Out[30]: (268, 9)
In [31]: data0.shape
Out[31]: (500, 9)
```

Findings-

dataset is imbalanced

ó

Need to do data augmentation or data resampling for balanced data

i

Outcome

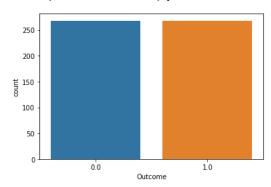
```
In [34]: #!pip install imblearn
In [35]: from imblearn.under_sampling import RandomUnderSampler
         from imblearn.over_sampling import RandomOverSampler, SMOTE
In [36]: print("The Number of Samples in the dataset: ", len(dataset))
         print('Class 0
                               :', round(dataset['Outcome'].value_counts()[0]
                               /len(dataset) * 100, 2), '% of the dataset')
         print('Class 1(Fraud) :', round(dataset['Outcome'].value_counts()[1]
                               /len(dataset) * 100, 2), '% of the dataset')
         The Number of Samples in the dataset: 768
                        : 65.1 % of the dataset
         Class 0
         Class 1(Fraud) : 34.9 % of the dataset
In [37]: X_data=dataset.iloc[:,:-1]
         Y_data=dataset.iloc[:,-1:]
In [38]: dataset['Outcome'].value_counts()
Out[38]: 0
              500
              268
         Name: Outcome, dtype: int64
In [42]: rus=RandomUnderSampler(random_state=42)
In [43]: X_res,Y_res=rus.fit_resample(X_data,Y_data)
         X_res = pd.DataFrame(X_res)
         Y_res = pd.DataFrame(Y_res)
         print("After Under Sampling Of Major Class Total Samples are :", len(Y_res))
         print('Class 0:', round(Y_res.value_counts()[0]
                         /len(Y_res) * 100, 2), '% of the dataset')
         print('Class 1(Fraud):', round(Y_res.value_counts()[1]
                         /len(Y_res) * 100, 2), '% of the dataset')
         After Under Sampling Of Major Class Total Samples are : 536
         Class 0: 50.0 % of the dataset
         Class 1(Fraud): 50.0 % of the dataset
In [44]: resampled_data=pd.concat([X_res,Y_res])
```

In [45]: sns.countplot(resampled_data['Outcome'])

C:\Users\Admin\anaconda3\lib\site-packages\seaborn_decorators.py:36: FutureWarning: Pass the following variable as a keyword a rg: x. From version 0.12, the only valid positional argument will be `data`, and passing other arguments without an explicit ke yword will result in an error or misinterpretation.

warnings.warn(

Out[45]: <AxesSubplot:xlabel='Outcome', ylabel='count'>



- 1. Now here we can data set is balanced, i.e, 50% of class 0 and 50% of class 1
- 2.For getting balanced data we use here resampling technique of undersampling, we can use oversampling too.
- 3. For building a model now we are going to use balanced dataset.

In [46]: Y_res.shape Out[46]: (536, 1) In [47]: X_res Out[47]: Pregnancies Glucose BloodPressure SkinThickness Insulin BMI DiabetesPedigreeFunction Age 30 0 97.0 70 40 38.1 0.218 0 78.0 5 48 0 0 33.7 0.654 25 2 3 111.0 31 22 58 44 29.5 0.430 3 2 129.0 84 0 0.284 27 0 28.0 102.0 40 0.204 45 74 105 37.2 531 128.0 88 39 110 36.5 1.057 37 532 123.0 72 0 36.3 0.258 52 533 92 35.5 0.278 534 170.0 74 31 0 44.0 0.403 43 535 60 0 30.1 0.349 47 536 rows × 8 columns

Data Modelling-part1

In [48]: from sklearn.model_selection import train_test_split
In [49]: x_train,x_test,y_train,y_test=train_test_split(X_res,Y_res,test_size=0.3,random_state=42)

