```
In [1]: import pandas as pd
         import numpy as np
         import matplotlib.pyplot as plt
         import seaborn as sns
In [2]: | df=pd.read_csv(r"C5_health care diabetes.csv")
Out[2]:
              Pregnancies Glucose BloodPressure SkinThickness Insulin BMI DiabetesPedigreeFunctio
            0
                        6
                               148
                                              72
                                                           35
                                                                      33.6
                                                                                             0.62
            1
                        1
                               85
                                              66
                                                           29
                                                                    0 26.6
                                                                                             0.35
            2
                        8
                               183
                                              64
                                                            0
                                                                    0
                                                                      23.3
                                                                                             0.67
            3
                        1
                               89
                                              66
                                                           23
                                                                   94
                                                                      28.1
                                                                                             0.16
            4
                               137
                                              40
                                                            35
                                                                  168 43.1
                                                                                             2.28
                                                                   ...
                                ...
                                              ...
                                                            ...
          763
                       10
                               101
                                              76
                                                            48
                                                                  180
                                                                      32.9
                                                                                             0.17
          764
                        2
                               122
                                              70
                                                           27
                                                                    0 36.8
                                                                                             0.34
          765
                        5
                               121
                                                                                             0.24
                                              72
                                                            23
                                                                  112 26.2
          766
                        1
                               126
                                              60
                                                             0
                                                                    0 30.1
                                                                                             0.34
          767
                        1
                               93
                                              70
                                                           31
                                                                    0 30.4
                                                                                             0.31
         768 rows × 9 columns
In [3]: df.info()
         <class 'pandas.core.frame.DataFrame'>
         RangeIndex: 768 entries, 0 to 767
         Data columns (total 9 columns):
          #
              Column
                                            Non-Null Count Dtype
               -----
          0
              Pregnancies
                                            768 non-null
                                                             int64
          1
              Glucose
                                            768 non-null
                                                             int64
          2
              BloodPressure
                                            768 non-null
                                                             int64
          3
              SkinThickness
                                            768 non-null
                                                             int64
          4
              Insulin
                                            768 non-null
                                                             int64
          5
                                            768 non-null
                                                             float64
          6
              DiabetesPedigreeFunction
                                           768 non-null
                                                             float64
          7
                                            768 non-null
              Age
                                                             int64
                                            768 non-null
              Outcome
                                                             int64
         dtypes: float64(2), int64(7)
         memory usage: 54.1 KB
```

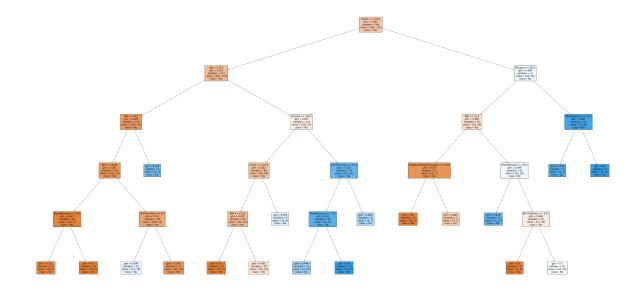
In [4]: df=df.dropna()

```
In [5]: df.isnull().sum()
 Out[5]: Pregnancies
                                         0
          Glucose
                                         0
          BloodPressure
                                         0
          SkinThickness
                                         0
          Insulin
                                         0
          BMI
                                         0
          DiabetesPedigreeFunction
                                         0
                                         0
          Outcome
                                         0
          dtype: int64
 In [6]: |df.describe()
 Out[6]:
                 Pregnancies
                                Glucose BloodPressure SkinThickness
                                                                        Insulin
                                                                                     BMI Diabetes
                  768.000000 768.000000
                                            768.000000
                                                         768.000000
                                                                    768.000000
                                                                               768.000000
           count
           mean
                     3.845052 120.894531
                                            69.105469
                                                          20.536458
                                                                     79.799479
                                                                                31.992578
             std
                    3.369578
                              31.972618
                                             19.355807
                                                          15.952218 115.244002
                                                                                 7.884160
                    0.000000
                               0.000000
                                             0.000000
                                                           0.000000
                                                                      0.000000
                                                                                 0.000000
            min
            25%
                    1.000000
                              99.000000
                                                           0.000000
                                                                      0.000000
                                                                                27.300000
                                            62.000000
            50%
                    3.000000 117.000000
                                            72.000000
                                                          23.000000
                                                                     30.500000
                                                                                32.000000
            75%
                    6.000000 140.250000
                                             80.000000
                                                          32.000000
                                                                    127.250000
                                                                                36.600000
                   17.000000 199.000000
                                            122.000000
                                                          99.000000 846.000000
                                                                                67.100000
            max
 In [7]: | df.columns
 Out[7]: Index(['Pregnancies', 'Glucose', 'BloodPressure', 'SkinThickness', 'Insulin',
                  'BMI', 'DiabetesPedigreeFunction', 'Age', 'Outcome'],
                 dtype='object')
 In [8]: |df['Outcome'].value counts()
 Out[8]: 0
                500
                268
          Name: Outcome, dtype: int64
 In [9]: df1=df[['Pregnancies', 'Glucose', 'BloodPressure', 'SkinThickness', 'Insulin',
                   'BMI', 'DiabetesPedigreeFunction', 'Age', 'Outcome']]
In [10]: | x=df1.drop('Outcome',axis=1)
          y=df1['Outcome']
In [11]: from sklearn.model_selection import train_test_split
          x_train,x_test,y_train,y_test=train_test_split(x,y,train_size=0.70)
```

```
In [12]: from sklearn.ensemble import RandomForestClassifier
         rfc=RandomForestClassifier()
         rfc.fit(x_train,y_train)
Out[12]: RandomForestClassifier()
In [13]: parameters={'max_depth':[1,2,3,4,5],
                      'min_samples_leaf':[5,10,15,20,25],
                      'n estimators':[10,20,30,40,50]}
In [14]: from sklearn.model_selection import GridSearchCV
         grid search=GridSearchCV(estimator=rfc,param grid=parameters,cv=2,scoring="accl
         grid_search.fit(x_train,y_train)
Out[14]: GridSearchCV(cv=2, estimator=RandomForestClassifier(),
                      param_grid={'max_depth': [1, 2, 3, 4, 5],
                                   'min_samples_leaf': [5, 10, 15, 20, 25],
                                   'n_estimators': [10, 20, 30, 40, 50]},
                      scoring='accuracy')
In [15]: grid_search.best_score_
Out[15]: 0.7690520446096654
In [16]: rfc_best=grid_search.best_estimator_
```

```
In [17]: from sklearn.tree import plot_tree
    plt.figure(figsize=(80,40))
    plot_tree(rfc_best.estimators_[5],feature_names=x.columns,class_names=['Yes','N
```

```
Out[17]: [Text(2590.7142857142853, 1993.2, 'Insulin <= 143.0\ngini = 0.459\nsamples =
               344\nvalue = [345, 192]\nclass = Yes'),
                 Text(1434.8571428571427, 1630.8000000000000, 'Age <= 31.5\ngini = 0.431\nsam
               ples = 273\nvalue = [291, 133]\nclass = Yes'),
                 Text(797.1428571428571, 1268.4, 'BMI <= 44.9\ngini = 0.294\nsamples = 150\nv
               alue = [183, 40]\nclass = Yes'),
                 Text(637.7142857142857, 906.0, 'BMI <= 26.35\ngini = 0.261\nsamples = 145\nv
               alue = [181, 33]\nclass = Yes'),
                 Text(318.85714285714283, 543.59999999999, 'BloodPressure <= 54.5\ngini =
               0.027\nsamples = 46\nvalue = [72, 1]\nclass = Yes'),
                 Text(159.42857142857142, 181.19999999999982, 'gini = 0.111\nsamples = 13\nva
               lue = [16, 1]\nclass = Yes'),
                 Text(478.2857142857142, 181.1999999999982, 'gini = 0.0\nsamples = 33\nvalue
               = [56, 0]\nclass = Yes'),
                 Text(956.5714285714284, 543.59999999999, 'SkinThickness <= 5.0\ngini = 0.3
               51\nsamples = 99\nvalue = [109, 32]\nclass = Yes'),
                 Text(797.1428571428571, 181.19999999999982, 'gini = 0.498\nsamples = 22\nval
               ue = [14, 16] \setminus nclass = No'),
                 Text(1116.0, 181.199999999999, 'gini = 0.247\nsamples = 77\nvalue = [95, 1
               6]\nclass = Yes'),
                 Text(956.5714285714284, 906.0, 'gini = 0.346\nsamples = 5\nvalue = [2, 7]\nc
               lass = No'),
                 Text(2072.5714285714284, 1268.4, 'Glucose <= 146.5 \neq 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.497 = 0.4
               123\nvalue = [108, 93]\nclass = Yes'),
                 Text(1753.7142857142856, 906.0, 'Insulin <= 107.5\ngini = 0.444\nsamples = 8
               9\nvalue = [98, 49]\nclass = Yes'),
                 Text(1594.2857142857142, 543.599999999999, 'BMI <= 27.25\ngini = 0.432\nsam
               ples = 80\nvalue = [93, 43]\nclass = Yes'),
                 Text(1434.8571428571427, 181.1999999999982, 'gini = 0.0\nsamples = 21\nvalu
               e = [37, 0]\nclass = Yes'),
                 Text(1753.7142857142856, 181.1999999999999, 'gini = 0.491\nsamples = 59\nva
               lue = [56, 43] \setminus class = Yes'),
                 Text(1913.1428571428569, 543.599999999999, 'gini = 0.496\nsamples = 9\nvalu
               e = [5, 6] \setminus nclass = No'),
                Text(2391.428571428571, 906.0, 'SkinThickness <= 30.5\ngini = 0.302\nsamples
               = 34\nvalue = [10, 44]\nclass = No'),
                 Text(2232.0, 543.599999999999, 'BloodPressure <= 77.0\ngini = 0.24\nsamples
               = 26\nvalue = [6, 37]\nclass = No'),
                Text(2072.5714285714284, 181.199999999999982, 'gini = 0.444\nsamples = 13\nva
               lue = [6, 12] \setminus class = No'),
                 Text(2391.428571428571, 181.19999999999982, 'gini = 0.0\nsamples = 13\nvalue
               = [0, 25] \setminus nclass = No'),
                 Text(2550.8571428571427, 543.599999999999, 'gini = 0.463\nsamples = 8\nvalu
               e = [4, 7] \setminus nclass = No'),
                 Text(3746.5714285714284, 1630.8000000000002, 'Glucose <= 165.5\ngini = 0.499
               \nsamples = 71 \nvalue = [54, 59] \nclass = No'),
                 Text(3347.99999999995, 1268.4, 'BMI <= 31.4\ngini = 0.488\nsamples = 56\nv
               alue = [53, 39]\nclass = Yes'),
                 Text(3029.142857142857, 906.0, 'DiabetesPedigreeFunction <= 0.746\ngini = 0.
               227\nsamples = 13\nvalue = [20, 3]\nclass = Yes'),
                Text(2869.7142857142853, 543.599999999999, 'gini = 0.0\nsamples = 8\nvalue
               = [15, 0]\nclass = Yes'),
                 Text(3188.5714285714284, 543.599999999999, 'gini = 0.469\nsamples = 5\nvalu
               e = [5, 3]\nclass = Yes'),
                 Text(3666.8571428571427, 906.0, 'BloodPressure <= 65.0\ngini = 0.499\nsample
               s = 43 \text{ nvalue} = [33, 36] \text{ nclass} = \text{No'}),
                 Text(3507.428571428571, 543.599999999999, 'gini = 0.278\nsamples = 8\nvalue
```



In [ ]: