```
In [2]:
         import pandas as pd
         data=pd.read_csv("C:\\Users\\pallavi pegdyal\\Downloads\\diabetes.csv")
In [3]:
In [4]:
        data.head()
Out[4]:
             Pregnancies
                         Glucose BloodPressure SkinThickness Insulin BMI DiabetesPedigreeFunct
          0
                      6
                             148
                                            72
                                                           35
                                                                     33.6
                                                                                             0.
          1
                       1
                              85
                                            66
                                                           29
                                                                   0
                                                                      26.6
                                                                                             0.
          2
                      8
                             183
                                            64
                                                           0
                                                                      23.3
                                                                                             0.
          3
                       1
                              89
                                            66
                                                           23
                                                                  94 28.1
                                                                                             0.
                      0
                                                                 168 43.1
                                                                                             2.
                             137
                                            40
                                                           35
In [5]:
         data.tail()
Out[5]:
               Pregnancies Glucose BloodPressure SkinThickness Insulin BMI DiabetesPedigreeFu
          763
                       10
                               101
                                                                       32.9
                                              76
                                                            48
                                                                   180
          764
                        2
                               122
                                              70
                                                            27
                                                                     0 36.8
          765
                        5
                                                                   112 26.2
                               121
                                              72
                                                            23
          766
                        1
                               126
                                              60
                                                             0
                                                                     0
                                                                       30.1
                                93
                                              70
                                                             31
                                                                       30.4
In [6]:
         data.shape
Out[6]: (768, 9)
         print("number of rows",data.shape[0])
         print("number of columns",data.shape[1])
         number of rows 768
         number of columns 9
         #checkin null values in dataset
In [8]: data.isnull().sum()
Out[8]: Pregnancies
                                         0
         Glucose
                                         0
         BloodPressure
                                         0
         SkinThickness
                                         0
         Insulin
                                         0
                                         0
         BMI
         DiabetesPedigreeFunction
                                         0
                                         0
         Age
         Outcome
                                         0
         dtype: int64
```

```
In [9]:
          #checking overall statistics
          data.describe()
 Out[9]:
                 Pregnancies
                                Glucose BloodPressure SkinThickness
                                                                        Insulin
                                                                                     BMI Diab
           count
                   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
             std
                    3.369578
                              31.972618
                                             19.355807
                                                          15.952218 115.244002
                                                                                 7.884160
             min
                    0.000000
                               0.000000
                                             0.000000
                                                           0.000000
                                                                      0.000000
                                                                                 0.000000
            25%
                    1.000000
                              99.000000
                                             62.000000
                                                           0.000000
                                                                      0.000000
                                                                                27.300000
            50%
                                                          23.000000
                    3.000000 117.000000
                                            72.000000
                                                                     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 [10]:
          import numpy as np
          data_copy=data.copy(deep=True)
In [11]:
          data.columns
In [12]:
Out[12]: Index(['Pregnancies', 'Glucose', 'BloodPressure', 'SkinThickness', 'Insuli
                  'BMI', 'DiabetesPedigreeFunction', 'Age', 'Outcome'],
                 dtype='object')
          data_copy[[ 'Glucose', 'BloodPressure', 'SkinThickness', 'Insulin',
In [13]:
                  'BMI']]=data_copy[[ 'Glucose', 'BloodPressure', 'SkinThickness', 'In
                  'BMI']].replace(0,np.nan)
In [14]: |data_copy.isnull().sum()
Out[14]: Pregnancies
                                           0
          Glucose
                                            5
          BloodPressure
                                          35
          SkinThickness
                                         227
          Insulin
                                         374
          BMI
                                          11
          DiabetesPedigreeFunction
                                           0
                                            0
                                            0
          Outcome
          dtype: int64
```

```
In [15]: data['Glucose'] = data['Glucose'].replace(0, data['Glucose'].mean())
    data['SkinThickness'] = data['SkinThickness'].replace(0, data['SkinThickness'].mean())
    data['Insulin'] = data['Insulin'].replace(0, data['Insulin'].mean())
    data['BMI'] = data['BMI'].replace(0, data['BMI'].mean())
    data['Pregnancies'] = data['Pregnancies'].replace(0, data['Pregnancies'].me
```

```
In [16]: X=data.drop('Outcome',axis=1)
Y=data['Outcome']
```

In [17]: #splitting the datasset into training set and testing set
from sklearn.model_selection import train_test_split
X_train,X_test,Y_train,Y_test=train_test_split(X,Y,test_size=0.20,random_st

In [40]: print(X.shape,X_train.shape,X_test.shape)
print(X_train,X_test,Y_train,Y_test)

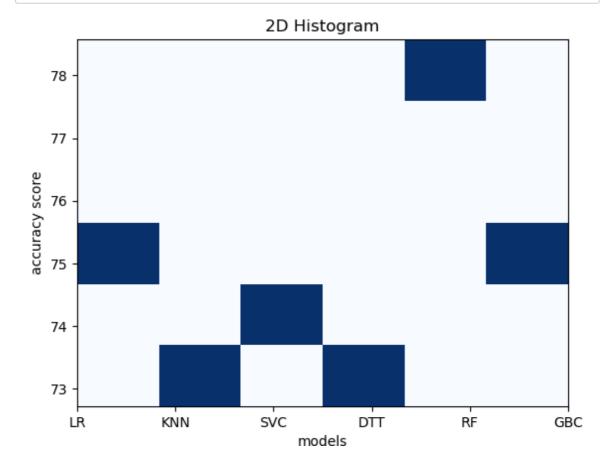
```
(768, 8) (614, 8) (154, 8)
                   Glucose BloodPressure SkinThickness
     Pregnancies
                                                                  Insulin
                                                                79.799479
60
         2.000000
                       84.0
                                           0
                                                   20.536458
618
         9.000000
                      112.0
                                          82
                                                   24.000000
                                                                79.799479
346
                                          46
         1.000000
                      139.0
                                                   19.000000
                                                                83.000000
294
                                                                79.799479
         3.845052
                      161.0
                                          50
                                                   20.536458
231
         6.000000
                      134.0
                                          80
                                                   37.000000
                                                               370.000000
. .
              . . .
                        . . .
                                         . . .
71
         5.000000
                                                               140.000000
                      139.0
                                          64
                                                   35.000000
                      96.0
                                                   20.536458
106
        1.000000
                                         122
                                                                79.799479
270
       10.000000
                      101.0
                                          86
                                                   37.000000
                                                                79.799479
        3.845052
435
                      141.0
                                                                79.799479
                                           0
                                                   20.536458
                                          96
                                                                79.799479
102
         3.845052
                      125.0
                                                   20.536458
                 DiabetesPedigreeFunction
                                              Age
60
     31.992578
                                      0.304
                                               21
618
     28.200000
                                       1.282
                                               50
346
     28.700000
                                      0.654
                                               22
294
     21.900000
                                      0.254
                                               65
231
     46.200000
                                      0.238
                                               46
                                         . . .
. .
            . . .
                                              . . .
71
                                      0.411
     28.600000
                                               26
106
     22.400000
                                      0.207
                                               27
270
     45.600000
                                      1.136
                                               38
435
     42.400000
                                      0.205
                                               29
102
     22.500000
                                      0.262
                                               21
[614 rows x 8 columns]
                              Pregnancies Glucose BloodPressure SkinThick
          Insulin
                     BMI
ness
668
              6.0
                       98.0
                                          58
                                                   33.000000
                                                               190.000000
                                                                            34.0
324
              2.0
                      112.0
                                          75
                                                   32.000000
                                                                79.799479
                                                                            35.7
624
              2.0
                      108.0
                                          64
                                                   20.536458
                                                                79.799479
                                                                            30.8
690
              8.0
                                          80
                                                   20.536458
                                                                79.799479
                                                                            24.6
                      107.0
473
              7.0
                      136.0
                                          90
                                                   20.536458
                                                                79.799479
                                                                            29.9
. .
              . . .
                        . . .
                                         . . .
355
              9.0
                      165.0
                                          88
                                                   20.536458
                                                                79.799479
                                                                            30.4
534
              1.0
                       77.0
                                          56
                                                   30.000000
                                                                56.000000
                                                                            33.3
344
              8.0
                       95.0
                                          72
                                                                79.799479
                                                   20.536458
                                                                            36.8
296
              2.0
                      146.0
                                          70
                                                   38.000000
                                                               360.000000
                                                                            28.0
462
              8.0
                       74.0
                                          70
                                                  40.000000
                                                                49.000000
                                                                           35.3
     DiabetesPedigreeFunction
                                  Age
668
                          0.430
                                   43
324
                          0.148
                                   21
624
                          0.158
                                   21
690
                          0.856
                                   34
473
                          0.210
                                   50
. .
                             . . .
                                  . . .
355
                          0.302
                                   49
                          1.251
534
                                   24
344
                          0.485
                                   57
296
                          0.337
                                   29
462
                          0.705
                                   39
[154 rows x 8 columns] 60
       1
618
346
       0
294
       0
```

```
71
      0
106
      0
270
       1
435
       1
102
Name: Outcome, Length: 614, dtype: int64 668
324
624
       0
690
       0
473
       0
355
      1
534
      0
344
      0
296
       1
462
Name: Outcome, Length: 154, dtype: int64
```

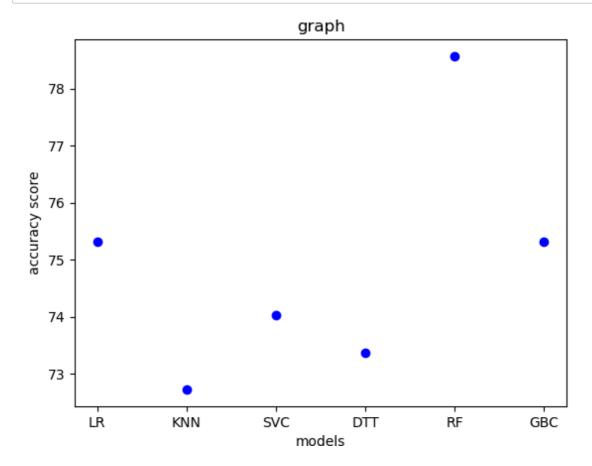
```
In [19]: from sklearn.preprocessing import StandardScaler
         from sklearn.linear_model import LogisticRegression
         from sklearn.neighbors import KNeighborsClassifier
         from sklearn.svm import SVC
         from sklearn.tree import DecisionTreeClassifier
         from sklearn.ensemble import RandomForestClassifier
         from sklearn.ensemble import GradientBoostingClassifier
         from sklearn.pipeline import Pipeline
In [20]: pipeline_lr=Pipeline([('scalar1',StandardScaler()),
                                ('lr_classifier',LogisticRegression())])
         pipeline_knn=Pipeline([('scalar2',StandardScaler()),
                                ('knn_classifier', KNeighborsClassifier())])
         pipeline_svc=Pipeline([('scaler3',StandardScaler()),
                                 ('svc_classifier',SVC())])
         pipeline_dt=Pipeline([('dt_classifier',DecisionTreeClassifier())])
         pipeline_rf=Pipeline([('rf_classifier',RandomForestClassifier(max_depth=3))
         pipeline_gbc=Pipeline([('gbc_classifier',GradientBoostingClassifier())])
         #pipeline_vt=Pipeline([('voting_classifier', VotingClassifier(estimators=[(p
In [21]: pipelines=[pipeline_lr,
                    pipeline_knn,
                    pipeline_svc,
                    pipeline dt,
                    pipeline rf,
                    pipeline_gbc]
```

```
In [22]: pipelines
Out[22]: [Pipeline(steps=[('scalar1', StandardScaler()),
                           ('lr_classifier', LogisticRegression())]),
          Pipeline(steps=[('scalar2', StandardScaler()),
                           ('knn_classifier', KNeighborsClassifier())]),
          Pipeline(steps=[('scaler3', StandardScaler()), ('svc_classifier', SVC
         ())]),
          Pipeline(steps=[('dt_classifier', DecisionTreeClassifier())]),
          Pipeline(steps=[('rf_classifier', RandomForestClassifier(max_depth=3))]),
          Pipeline(steps=[('gbc_classifier', GradientBoostingClassifier())])]
In [23]:
         for pipe in pipelines:
             pipe.fit(X_train,Y_train)
In [24]: pipe dict={0:'LR',
                    1: 'KNN',
                    2:'SVC',
                    3:'DT',
                    4:'RF',
                    5:'GBC'}
In [25]: pipe_dict
Out[25]: {0: 'LR', 1: 'KNN', 2: 'SVC', 3: 'DT', 4: 'RF', 5: 'GBC'}
In [26]: for i,model in enumerate(pipelines):
             print("{}Test Accuracy:{}".format(pipe_dict[i],model.score(X_test,Y_test))
         LRTest Accuracy:75.32467532467533
         KNNTest Accuracy:72.727272727273
         SVCTest Accuracy:74.02597402597402
         DTTest Accuracy:73.37662337662337
         RFTest Accuracy:77.272727272727
         GBCTest Accuracy:75.32467532467533
In [27]: | import matplotlib.pyplot as plt
```

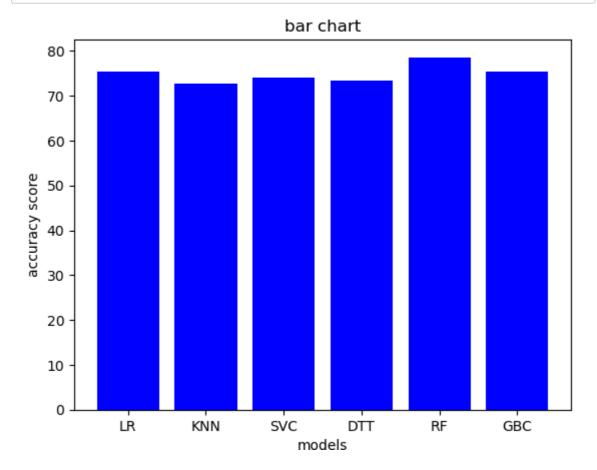
```
In [28]: x=np.array([0,1,2,3,4,5])
    y=np.array([75.3246,72.7272,74.0259,73.3766,78.5714,75.3246])
    plt.hist2d(x,y, bins=(6,6), cmap=plt.cm.Blues)
    plt.xticks([0,1,2,3,4,5],['LR','KNN','SVC','DTT','RF','GBC'])
    plt.xlabel('models')
    plt.ylabel('accuracy score')
    plt.title('2D Histogram')
    plt.show()
```



```
In [29]: x=np.array([0,1,2,3,4,5])
    y=np.array([75.3246,72.7272,74.0259,73.3766,78.5714,75.3246])
    plt.plot(x,y,'o',color='blue')
    plt.xticks([0,1,2,3,4,5],['LR','KNN','SVC','DTT','RF','GBC'])
    plt.xlabel('models')
    plt.ylabel('accuracy score')
    plt.title('graph')
    plt.show()
```



```
In [30]: x=['LR','KNN','SVC','DTT','RF','GBC']
    y=[75.3246,72.7272,74.0259,73.3766,78.5714,75.3246]
    plt.bar(x,y,color='blue')
    plt.xlabel('models')
    plt.ylabel('accuracy score')
    plt.title('bar chart')
    plt.show()
```



```
In [31]: from sklearn.ensemble import RandomForestClassifier
In [32]: X=data.drop('Outcome',axis=1)
Y=data['Outcome']
In [33]: rf=RandomForestClassifier(max_depth=3)
In [34]: rf.fit(X,Y)
```

Out[34]: RandomForestClassifier(max_depth=3)

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 [35]:
         feature_importances=rf.feature_importances_
         for i,feature in enumerate(X.columns):
             print(f"{feature}:{feature_importances[i]*100}")
         Pregnancies:10.041377590538454
         Glucose:35.730179879623385
         BloodPressure: 1.6611141066173896
         SkinThickness:4.99578294452019
         Insulin:7.302576836746192
         BMI:16.297431208069245
         DiabetesPedigreeFunction:8.00721601873113
         Age:15.964321415154012
In [37]: #prediction on new data
         new data=pd.DataFrame({
             'Pregnancies':6,
              'Glucose':148.0,
              'BloodPressure':72.0,
             'SkinThickness':35.0,
             'Insulin':0.0,
              'BMI':33.6,
              'DiabetesPedigreeFunction':0.627,
              'Age':50
         },index=[0])
In [38]: p=rf.predict(new_data)
In [39]: if p[0] == 0:
             print('non-diabetic')
         else:
             print('diabetic')
         diabetic
In [48]:
         #save the model
         import joblib
In [49]: |joblib.dump(rf, 'diabetes_model')
Out[49]: ['diabetes_model']
In [50]: model=joblib.load('diabetes_model')
In [51]: model.predict(new_data)
Out[51]: array([1], dtype=int64)
In [52]
         #creating gui
         from tkinter import *
         import joblib
```

```
In [ ]:
          from tkinter import *
          import joblib
          import numpy as np
          from sklearn import *
          def show_entry_fields():
               p1=float(e1.get())
               p2=float(e2.get())
               p3=float(e3.get())
               p4=float(e4.get())
               p5=float(e5.get())
               p6=float(e6.get())
               p7=float(e7.get())
               p8=float(e8.get())
               model=joblib.load('diabetes model')
               result=model.predict([[p1,p2,p3,p4,p5,p6,p7,p8]])
               if result==0:
                   Label(master, text="Non-Diabetic").grid(row=31)
               else:
                   Label(master, text="Diabetic").grid(row=31)
          master = Tk()
          master.title("Diabetes Prediction Using Machine Learning")
          label=Label(master,text="Diabetes Prediction using Machine Learning"
                                  ,bg="yellow",fg="red"). \
                                      grid(row=0,columnspan=2)
          Label(master, text="Pregnancies").grid(row=1)
          Label(master, text="Glucose").grid(row=2)
          Label(master, text="Enter value of BloodPressure").grid(row=3)
          Label(master, text="Enter value of SkinThickness").grid(row=4)
          Label(master, text="Enter value of Insulin").grid(row=5)
          Label(master, text="Enter value of BMI").grid(row=6)
          Label(master, text="Enter value of DiabetesPedigreeFunction").grid(row=7)
          Label(master, text="Enter value of Age").grid(row=8)
          e1=Entry(master)
          e2=Entry(master)
          e3=Entry(master)
          e4=Entry(master)
          e5=Entry(master)
          e6=Entry(master)
          e7=Entry(master)
          e8=Entry(master)
          e1.grid(row=1,column=1)
          e2.grid(row=2,column=1)
          e3.grid(row=3,column=1)
          e4.grid(row=4,column=1)
          e5.grid(row=5,column=1)
          e6.grid(row=6,column=1)
          e7.grid(row=7,column=1)
          e8.grid(row=8,column=1)
          Button(master, text='predict',command=show_entry_fields).grid()
```

	<pre>mainloop()</pre>
In []:	
[].	
In []:	
In []:	
T . F .	
In []:	
T- [].	
In []:	
In []:	
Tn [].	
In []:	