```
In [50]: x_train
Out[50]:
                Pregnancies Glucose BloodPressure SkinThickness Insulin BMI DiabetesPedigreeFunction Age
            57
                         3
                              180.0
                                               64
                                                             25
                                                                    70
                                                                       34 0
                                                                                              0.271
                                                                                                     26
           227
                              165.0
                                                                                              0.631
                                                                                                     49
                         6
                                               68
                                                             26
                                                                   168
                                                                       33.6
                         6
                              111.0
                                               64
                                                             39
                                                                                                     24
            24
                                                                    0 34.2
                                                                                              0.260
            17
                         0
                              100.0
                                               88
                                                             60
                                                                                              0.962
                                                                                                     31
                                                                   110
                                                                       46.8
           210
                         2
                              123.0
                                               48
                                                             32
                                                                                                     26
                                                                   165
                                                                       42.1
                                                                                              0.520
            71
                         0
                              111.0
                                               65
                                                             0
                                                                     0
                                                                       24.6
                                                                                              0.660
                                                                                                     31
           106
                              126.0
                                               74
                                                             38
                                                                       25.9
                                                                                              0.162
           270
                         0
                              137.0
                                               40
                                                             35
                                                                   168
                                                                       43.1
                                                                                              2.288
                                                                                                     33
           435
                         0
                              189.0
                                              104
                                                            25
                                                                     0 34.3
                                                                                              0.435
                                                                                                     41
           102
                              133.0
                                               84
                                                             0
                                                                     0 40.2
                                                                                              0.696
                                                                                                     37
          375 rows × 8 columns
In [51]: x_test
Out[51]:
                Pregnancies Glucose BloodPressure SkinThickness Insulin BMI DiabetesPedigreeFunction
                                                                                                    Age
           117
                              109.0
                                               60
                                                             8
                                                                   182
                                                                       25.4
                                                                                              0.947
                                                                                                     21
           132
                               83.0
                                               86
                                                             19
                                                                     0 29.3
                                                                                              0.317
                                                                                                     34
           154
                         6
                               80.0
                                               80
                                                             36
                                                                     0
                                                                       39.8
                                                                                              0.177
                                                                                                     28
           245
                         4
                              151.0
                                               90
                                                             38
                                                                       29.7
                                                                                              0.294
                                                                                                      36
            84
                         3
                               96.0
                                               78
                                                             39
                                                                     0
                                                                       37.3
                                                                                              0.238
                                                                                                      40
            31
                         3
                              125.0
                                               58
                                                             0
                                                                     0
                                                                       31.6
                                                                                              0.151
                                                                                                     24
           113
                              103.0
                                               80
                                                             11
                                                                    82 194
                                                                                              0.491
                                                                                                     22
           496
                         6
                              115.0
                                               60
                                                             39
                                                                     0 33.7
                                                                                              0.245
                                                                                                     40
                                                             37
                        10
                                               86
                                                                                                     38
           371
                              101.0
                                                                     0 45.6
                                                                                              1.136
           483
                         0
                              119 0
                                               0
                                                                     0 324
                                                                                              0 141
                                                                                                     24
          161 rows × 8 columns
In [52]: | from sklearn.tree import DecisionTreeClassifier
In [53]: CDT=DecisionTreeClassifier(criterion='gini')
In [54]: CDT.fit(x_train,y_train)
Out[54]: DecisionTreeClassifier()
          In a Jupyter environment, please rerun this cell to show the HTML representation or trust the notebook.
          On GitHub, the HTML representation is unable to render, please try loading this page with nbviewer.org.
In [55]: y_pred=CDT.predict(x_test)
In [56]: y_pred
Out[56]: array([0, 0, 0, 0, 0, 1, 0, 1, 1, 0, 0, 1, 0, 0, 0, 0, 1, 1, 0, 1, 0, 1,
                  0, 1, 0, 0, 1, 1, 1, 1, 0, 1, 1, 0, 0, 0, 0, 1, 0, 0, 0, 1,
                     0, 1, 1, 0, 0,
                                      0, 1, 0, 1, 0, 1, 1, 1,
                                                                1, 1,
                                                                       0, 0, 0,
                     0, 0, 1, 0, 1, 0, 1, 0, 0, 1, 0, 1, 0, 0, 1, 0, 0, 0,
                  0, 1, 1, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 1, 0, 1, 1, 1, 1, 1, 1,
                  0, 0, 1, 0, 1, 0, 0, 1, 0, 1, 0, 0, 1, 0, 0, 0, 1, 1, 1, 0, 0, 1,
                  1, 0, 1, 1, 0, 0, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 0, 1, 1, 1, 0,
                  0, 0, 1, 0, 0, 1, 0], dtype=int64)
          PERFORMANCE ANALYSIS
```

In [57]: from sklearn.metrics import confusion_matrix,classification_report,accuracy_score

```
In [58]: print("Confusion_matrix",classification_report(y_pred,y_test))
                                                     recall f1-score support
         Confusion_matrix
                                        precision
                            0.82
                                      0.67
                                                0.74
                                                             92
                            0.65
                                      0.80
                                                0.71
                                                             69
                    1
             accuracy
                                                0.73
                                                            161
                            0.73
                                      0.74
            macro avg
                                                            161
                            0.74
                                                0.73
         weighted avg
                                      0.73
                                                            161
In [59]: |print(accuracy_score(y_pred,y_test))
         0.7267080745341615
In [60]: x_pred=CDT.predict(x_train)
In [61]: print(accuracy_score(x_pred,y_train))
         1.0
```

Now I observing here overfitting condition because, this model has accuracy for train data is 1 but for test data is 0.7, so to avoid this overfitting condition we have to apply below techninques

- a.Pruning the DT
- b.Entrpy instead of Gini
- c.Crossvalidation
- d.GridsearchCV
- b.Entropy instead of gini

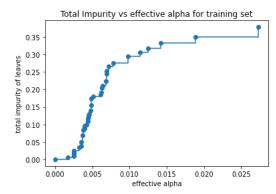
so,here I observed by using criterion entropy instead of gini,some improvement in accuracy but not as expected

a.Pruining

```
In [66]: path=DecisionTreeClassifier(random_state=1).\
    cost_complexity_pruning_path(x_train, y_train)
    ccp_alphas, impurities = path.ccp_alphas, path.impurities
```

```
In [67]: fig, ax = plt.subplots()
    ax.plot(ccp_alphas[:-1], impurities[:-1], marker='o', drawstyle="steps-post")
    ax.set_xlabel("effective alpha")
    ax.set_ylabel("total impurity of leaves")
    ax.set_title("Total Impurity vs effective alpha for training set")
```

Out[67]: Text(0.5, 1.0, 'Total Impurity vs effective alpha for training set')



As it is evident from the above plot, the zero value of alpha corresponds to minimum impurity(unpruned tree) and as the value of alpha tends to infinity the tree tends to be more impure.

Next, we build a forest of trees with different values of ccp_alpha values extracted from cost_complexity_pruning_path in order. The last tree will be the root node.

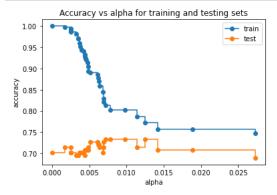
```
In [68]: clfs = []
    for ccp_alpha in ccp_alphas:
        clf = DecisionTreeClassifier(random_state=1,ccp_alpha=ccp_alpha)
        clf.fit(x_train, y_train)
        clfs.append(clf)
```

In [69]: print("Number of nodes in the last tree is: {} with ccp_alpha: {} and a depth of: {}".format(
 clfs[-1].tree_.node_count, ccp_alphas[-1].clfs[-1].tree_.max_depth))

Number of nodes in the last tree is: 1 with ccp_alpha: 0.121420542713568 and a depth of: 0

```
In [70]: clfs = clfs[:-1]
ccp_alphas = ccp_alphas[:-1]
```

```
In [71]:
    train_scores = [clf.score(x_train, y_train) for clf in clfs]
    test_scores = [clf.score(x_test, y_test) for clf in clfs]
    fig, ax = plt.subplots()
    ax.set_xlabel("alpha")
    ax.set_ylabel("accuracy")
    ax.set_title("Accuracy vs alpha for training and testing sets")
    ax.plot(ccp_alphas, train_scores, marker='o', label="train",drawstyle="steps-post")
    ax.plot(ccp_alphas, test_scores, marker='o', label="test",drawstyle="steps-post")
    ax.legend()
    plt.show()
```



```
In [72]: import sklearn.tree as tree
      clf=DecisionTreeClassifier(random_state=0,ccp_alpha=0.02)
      clf.fit(x_train,y_train)
      plt.figure(figsize=(12,8))
      tree.plot_tree(clf,rounded=True,filled=True)
      plt.show()
                      X[1] \le 124.5
                         gini = 0.5
                      samples = 375
                    value = [192, 183]
                                   X[5] \le 29.95
           gini = 0.381
                                    gini = 0.375
          samples = 199
                                  samples = 176
        value = [148, 51]
                                 value = [44, 132]
                        gini = 0.496
                                                gini = 0.257
                                              samples = 132
                       samples = 44
                                             value = [20, 112]
                     value = [24, 20]
```

In [73]: accuracy_score(y_test,clf.predict(x_test))

Out[73]: 0.7080745341614907

c)CROSS VALIDATION

```
In [74]: from sklearn.model_selection import cross_val_score
             from sklearn.linear_model import LogisticRegression
             #by kfold method
             # apply classifier
             clf = LogisticRegression()
             # get cv scores
             cv_scores = cross_val_score(clf,x_train,y_train, cv = 5)
             print('Cross validation scores (5 folds): {}'.format(cv_scores))
             print('The average cross validation score (5 folds): {} format(np.mean(cv_scores)))
             ## final result ##
             C:\Users\Admin\anaconda3\lib\site-packages\sklearn\utils\validation.py:1111: DataConversionWarning: A column-vector y was passe
             d when a 1d array was expected. Please change the shape of y to (n_samples, ), for example using ravel().
               y = column_or_1d(y, warn=True)
             C:\Users\Admin\anaconda3\lib\site-packages\sklearn\linear model\ logistic.py:444: ConvergenceWarning: lbfgs failed to converge
             (status=1):
             STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.
             Increase the number of iterations (max_iter) or scale the data as shown in:
                  https://scikit-learn.org/stable/modules/preprocessing.html (https://scikit-learn.org/stable/modules/preprocessing.html)
             Please also refer to the documentation for alternative solver options:
                  https://scikit-learn.org/stable/modules/linear_model.html#logistic-regression (https://scikit-learn.org/stable/modules/line
             ar_model.html#logistic-regression)
               n iter i = check optimize result(
             C:\Users\Admin\anaconda3\lib\site-packages\sklearn\utils\validation.py:1111: DataConversionWarning: A column-vector y was passe
             d when a 1d array was expected. Please change the shape of y to (n_samples, ), for example using ravel().
                y = column_or_1d(y, warn=True)
             C:\Users\Admin\anaconda3\lib\site-packages\sklearn\linear_model\_logistic.py:444: ConvergenceWarning: lbfgs failed to converge
             (status=1):
             STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.
             Increase the number of iterations (max_iter) or scale the data as shown in:
                  https://scikit-learn.org/stable/modules/preprocessing.html (https://scikit-learn.org/stable/modules/preprocessing.html)
             Please also refer to the documentation for alternative solver options:
                  https://scikit-learn.org/stable/modules/linear_model.html#logistic-regression (https://scikit-learn.org/stable/modules/line
             ar_model.html#logistic-regression)
               n_iter_i = _check_optimize_result(
             C:\Users\Admin\anaconda3\lib\site-packages\sklearn\utils\validation.py:1111: DataConversionWarning: A column-vector y was passe
             d when a 1d array was expected. Please change the shape of y to (n_samples, ), for example using ravel().
               y = column_or_1d(y, warn=True)
             C:\Users\Admin\anaconda3\lib\site-packages\sklearn\linear_model\_logistic.py:444: ConvergenceWarning: lbfgs failed to converge
             (status=1):
             STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.
             Increase the number of iterations (max_iter) or scale the data as shown in:
                  https://scikit-learn.org/stable/modules/preprocessing.html (https://scikit-learn.org/stable/modules/preprocessing.html)
             Please also refer to the documentation for alternative solver options:
                  https://scikit-learn.org/stable/modules/linear_model.html#logistic-regression (https://scikit-learn.org/stable/modules/line
             ar_model.html#logistic-regression)
               n_iter_i = _check_optimize_result(
             C:\Users\Admin\anaconda3\lib\site-packages\sklearn\utils\validation.py:1111: DataConversionWarning: A column-vector y was passe
             d when a 1d array was expected. Please change the shape of y to (n_samples, ), for example using ravel().
                y = column_or_1d(y, warn=True)
             C:\Users\Admin\anaconda3\lib\site-packages\sklearn\linear_model\_logistic.py:444: ConvergenceWarning: lbfgs failed to converge
             (status=1):
             STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.
             Increase the number of iterations (max_iter) or scale the data as shown in:
                  https://scikit-learn.org/stable/modules/preprocessing.html (https://scikit-learn.org/stable/modules/preprocessing.html)
             Please also refer to the documentation for alternative solver options:
                  https://scikit-learn.org/stable/modules/linear_model.html#logistic-regression (https://scikit-learn.org/stable/modules/line
             ar_model.html#logistic-regression)
                n_iter_i = _check_optimize_result(
             C:\Users\Admin\anaconda3\lib\site-packages\sklearn\utils\validation.py:1111: DataConversionWarning: A column-vector y was passe
             d when a 1d array was expected. Please change the shape of y to (n_samples, ), for example using ravel().
               y = column_or_1d(y, warn=True)
             Cross validation scores (5 folds): [0.69333333 0.76
                                                                                             0.84
                                                                                                             0.74666667 0.813333331
             The average cross validation score (5 folds): 0.770666666666667
             C:\Users\Admin\anaconda3\lib\site-packages\sklearn\linear_model\_logistic.py:444: ConvergenceWarning: lbfgs failed to converge
             (status=1):
             STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.
             Increase the number of iterations (max_iter) or scale the data as shown in:
                  https://scikit-learn.org/stable/modules/preprocessing.html (https://scikit-learn.org/stable/modules/preprocessing.html)
             Please also refer to the documentation for alternative solver options:
                  https://scikit-learn.org/stable/modules/linear_model.html#logistic-regression (https://scikit-learn.org/stable/modules/linear_model.html#logistic-regression (https://scikit-regression) (https://scikit-regression) (https://scikit-
             ar_model.html#logistic-regression)
```

D) GridSearchCv

n_iter_i = _check_optimize_result(

```
In [75]: from sklearn.neighbors import KNeighborsClassifier
         from sklearn.model_selection import GridSearchCV
         from sklearn.svm import SVC
In [76]: | param_grid = {'max_features': ['auto', 'sqrt', 'log2'],
                        ccp_alpha': [0.1, .01, .001],
                        'max_depth' : [5, 6, 7, 8, 9],
'criterion' :['gini', 'entropy']
         gscv_DT=GridSearchCV(estimator=CDT,param_grid=param_grid,cv=5,verbose=0)
         gscv_DT.fit(x_train,y_train)
         C:\Users\Admin\anaconda3\lib\site-packages\sklearn\tree\_classes.py:298: FutureWarning: `max_features='auto'` has been deprec
         ated in 1.1 and will be removed in 1.3. To keep the past behaviour, explicitly set `max_features='sqrt'
         C:\Users\Admin\anaconda3\lib\site-packages\sklearn\tree\_classes.py:298: FutureWarning: `max_features='auto'` has been deprec
         ated in 1.1 and will be removed in 1.3. To keep the past behaviour, explicitly set `max_features='sqrt'
           warnings.warn(
         C:\Users\Admin\anaconda3\lib\site-packages\sklearn\tree\_classes.py:298: FutureWarning: `max_features='auto'` has been deprec
         ated in 1.1 and will be removed in 1.3. To keep the past behaviour, explicitly set `max_features='sqrt'`.
           warnings.warn(
         C:\Users\Admin\anaconda3\lib\site-packages\sklearn\tree\_classes.py:298: FutureWarning: `max_features='auto'` has been deprec
         ated in 1.1 and will be removed in 1.3. To keep the past behaviour, explicitly set `max_features='sqrt'`.
         C:\Users\Admin\anaconda3\lib\site-packages\sklearn\tree\_classes.py:298: FutureWarning: `max_features='auto'` has been deprec
         ated in 1.1 and will be removed in 1.3. To keep the past behaviour, explicitly set `max_features='sqrt'`.
         C:\Users\Admin\anaconda3\lib\site-packages\sklearn\tree\_classes.py:298: FutureWarning: `max_features='auto'` has been deprec
         ated in 1.1 and will be removed in 1.3. To keep the past behaviour, explicitly set `max_features='sqrt'`.
           warnings.warn(
         C:\Users\Admin\anaconda3\lib\site-packages\sklearn\tree\_classes.py:298: FutureWarning: `max_features='auto'` has been deprec
In [77]: gscv_DT.best_params_
Out[77]: {'ccp_alpha': 0.01,
           criterion': 'gini',
           'max_depth': 8,
           'max_features': 'sqrt'}
In [78]: gscv_DT.best_score_
Out[78]: 0.7413333333333333
In [79]: GSCV_DT=DecisionTreeClassifier(ccp_alpha=0.01,
          criterion='gini',
          max_depth=8,
          max features='auto')
In [80]: GSCV_DT.fit(x_train,y_train)
         C:\Users\Admin\anaconda3\lib\site-packages\sklearn\tree\_classes.py:298: FutureWarning: `max_features='auto'` has been deprecat
         ed in 1.1 and will be removed in 1.3. To keep the past behaviour, explicitly set `max_features='sqrt'`.
           warnings.warn(
Out[80]: DecisionTreeClassifier(ccp_alpha=0.01, max_depth=8, max_features='auto')
         In a Jupyter environment, please rerun this cell to show the HTML representation or trust the notebook.
         On GitHub, the HTML representation is unable to render, please try loading this page with nbviewer.org.
In [81]: y_gscv_dt=GSCV_DT.predict(x_test)
In [82]: y_gscv_dt
Out[82]: array([0, 0, 0, 1, 0, 1, 0, 1, 0, 0, 1, 1, 0, 0, 0, 1, 1, 0, 1,
                0, 1, 0, 1, 1, 1, 1, 1, 1, 1, 1, 0, 1, 0, 1, 0, 0, 0, 0, 1,
                0, 0, 1, 1, 0, 0, 0, 0, 1, 0, 0, 1, 1, 0, 0, 1, 0, 0, 0, 1, 1, 1,
                0, 0, 0, 1, 1, 1, 0, 0, 0, 0, 1, 0, 1, 0, 1, 1, 0, 0, 1, 0,
                   0, 0, 0, 0, 0, 0, 0, 1, 0, 1, 0, 0, 1, 0, 0, 1, 0, 1, 0, 1,
                1, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 1, 0, 1, 0, 0, 1, 1, 1, 0, 0,
                1, 0, 1, 1, 0, 0, 0, 0, 1, 1, 1, 0, 0, 0, 1, 0, 1, 0, 1, 0,
                0, 1, 1, 0, 0, 0, 0], dtype=int64)
In [83]: print(accuracy_score(y_gscv_dt,y_test))
```

Validating result by SVM

0.6894409937888198

```
In [84]: from sklearn.svm import SVC
In [85]: SVM=SVC(kernel='linear', random_state=0)
In [86]: SVM.fit(x_train,y_train)
                        {\tt C:\Users\Admin\anaconda3\lib\site-packages\sklearn\utils\validation.py:1111:\ DataConversion\Warning:\ A\ column-vector\ y\ was\ passe}
                        d when a 1d array was expected. Please change the shape of y to (n_samples, ), for example using ravel().
                            y = column_or_1d(y, warn=True)
Out[86]: SVC(kernel='linear', random_state=0)
                        In a Jupyter environment, please rerun this cell to show the HTML representation or trust the notebook.
                        On GitHub, the HTML representation is unable to render, please try loading this page with nbviewer.org.
In [87]: y_svm=SVM.predict(x_test)
In [88]: y_svm
Out[88]: array([0, 0, 0, 1, 0, 1, 0, 1, 0, 0, 0, 1, 0, 0, 0, 1, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1,
                                         0, 1, 0, 1, 1, 1, 1, 0, 1, 1, 1, 0, 1, 1, 0, 1, 0, 0, 0, 0, 0,
                                         0, 0, 0, 1, 0, 0, 0, 0, 1, 0, 0, 1, 0, 0, 1, 0, 0, 1, 1, 1, 1,
                                         0, 0, 0, 1, 1, 1, 0, 1, 0, 0, 1, 0, 1, 0, 1, 1, 0, 0, 1, 0, 1, 0,
                                         0, 0, 1, 0, 0, 0, 0, 1, 0, 0, 1, 0, 0, 1, 0, 0, 0, 1, 1, 0, 0,
                                         1, 0, 0, 0, 0, 0, 1, 1, 0, 1, 1, 1, 0, 1, 0, 1, 1, 1, 1, 0, 0,
                                         1, 1, 1, 1, 0, 0, 0, 1, 0, 1, 0, 0, 0, 0, 0, 1, 0, 1, 1, 1, 1, 0,
                                         0, 1, 0, 0, 0, 1, 0], dtype=int64)
In [89]: print(accuracy_score(y_svm,y_test))
                        0.7391304347826086
                        now we will check the result after grid_search_cv
```

```
In [90]: params={'c':[5,10,15],'gamma':[0.01,0.02,0.05]}
         gscv_svm_2=GridSearchCV(estimator=SVM,param_grid=params,cv=5,verbose=0)
         gscv_svm_2.fit(x_train,y_train)
         ValueError
                                                    Traceback (most recent call last)
         Input In [90], in <cell line: 3>()
               1 params={'c':[5,10,15],'gamma':[0.01,0.02,0.05]}
               2 gscv_svm_2=GridSearchCV(estimator=SVM,param_grid=params,cv=5,verbose=0)
         ----> 3 gscv_svm_2.fit(x_train,y_train)
         File ~\anaconda3\lib\site-packages\sklearn\model_selection\_search.py:875, in BaseSearchCV.fit(self, X, y, groups, **fit_para
             869
                      results = self. format results(
             870
                         all_candidate_params, n_splits, all_out, all_more_results
             871
             873
                     return results
         --> 875 self._run_search(evaluate_candidates)
             877 # multimetric is determined here because in the case of a callable
             878 # self.scoring the return type is only known after calling
             879 first_test_score = all_out[0]["test_scores"]
         File ~\anaconda3\lib\site-packages\sklearn\model_selection\_search.py:1379, in GridSearchCV._run_search(self, evaluate_candid
 In [ ]: #Now we have to follow the same steps with different algorithms
         #KNN, RF, XGB,
 In [ ]: ## save the moodels with joblib,pickle
In [92]: from sklearn.neighbors import KNeighborsClassifier
         knn_classifier= KNeighborsClassifier(n_neighbors=5, metric='minkowski', p=2)
         knn_classifier.fit(x_train, y_train)
         C:\Users\Admin\anaconda3\lib\site-packages\sklearn\neighbors\_classification.py:207: DataConversionWarning: A column-vector y w
         as passed when a 1d array was expected. Please change the shape of y to (n_samples,), for example using ravel().
           return self._fit(X, y)
Out[92]: KNeighborsClassifier()
         In a Jupyter environment, please rerun this cell to show the HTML representation or trust the notebook.
         On GitHub, the HTML representation is unable to render, please try loading this page with nbviewer.org.
In [93]: y_pred_knn_classifier=knn_classifier.predict(x_test)
```

```
In [94]: print(accuracy_score(y_pred_knn_classifier,y_test))
```

0.7267080745341615

Now we check with Grid searchCV

```
In [95]: k_range = list(range(1, 31))
          param_grid = dict(n_neighbors=k_range)
          grid = GridSearchCV(knn_classifier, param_grid, cv=10, scoring='accuracy', return_train_score=False,verbose=1)
In [96]: grid_search=grid.fit(x_train,y_train)
          Fitting 10 folds for each of 30 candidates, totalling 300 fits
          C:\Users\Admin\anaconda3\lib\site-packages\sklearn\neighbors\_classification.py:207: DataConversionWarning: A column-vector y
          was passed when a 1d array was expected. Please change the shape of y to (n_samples,), for example using ravel().
            return self. fit(X, y)
          C:\Users\Admin\anaconda3\lib\site-packages\sklearn\neighbors\_classification.py:207: DataConversionWarning: A column-vector y
          was passed when a 1d array was expected. Please change the shape of y to (n_samples,), for example using ravel().
            return self._fit(X, y)
          C:\Users\Admin\anaconda3\lib\site-packages\sklearn\neighbors\_classification.py:207: DataConversionWarning: A column-vector y
          was passed when a 1d array was expected. Please change the shape of y to (n_samples,), for example using ravel().
            return self. fit(X, y)
          C:\Users\Admin\anaconda3\lib\site-packages\sklearn\neighbors\_classification.py:207: DataConversionWarning: A column-vector y
          was passed when a 1d array was expected. Please change the shape of y to (n_samples,), for example using ravel().
            return self. fit(X, y)
          C:\Users\Admin\anaconda3\lib\site-packages\sklearn\neighbors\_classification.py:207: DataConversionWarning: A column-vector y
          was passed when a 1d array was expected. Please change the shape of y to (n_samples,), for example using ravel().
            return self. fit(X, y)
          C:\Users\Admin\anaconda3\lib\site-packages\sklearn\neighbors\_classification.py:207: DataConversionWarning: A column-vector y
          was passed when a 1d array was expected. Please change the shape of y to (n_samples,), for example using ravel().
            naturn calf fit/Y v)
In [97]: grid_search.best_params_
Out[97]: {'n neighbors': 9}
In [98]: knn_classifier= KNeighborsClassifier(n_neighbors=9 )
In [99]: knn_classifier.fit(x_train,y_train)
          C:\Users\Admin\anaconda3\lib\site-packages\sklearn\neighbors\_classification.py:207: DataConversionWarning: A column-vector y w
          as passed when a 1d array was expected. Please change the shape of y to (n samples,), for example using ravel().
            return self._fit(X, y)
Out[99]: KNeighborsClassifier(n neighbors=9)
          In a Jupyter environment, please rerun this cell to show the HTML representation or trust the notebook.
          On GitHub, the HTML representation is unable to render, please try loading this page with nbviewer.org.
In [100]: y gscv knn=knn classifier.predict(x test)
In [101]: print(accuracy_score(y_gscv_knn,y_test))
          0.7515527950310559
In [102]: from sklearn.ensemble import RandomForestClassifier
In [103]: rf=RandomForestClassifier(random_state=42)
In [104]: rf.fit(x_train,y_train)
          C:\Users\Admin\AppData\Local\Temp\ipykernel_1892\1149647727.py:1: DataConversionWarning: A column-vector y was passed when a 1d
          array was expected. Please change the shape of y to (n_samples,), for example using ravel().
            rf.fit(x train,y train)
Out[104]: RandomForestClassifier(random_state=42)
          In a Jupyter environment, please rerun this cell to show the HTML representation or trust the notebook.
          On GitHub, the HTML representation is unable to render, please try loading this page with nbviewer.org.
In [105]: y_rf=rf.predict(x_test)
In [106]: print('Accuracy score of Random Forest without GridDSearchCV:',accuracy_score(y_rf,y_test))
          Accuracy score of Random Forest without GridDSearchCV: 0.7950310559006211
In [107]: # Now with the help of gridsearchCV we are going to find best fit parameters
```

```
In [108]: param_grid = {
                          'n_estimators': [200, 500],
'max_features': ['auto', 'sqrt', 'log2'],
                          'max_depth' : [4,5,6,7,8],
                          'criterion' :['gini', 'entropy']
                  CV rfc = GridSearchCV(estimator=rf, param grid=param grid, cv= 5)
                  CV_rfc.fit(x_train,y_train)
                  C:\Users\Admin\anaconda3\lib\site-packages\sklearn\model_selection\_validation.py:686: DataConversionWarning: A column-vector
                  y was passed when a 1d array was expected. Please change the shape of y to (n_samples,), for example using ravel().
                      estimator.fit(X_train, y_train, **fit_params)
                  C:\Users\Admin\anaconda3\lib\site-packages\sklearn\ensemble\_forest.py:427: FutureWarning: `max_features='auto'` has been dep
                  recated in 1.1 and will be removed in 1.3. To keep the past behaviour, explicitly set `max_features='sqrt'` or remove this pa
                  rameter as it is also the default value for RandomForestClassifiers and ExtraTreesClassifiers.
                      warn(
                  C:\Users\Admin\anaconda3\lib\site-packages\sklearn\model_selection\_validation.py:686: DataConversionWarning: A column-vector
                  y was passed when a 1d array was expected. Please change the shape of y to (n_samples,), for example using ravel().
                      estimator.fit(X_train, y_train, **fit_params)
                   \verb|C:\Users\Admin\anaconda3\lib\site-packages\sklearn\ensemble\_forest.py: 427: Future \verb|Warning:`max_features='auto'` has been departed to the packages of 
                  recated in 1.1 and will be removed in 1.3. To keep the past behaviour, explicitly set `max_features='sqrt'` or remove this pa
                  rameter as it is also the default value for RandomForestClassifiers and ExtraTreesClassifiers.
                  C:\Users\Admin\anaconda3\lib\site-packages\sklearn\model_selection\_validation.py:686: DataConversionWarning: A column-vector
                  y was passed when a 1d array was expected. Please change the shape of y to (n_samples,), for example using ravel().
                      estimator.fit(X_train, y_train, **fit_params)
                  C:\Users\Admin\anaconda3\lib\site-packages\sklearn\ensemble\_forest.py:427: FutureWarning: `max_features='auto'` has been dep
                  recated in 1.1 and will be removed in 1.3. To keep the past behaviour, explicitly set `max_features='sqrt'` or remove this pa
   In [ ]: CV_rfc.best_params_
   In [ ]: |rf_CV=RandomForestClassifier( n_estimators=500,
                         criterion='gini',
                         max_depth=4,
                         min_samples_split=2,
                         min_samples_leaf=1,
                         min_weight_fraction_leaf=0.0,
                         max_features='auto',
                         max_leaf_nodes=None,
                         min_impurity_decrease=0.0,
                         bootstrap=True,
                         oob score=False.
                         n_jobs=None,
                         random state=42,
                         verbose=0,
                         warm start=False.
                         class_weight=None,
                         ccp_alpha=0.0,
                         max_samples=None)
   In [ ]: rf_CV.fit(x_train,y_train)
   In [ ]: y_pred_rfcv=rf_CV.predict(x_test)
   In [ ]: print('Accuracy score of Random Forest with GridDSearchCV:',accuracy_score(y_pred_rfcv,y_test))
```

XGB boosting algorithm

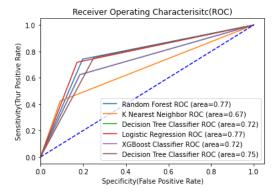
In a Jupyter environment, please rerun this cell to show the HTML representation or trust the notebook. On GitHub, the HTML representation is unable to render, please try loading this page with nbviewer.org.

```
In [111]: y_xgb=XGB.predict(x_test)
In [112]: y_xgb
Out[112]: array([0, 0, 0, 1, 0, 1, 0, 1, 0, 0, 0, 1, 1, 0, 1, 1, 1, 1, 0, 1, 0, 1,
                               0, 1, 0, 1, 1, 1, 1, 1, 1, 1, 1, 1, 0, 1, 1, 0, 1, 0, 0, 0, 0, 1,
                              0, 1, 1, 0, 0, 0, 0, 1, 1, 1, 0, 1, 1, 0, 1, 1, 0, 0, 1, 1, 1, 1,
                              0, 0, 0, 1, 1, 0, 0, 1, 0, 0, 1, 0, 1, 0, 1, 1, 0, 0, 0, 0, 1, 0,
                              0, 0, 1, 0, 0, 0, 0, 1, 0, 0, 1, 0, 0, 1, 0, 1, 1, 1, 1, 0, 0,
                              0, 0, 1, 0, 0, 1, 1, 1, 0, 1, 1, 0, 1, 0, 1, 0, 1, 1, 1, 1, 0, 0, 1,
                              1, 1, 1, 1, 1, 0, 0, 1, 0, 1, 0, 0, 0, 1, 0, 1, 0, 1, 0, 1, 1, 0,
                              0, 1, 1, 0, 1, 1, 1])
In [113]: |print('Accuracy_score of XGB Algorithm:',accuracy_score(y_xgb,y_test))
                  Accuracy_score of XGB Algorithm: 0.7515527950310559
In [114]: from sklearn.linear_model import LogisticRegression
In [115]: lr=LogisticRegression()
In [116]: lr.fit(x train,y train)
                  C:\Users\Admin\anaconda3\lib\site-packages\sklearn\utils\validation.py:1111: DataConversionWarning: A column-vector y was passe
                  d when a 1d array was expected. Please change the shape of y to (n_samples, ), for example using ravel().
                      y = column_or_1d(y, warn=True)
                  C:\Users\Admin\anaconda3\lib\site-packages\sklearn\linear_model\_logistic.py:444: ConvergenceWarning: lbfgs failed to converge
                  (status=1):
                  STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.
                  Increase the number of iterations (max_iter) or scale the data as shown in:
                         https://scikit-learn.org/stable/modules/preprocessing.html) (https://scikit-learn.org/stable/modules/preprocessing.html)
                  Please also refer to the documentation for alternative solver options:
                         https://scikit-learn.org/stable/modules/linear_model.html#logistic-regression (https://scikit-learn.org/stable/modules/linear_model.html#logistic-regression (https://scikit-regression) (https://scikit-regression) (https://scikit-
                  ar_model.html#logistic-regression)
                      n_iter_i = _check_optimize_result(
Out[116]: LogisticRegression()
                  In a Jupyter environment, please rerun this cell to show the HTML representation or trust the notebook.
                  On GitHub, the HTML representation is unable to render, please try loading this page with nbviewer.org.
In [117]: y_pred_lr=lr.predict(x_test)
In [118]: |print('Accuracy_score of Logistic Regression:',accuracy_score(y_pred_lr,y_test))
                  Accuracy_score of Logistic Regression: 0.7701863354037267
                  with GridSearchCV
In [119]: | grid={"C":np.logspace(-3,3,7), "penalty":["11","12"]}# l1 Lasso l2 ridge
                  logreg=LogisticRegression()
                  logreg_cv=GridSearchCV(logreg,grid,cv=10)
                  logreg_cv.fit(x_train,y_train)
                  C:\Users\Admin\anaconda3\lib\site-packages\sklearn\utils\validation.py:1111: DataConversionWarning: A column-vector y was pas
                  sed when a 1d array was expected. Please change the shape of y to (n samples, ), for example using ravel().
```

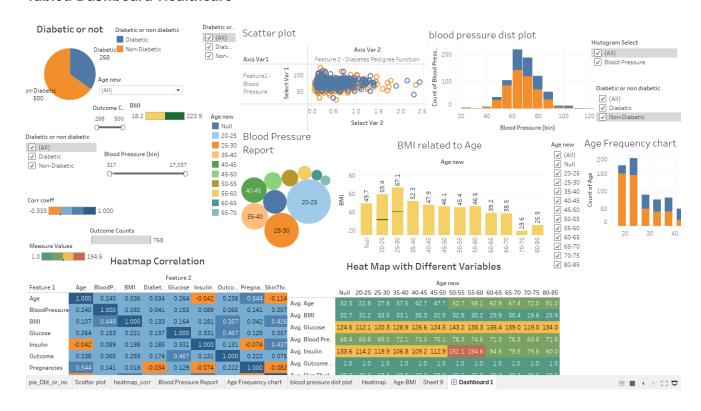
```
y = column_or_1d(y, warn=True)
C:\Users\Admin\anaconda3\lib\site-packages\sklearn\linear_model\_logistic.py:444: ConvergenceWarning: lbfgs failed to converg
e (status=1):
STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.
Increase the number of iterations (max_iter) or scale the data as shown in:
   https://scikit-learn.org/stable/modules/preprocessing.html (https://scikit-learn.org/stable/modules/preprocessing.html)
Please also refer to the documentation for alternative solver options:
   https://scikit-learn.org/stable/modules/linear_model.html#logistic-regression (https://scikit-learn.org/stable/modules/li
near model.html#logistic-regression)
 n_iter_i = _check_optimize_result(
C:\Users\Admin\anaconda3\lib\site-packages\sklearn\utils\validation.py:1111: DataConversionWarning: A column-vector y was pas
sed when a 1d array was expected. Please change the shape of y to (n_samples, ), for example using ravel().
 y = column or 1d(y, warn=True)
C:\Users\Admin\anaconda3\lib\site-packages\sklearn\linear model\ logistic.py:444: ConvergenceWarning: lbfgs failed to converg
e (status=1):
STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.
```

AUC curves for various Algorithms

```
In [127]: import sklearn.metrics as metrics
                   plt.figure()
                   models=[
                          { 'label':'Random Forest',
                               'model':RandomForestClassifier()
                          },
                          {
                                  'label':'K Nearest Neighbor',
                                   'model':KNeighborsClassifier(n_neighbors=2)
                          },
                          {
                                  'label': 'Decision Tree Classifier',
                                   'model':DecisionTreeClassifier()
                          },
                                  'label': 'Logistic Regression',
                                   'model':LogisticRegression()
                          },
                          {
                                  'label': 'XGBoost Classifier',
                                   'model':DecisionTreeClassifier()
                          },
                                  'label': 'Decision Tree Classifier',
                                   'model':xgb.XGBClassifier()
                          }
                   ]
                   for mod in models:
                          model=mod['model']
                          model.fit(x_train,y_train)
                          y_pred=model.predict(x_test)
                          fpr,tpr,thresholds=metrics.roc_curve(y_test,y_pred)
                          auc=metrics.roc_auc_score(y_test,model.predict(x_test))
                          plt.plot(fpr,tpr,label='%s ROC (area=%0.2f)'%(mod['label'],auc))
                   plt.plot([0,1],[0,1],'b--')
                   plt.xlim([0.0,1.0])
                   plt.xlim([0.0,1.05])
                   plt.xlabel('Specificity(False Positive Rate)')
                   plt.ylabel('Sensitivity(Trur Positive Rate)')
                   plt.title('Receiver Operating Characterisitc(ROC)')
                   plt.legend(loc='lower right')
                   plt.show()
                   C:\Users\Admin\AppData\Local\Temp\ipykernel_1892\1919318965.py:33: DataConversionWarning: A column-vector y was passed when a 1
                   d array was expected. Please change the shape of y to (n_samples,), for example using ravel().
                      model.fit(x train,y train)
                    \hbox{\tt C:\backslash Users\backslash Admin\backslash anaconda3\backslash lib\backslash site-packages\backslash sklearn\backslash neighbors\backslash classification.py:207: } DataConversionWarning: A column-vector y where the packages is the packages of the packag
                   as passed when a 1d array was expected. Please change the shape of y to (n_samples,), for example using ravel().
                       return self._fit(X, y)
                   C:\Users\Admin\anaconda3\lib\site-packages\sklearn\utils\validation.py:1111: DataConversionWarning: A column-vector y was passe
                   d when a 1d array was expected. Please change the shape of y to (n_samples, ), for example using ravel().
                       y = column_or_1d(y, warn=True)
                   C:\Users\Admin\anaconda3\lib\site-packages\sklearn\linear_model\_logistic.py:444: ConvergenceWarning: lbfgs failed to converge
                   (status=1):
                   STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.
                   Increase the number of iterations (max_iter) or scale the data as shown in:
                          https://scikit-learn.org/stable/modules/preprocessing.html (https://scikit-learn.org/stable/modules/preprocessing.html)
                   Please also refer to the documentation for alternative solver options:
                          https://scikit-learn.org/stable/modules/linear_model.html#logistic-regression (https://scikit-learn.org/stable/modules/linear_model.html
                   ar model.html#logistic-regression)
                       n_iter_i = _check_optimize_result(
```



Tableu Dashboard-Healthcare



End Of Project

Thank You.....Arun R Chougale

In []: