```
In [1]: import numpy as np
         import pandas as pd
         import matplotlib.pyplot as plt
In [2]: data=pd.read_csv(r"C:\Users\ASUS\Downloads\archive (7)\Heart Attack.csv")
In [3]: data.head()
Out[3]:
            age
                 gender
                          impluse pressurehight pressurelow
                                                                glucose
                                                                           kcm
                                                                               troponin
                                                                                             class
         0
              64
                                66
                                              160
                                                                   160.0
                                                                           1.80
                                                                                    0.012 negative
                       1
          1
              21
                                94
                                               98
                                                                   296.0
                                                                           6.75
                                                                                    1.060
                                                                                           positive
         2
              55
                       1
                                64
                                              160
                                                            77
                                                                   270.0
                                                                           1.99
                                                                                    0.003
                                                                                          negative
                                                                   270.0
         3
              64
                                70
                                              120
                                                            55
                                                                          13.87
                                                                                    0.122
                                                                                           positive
                                                                   300.0
              55
                       1
                                64
                                              112
                                                            65
                                                                           1.08
                                                                                    0.003 negative
In [4]:
        data.tail()
Out[4]:
                     gender impluse
                                       pressurehight pressurelow
                                                                   glucose
                                                                                                class
                age
                                                                              kcm
                                                                                    troponin
          1314
                           1
                                   94
                                                 122
                                                                67
                 44
                                                                      204.0
                                                                              1.63
                                                                                       0.006 negative
          1315
                                                 125
                                                                      149.0
                                                                              1.33
                 66
                           1
                                   84
                                                                55
                                                                                       0.172
                                                                                               positive
                                                 168
                                                               104
                                                                       96.0
                                                                                       4.250
          1316
                           1
                                   85
                                                                              1.24
                                                                                               positive
          1317
                 54
                                   58
                                                 117
                                                                68
                                                                      443.0
                                                                              5.80
                                                                                       0.359
                                                                                               positive
                                                 157
          1318
                 51
                           1
                                   94
                                                                79
                                                                      134.0 50.89
                                                                                       1.770
                                                                                               positive
         data.sample(15)
In [5]:
Out[5]:
                     gender impluse
                                      pressurehight pressurelow
                                                                    glucose
                                                                               kcm troponin
                                                                                                 class
                age
            86
                 40
                                   76
                                                 157
                                                                      193.0
                                                                               4.66
                                                                                        0.003
                                                                                              negative
                 75
                           0
                                   75
                                                 134
           993
                                                                85
                                                                      201.0
                                                                               1.24
                                                                                        0.007
                                                                                               negative
           231
                 50
                           1
                                   75
                                                 142
                                                                75
                                                                      122 0
                                                                               6 27
                                                                                        0.004
                                                                                               negative
           925
                 67
                                   58
                                                  93
                                                                78
                                                                      108.0
                                                                               3.13
                                                                                        0.009
                                                                                               negative
          1244
                 37
                           1
                                   88
                                                 119
                                                                66
                                                                      118.0
                                                                               5.78
                                                                                        0.006
                                                                                               negative
            61
                 90
                           0
                                   58
                                                 120
                                                                69
                                                                      191.0
                                                                               5.22
                                                                                        0.015
                                                                                                positive
           447
                 50
                           1
                                   73
                                                 135
                                                                79
                                                                      238.0
                                                                               1.87
                                                                                        0.005
                                                                                               negative
           761
                 55
                           0
                                   68
                                                 116
                                                                74
                                                                      143.0
                                                                               1.34
                                                                                        0.094
                                                                                                positive
          1142
                 63
                           1
                                   94
                                                 105
                                                                81
                                                                      168.0
                                                                               1.58
                                                                                        0.005
                                                                                               negative
           627
                 45
                                   96
                                                  97
                                                                55
                                                                      144.0
                                                                               2.87
                                                                                        1.480
                                                                                                positive
            93
                 60
                           0
                                   60
                                                 130
                                                                56
                                                                      294.0
                                                                               2.13
                                                                                        0.103
                                                                                                positive
           560
                 60
                           1
                                   60
                                                 179
                                                                83
                                                                      347.0
                                                                               1.25
                                                                                        0.043
                                                                                                positive
            71
                 58
                           0
                                   91
                                                 120
                                                                80
                                                                      177.0
                                                                              18.15
                                                                                        0.005
                                                                                                positive
          1067
                 52
                           0
                                   58
                                                 120
                                                                69
                                                                       97.0
                                                                               5.17
                                                                                        0.083
                                                                                                positive
             7
                 63
                                   60
                                                 214
                                                                82
                                                                       87.0 300.00
                                                                                        2.370
                           1
                                                                                                positive
In [6]: data.info()
        <class 'pandas.core.frame.DataFrame'>
        RangeIndex: 1319 entries, 0 to 1318
        Data columns (total 9 columns):
         #
             Column
                               Non-Null Count Dtype
        - - -
         0
                                1319 non-null
                                                   int64
              age
         1
              gender
                                1319 non-null
                                                   int64
                                1319 non-null
              impluse
                                                   int64
              pressurehight 1319 non-null
         3
                                                   int64
              pressurelow
                                1319 non-null
                                                   int64
         5
                                1319 non-null
              glucose
                                                   float64
              kcm
                                1319 non-null
                                                   float64
              troponin
                                1319 non-null
                                                   float64
         7
```

8

class

memory usage: 92.9+ KB

1319 non-null

dtypes: float64(3), int64(5), object(1)

object

```
kcm
                                  gender
                                             impluse pressurehight pressurelow
                                                                                    glucose
                                                                                                            troponin
                        age
          count 1319.000000
                             1319.000000
                                         1319.000000
                                                                                 1319.000000
                                                                                             1319.000000
                                                                                                         1319.000000
                                                        1319.000000
                                                                    1319.000000
                   56.191812
                                0.659591
                                            78.336619
                                                         127.170584
                                                                      72.269143
                                                                                  146.634344
                                                                                               15.274306
                                                                                                            0.360942
          mean
            std
                   13.647315
                                0.474027
                                            51.630270
                                                          26.122720
                                                                      14.033924
                                                                                   74.923045
                                                                                               46.327083
                                                                                                            1.154568
                   14.000000
                                0.000000
                                            20.000000
                                                          42.000000
                                                                      38.000000
                                                                                  35.000000
                                                                                                0.321000
                                                                                                            0.001000
            min
           25%
                   47.000000
                                0.000000
                                            64.000000
                                                         110.000000
                                                                      62.000000
                                                                                   98.000000
                                                                                                1.655000
                                                                                                            0.006000
           50%
                   58.000000
                                1.000000
                                            74.000000
                                                         124.000000
                                                                      72.000000
                                                                                  116.000000
                                                                                                2.850000
                                                                                                            0.014000
           75%
                   65.000000
                                1.000000
                                            85.000000
                                                         143.000000
                                                                      81.000000
                                                                                  169.500000
                                                                                                5.805000
                                                                                                            0.085500
                  103.000000
           max
                                1.000000
                                          1111.000000
                                                         223.000000
                                                                     154.000000
                                                                                  541.000000
                                                                                              300.000000
                                                                                                            10.300000
 In [8]: data.shape
 Out[8]:
          (1319, 9)
 In [9]: data.size
 Out[9]: 11871
In [10]: data.isnull().sum()
Out[10]:
                             0
          age
          gender
                             0
          impluse
                             0
          pressurehight
                             0
          pressurelow
                             0
                             0
          glucose
          kcm
                             0
          troponin
                             0
          class
                             0
          dtype: int64
In [11]: data.age.value_counts()
          age
                  106
          60
          70
                   73
          50
                   68
          63
                   64
          65
                   62
          88
                    1
          100
                    1
          14
                    1
          91
                    1
          84
                    1
          Name: count, Length: 75, dtype: int64
In [12]: data['class'].value_counts()
Out[12]: class
          positive
                       810
          negative
                       509
          Name: count, dtype: int64
In [13]: negative = data[data['class'] == "negative"]
          positive = data[data['class'] == "positive"]
In [14]: negative.shape
Out[14]: (509, 9)
In [15]: positive.shape
Out[15]: (810, 9)
In [16]: positive_sample = positive.sample(n=509)
          positive_sample.shape
Out[16]: (509, 9)
In [17]: newdata = pd.concat([positive sample , negative], axis=0)
```

In [7]: data.describe()

In [18]: newdata.sample(15)

```
1037
                  55
                                    74
                                                 150
                                                               90
                                                                      117.0
                                                                               7.610
                                                                                        0.104
                                                                                                positive
            860
                  54
                           1
                                    72
                                                 154
                                                               84
                                                                      127.0
                                                                               2.970
                                                                                        0.007
                                                                                               negative
            617
                  60
                                    80
                                                 135
                                                               75
                                                                       94.0
                                                                             147.400
                                                                                        3.850
                                                                                                positive
                           1
                                                 170
                                                                               2.610
            223
                  63
                                   119
                                                              107
                                                                      129 0
                                                                                        0.005
                                                                                               negative
            602
                  45
                           0
                                    80
                                                 117
                                                               83
                                                                      143.0
                                                                               2.490
                                                                                        0.003
                                                                                               negative
            760
                  69
                           0
                                    73
                                                 135
                                                               81
                                                                       69.0
                                                                               4.950
                                                                                         0.007
                                                                                               negative
            419
                  45
                           1
                                    90
                                                 110
                                                               65
                                                                       83.0
                                                                               2.420
                                                                                        0.096
                                                                                                positive
                           0
                                                 128
                                                                               2.330
            254
                  67
                                    69
                                                               70
                                                                      382 0
                                                                                        0.007
                                                                                               negative
           1021
                  61
                           0
                                    93
                                                 120
                                                               71
                                                                      121.0
                                                                              79.620
                                                                                        0.007
                                                                                                positive
            888
                  71
                           1
                                    59
                                                 107
                                                               64
                                                                       97.0
                                                                               1.970
                                                                                         1.450
                                                                                                positive
                           0
            983
                  50
                                    81
                                                 124
                                                               75
                                                                      114.0
                                                                               0.321
                                                                                        0.003
                                                                                               negative
                           1
                                                 130
                                                               72
                                                                       96.0
                                                                               2 800
            363
                  55
                                    60
                                                                                        0 117
                                                                                                positive
           1036
                  72
                           0
                                    75
                                                 160
                                                               70
                                                                      130.0
                                                                               8.540
                                                                                        0.015
                                                                                                positive
            208
                  49
                           0
                                    67
                                                 120
                                                               55
                                                                      100.0
                                                                               0.676
                                                                                         0.005
                                                                                                positive
In [19]:
          newdata['class'].value_counts()
Out[19]:
           class
                         509
           positive
                         509
           negative
           Name: count, dtype: int64
In [20]:
          newdata.shape
Out[20]:
           (1018, 9)
In [21]: X=newdata.drop("class" , axis="columns")
           X.head()
Out[21]:
                     gender impluse pressurehight pressurelow
                age
                                                                  glucose
                                                                             kcm troponin
            19
                 45
                                                                      89.0
                                                                             1.60
                                                                                     0.020
                           1
                                                109
            23
                 30
                           1
                                  63
                                                110
                                                              68
                                                                     107.0
                                                                            50.46
                                                                                     0.003
                                                                     106.0
                 78
                           1
                                   62
                                                157
                                                              66
                                                                                     0.024
           189
                                                                             3.77
                                                                                     0.017
           149
                 69
                                   82
                                                 86
                                                              70
                                                                      87.0
                                                                             4.37
                 42
                           0
                                   73
                                                162
                                                              99
                                                                     109.0 15.83
                                                                                     0.100
           571
In [22]:
          X.shape
           (1018, 8)
In [23]:
           Y=newdata['class'].values
In [24]:
Out[24]: array(['positive', 'positive', 'positive', ..., 'negative', 'negative',
                    'negative'], dtype=object)
In [25]:
          Y.shape
Out[25]:
           (1018,)
In [26]:
          print(Y)
          ['positive' 'positive' 'positive' ... 'negative' 'negative' 'negative']
In [27]:
          from sklearn.preprocessing import LabelEncoder
           le=LabelEncoder()
           Y=le.fit transform(Y)
In [28]:
Out[28]: array([1, 1, 1, ..., 0, 0, 0])
In [29]: from sklearn.model selection import train test split
```

X_train , X_test , y_train , y_test = train_test_split(X , Y , test_size= 0.2)

pressurehight pressurelow

152

glucose

99 0

87

kcm troponin

1 210

class

0.004 negative

Out[18]:

age

82

596

gender impluse

88

0

```
In [30]: len(X train)
Out[30]: 814
In [31]: len(X test)
Out[31]:
         204
In [32]: len(y train)
Out[32]: 814
In [33]: len(y test)
Out[33]:
         204
In [34]: X train.shape
Out[34]: (814, 8)
In [35]: y_train.shape
Out[35]: (814,)
In [36]: X_train.head()
Out[36]:
                   gender impluse pressurehight pressurelow glucose kcm troponin
               age
          380
                29
                               90
                                           129
                                                        90
                                                              135.0 8.93
                                                                           0.003
                29
                                                                           0.004
          1093
                                           157
                                                              242.0 4.79
                               76
                                                        93
                50
                               89
                                           162
                                                              100.0 1.83
                                                                           0.005
          449
                        1
                                                        99
                                                                           0.046
          1147
                63
                                           122
                                                        60
                                                              188.0 2.19
          874
                48
                        0
                               64
                                           140
                                                        90
                                                              168.0 3.53
                                                                           0.026
In [37]: y_train
Out[37]: array([1, 0, 0, 1, 1, 1, 0, 0, 1, 1, 1, 0, 1, 1, 0, 1, 0, 0, 0, 1, 1,
                                      0,
                 1, 1, 1, 1, 0, 1, 1,
                                         0, 1, 1, 0,
                                                     0, 0, 0, 0, 1, 1,
                                                                        Θ,
                                                                           0, 0, 0,
                 0, 0, 0, 0, 0, 0, 1, 1, 1, 0, 0, 1, 0, 1, 0, 0, 1, 1, 0, 1, 1, 1,
                 0, 0, 0, 0, 1, 1, 0, 0, 1, 0, 1, 0, 1, 1, 0, 0, 0, 1, 0, 0, 0,
                 0, 1, 1, 0, 0, 0, 0, 0, 1, 1, 1, 0, 0, 1, 1, 1, 1, 0, 1, 1, 0,
                 0, 0, 1, 0, 1, 1, 0, 0, 0, 0, 1, 1,
                                                     0, 1,
                                                           1, 0, 1, 0, 1,
                                                                           0, 1, 0,
                 1, 0, 1, 0, 1, 0, 1, 1, 0, 1, 0, 0, 1, 0, 0, 0, 0, 0, 1, 0, 1, 0,
                 1, 1, 1, 1, 1, 1, 0, 1, 1, 1, 1, 1, 0, 0, 0, 0, 1, 1, 0, 1, 1, 1,
                 1, 1, 1, 0, 1, 1, 0, 0, 0, 1, 1, 1, 1, 0, 1,
                                                              1, 0, 1, 1, 0, 1, 0,
                 1, 1, 1, 1, 0, 0, 0, 0, 1,
                                            1, 1, 0,
                                                     0, 1, 0, 0, 1, 1,
                                                                       1,
                                                                           0, 0, 1,
                 0, 0, 1, 1, 1, 1, 0, 0, 1, 1, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 1, 0,
                 1,\ 0,\ 0,\ 1,\ 1,\ 0,\ 0,\ 1,\ 0,\ 1,\ 1,\ 0,\ 1,\ 1,\ 1,\ 0,\ 1,\ 0,\ 1,\ 1,\ 0,
                 0, 0, 0, 0, 0, 0, 1, 0, 1, 1, 1, 0, 0, 1, 1, 1, 0, 1, 1, 1, 0, 1,
                 1, 1, 0, 0, 0, 1, 1, 0, 0, 0, 1, 1, 0, 0, 0, 1, 1, 0, 0, 0, 0,
                 1, 1, 0, 0, 1, 1, 1, 0, 0, 0, 0, 1, 0, 0, 0, 1, 1, 0, 0, 0, 1, 1,
                 1, 1, 0, 0, 1, 1, 0, 0, 1, 0, 0, 1, 1, 0, 0, 1, 1, 0, 0, 1, 1, 0,
                   1, 1, 1, 0, 0, 0, 1, 0, 1, 1, 1,
                                                     1, 1, 0, 0, 1, 0, 1,
                                                                           0, 0, 0,
                 1, 0, 0, 1, 0, 0, 1, 1, 0, 0, 0, 1, 1, 1,
                                                           1,
                                                              1, 1, 0, 1,
                                                                           1,
                                                                             1, 1,
                 1, 0, 1, 1, 1, 0, 0, 0, 0, 0, 0, 0, 1, 1, 0, 0, 0, 0, 0, 0,
                 0, 0, 1, 0, 0, 0, 1, 0, 0, 1, 1, 0, 1, 1, 0, 1, 0, 1, 1, 1, 0,
                 0, 1, 1, 1, 1, 0, 1, 0, 1, 0, 0, 0, 1, 1, 1, 1, 1, 0, 1, 0, 1, 1,
                 1, 1, 0, 1, 0, 0, 1, 0, 0, 1, 1, 0, 0, 1, 0, 1, 0, 1, 1, 0, 1, 0,
                 1, 0, 1, 1, 0, 0, 1, 0, 0, 1, 1, 1, 1, 1, 0, 1, 0, 1, 0, 1, 1, 0,
                 1, 1, 1, 1, 1, 0, 1, 1, 0, 0, 0, 0, 0, 1, 1, 0, 0, 0, 1, 0, 0,
                   1, 1, 1, 0, 0, 1, 0, 0, 0, 1, 0, 1, 0, 0,
                                                              0, 1, 1, 1,
                                                                           0, 0, 0,
                 0, 0, 1, 0, 1, 0, 0, 1, 0, 0, 1, 0, 0, 0, 0, 0, 1, 0, 0, 1, 0, 1,
                 1, 1, 0, 0, 1, 0, 1, 1, 1, 0, 0, 0, 0, 1, 1, 1, 0, 1, 1, 0, 1, 1,
                 1, 0, 0, 1, 1, 1, 1, 1, 0, 1, 0, 0, 1, 0, 0, 1, 1, 1, 1, 1, 0, 0,
                 1, 0, 1, 1, 1, 0, 0, 1, 0, 1, 1, 1, 0, 0, 0, 0, 1, 0, 0, 1, 0, 0,
                 1, 1, 1, 1, 1, 1, 1, 1, 0, 1, 0, 1, 0, 0, 1, 0, 1, 1, 1, 0, 0, 1,
                 0, 1, 0, 1, 0, 1, 0, 1, 0, 0, 1, 1, 0, 1, 0, 0, 1, 0, 0, 0, 1,
                 1, 0, 0, 0, 1, 0, 0, 0, 1, 1, 0, 1, 0, 1, 1, 0, 0, 1, 0, 1, 0, 0,
                       0, 0, 0, 1, 0, 0, 0, 0, 1, 0,
                                                     0, 0, 1,
                                                              0, 1, 1,
                                                                       1,
                 0, 0, 1, 0, 1, 0, 1, 1, 1, 1, 1, 1, 1, 0, 1, 1, 0, 1, 0, 0, 0, 1,
                 1, 1, 1, 0, 1, 1, 0, 0, 0, 1, 1, 1, 1, 1, 1, 0, 1, 1, 1, 0, 1, 0,
                 1, 1, 0, 1, 1, 0, 0, 0, 0, 1, 0, 0, 0, 1, 1, 1, 1, 0, 0, 1, 0,
                 0, 0, 1, 0, 1, 0, 1, 1, 1, 0, 0, 1, 1, 0, 1, 1, 0, 0, 1, 0, 0, 0]
In [38]: from sklearn.linear model import LogisticRegression
         from sklearn.tree import DecisionTreeClassifier
```

from sklearn.ensemble import RandomForestClassifier

```
In [39]: lrmodel = LogisticRegression()
         lrmodel.fit(X train , y train)
        C:\Users\ASUS\AppData\Roaming\Python\Python311\site-packages\sklearn\linear model\ logistic.py:460: ConvergenceW
        arning: 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
        Please also refer to the documentation for alternative solver options:
            https://scikit-learn.org/stable/modules/linear model.html#logistic-regression
          n_iter_i = _check_optimize_result(
Out[39]: v LogisticRegression
         LogisticRegression()
In [40]: lrmodel.predict(X_test)
Out[40]: array([0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 1, 1, 0, 0, 1, 0, 1, 1, 0, 0, 1,
                 0, 0, 1, 1, 0, 0, 0, 1, 1, 0, 0, 1, 0, 0, 0, 1, 1, 0, 0, 0,
                 1, 0, 0, 1, 0, 0, 1, 0, 1, 1, 0, 0, 1, 0, 1, 1, 1, 0, 0, 0, 0, 1,
                 0,\ 1,\ 1,\ 1,\ 1,\ 0,\ 0,\ 1,\ 0,\ 0,\ 1,\ 0,\ 0,\ 0,\ 0,\ 0,\ 0,\ 1,\ 1,\ 0,\ 0,
                0, 0, 0, 1, 1, 1, 0, 0, 1, 0, 0, 0, 0, 1, 1, 1, 0, 1, 0, 1, 0, 1,
                1, 0, 1, 0, 1, 1, 1, 1, 0, 1, 0, 1, 0, 0, 1, 0, 1, 0, 1, 1, 1, 0,
                0, 0, 0, 0, 1, 1, 0, 1, 1, 0, 1, 0, 1, 0, 1, 1, 0, 1, 0, 0, 0,
                 1,\ 1,\ 0,\ 1,\ 0,\ 0,\ 0,\ 0,\ 1,\ 0,\ 0,\ 0,\ 1,\ 1,\ 1,\ 1,\ 1,\ 0,\ 0,\ 1,\ 0,
                0, 0, 0, 0, 0, 0, 1, 1, 1, 1, 0, 0, 0, 1, 1, 0, 1, 1, 0, 1, 1, 0,
                0. 0. 0. 0. 0. 01)
In [41]:
         from sklearn.metrics import accuracy score
         X_train_prediction = lrmodel.predict(X_train)
         training data accuracy = accuracy score(X train prediction , y train)
In [42]: training data accuracy
Out[42]: 0.8366093366093366
In [43]: from sklearn.metrics import accuracy score
         X_test_prediction = lrmodel.predict(X_test)
         testing data accuracy = accuracy score(X test prediction , y test)
In [44]: testing data accuracy
Out[44]: 0.8088235294117647
In [45]: dtmodel = DecisionTreeClassifier(random state=42)
         dtmodel.fit(X_train, y_train)
Out[45]: v
                   DecisionTreeClassifier
         DecisionTreeClassifier(random_state=42)
In [46]: from sklearn.metrics import accuracy_score
         X train prediction = dtmodel.predict(X train)
         training data accuracy = accuracy score(X train prediction , y train)
In [47]: training data accuracy
Out[47]: 1.0
In [48]: from sklearn.metrics import accuracy_score
         X test prediction = dtmodel.predict(X test)
         testing_data_accuracy = accuracy_score(X_test_prediction , y_test)
In [49]: testing_data_accuracy
Out[49]: 0.9803921568627451
In [50]: 111
         gender(0 for Female, 1 for Male)
         heart rate (impulse)
         systolic BP (pressurehight)
         diastolic BP (pressurelow)
         blood sugar(glucose)
         CK-MB (kcm)
         Test-Troponin (troponin)
         negative refers to the absence of a heart attack, while positive refers to the presence of a heart attack
```

```
Out[50]: '\nAqe\ngender(0 for Female, 1 for Male)\nheart rate (impulse)\nsystolic BP (pressurehight) \ndiastolic BP (pre
                             ssurelow)\nblood sugar(glucose) \nCK-MB (kcm)\nTest-Troponin (troponin) \nnegative refers to the absence of a h
                             eart attack, while positive refers to the presence of a heart attack\n\
In [51]: dtmodel.predict([[64,1,66,160,83,160.0,1.80,0.012]])
                         \verb|C:\USers\ASUS\AppData\Roaming\Python\Python311\site-packages\sklearn\base.py: 465: UserWarning: X does not have value of the packages of 
                        alid feature names, but DecisionTreeClassifier was fitted with feature names
                            warnings.warn(
Out[51]: array([0])
In [52]: input_data = (64,1,66,160,83,160.0,1.80,0.012)
                            input data as numpy array = np.asarray(input data)
                            input data reshaped = input_data_as_numpy_array.reshape(1 , -1)
                            prediction = dtmodel.predict(input data reshaped)
                            prediction = le.inverse_transform(prediction)
                            prediction
                        C:\Users\ASUS\AppData\Roaming\Python\Python311\site-packages\sklearn\base.py:465: UserWarning: X does not have v
                        alid feature names, but DecisionTreeClassifier was fitted with feature names
                            warnings.warn(
Out[52]: array(['negative'], dtype=object)
In [53]: input_data = (64,1,66,160,83,160.0,1.80,0.012)
                            input data as numpy array = np.asarray(input data)
                            input_data_reshaped = input_data_as_numpy_array.reshape(1 , -1)
                            prediction = dtmodel.predict(input_data_reshaped)
                            prediction = le.inverse_transform(prediction)
                            if prediction == "negative":
                                    print("The person does not have Heart attack problem.")
                            else:
                                    print("The person has Heart attack problem.")
                        The person does not have Heart attack problem.
                         \verb|C:\USers\ASUS\AppData\Roaming\Python\Python311\site-packages\sklearn\base.py: 465: UserWarning: X does not have value of the packages of 
                        alid feature names, but DecisionTreeClassifier was fitted with feature names
                          warnings.warn(
In [54]: import pickle
                            with open("Heart_attack model pickle" , "wb") as file:
                                      pickle.dump(dtmodel , file)
In [55]: import numpy as np
                            import pickle
                            with open("Heart attack model pickle" , "rb") as file:
                                        dtmodel = pickle.load(file)
In [56]: input data = (64,1,66,160,83,160.0,1.80,0.012)
                            input_data_as_numpy_array = np.asarray(input_data)
                            input_data_reshaped = input_data_as_numpy_array.reshape(1 , -1)
                            prediction = dtmodel.predict(input_data_reshaped)
                            prediction = le.inverse_transform(prediction)
                            if prediction == "negative":
                                    print("The person does not have Heart attack problem.")
                            else:
                                    print("The person has Heart attack problem.")
                        The person does not have Heart attack problem.
                         \verb|C:\USers\ASUS\AppData\Roaming\Python\Python311\site-packages\sklearn\base.py: 465: UserWarning: X does not have value of the packages of 
                        alid feature names, but DecisionTreeClassifier was fitted with feature names
                            warnings.warn(
In [58]: #age in years
                            age = int(input(""))
                            #gender(0 for Female, 1 for Male)
                            gender = int(input(""))
                            #impluse
                            impluse = int(input(""))
                            #pressurehight
                            pressurehight = int(input(""))
                            #pressurelow
                            pressurelow = int(input(""))
                            #glucose
                            glucose = float(input(""))
                            kcm = float(input(""))
                            #troponin
                            troponin = float(input(""))
```

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```
input_data = (age,gender,impluse,pressurehight,pressurelow,glucose,kcm,troponin)
input_data_as_numpy_array = np.asarray(input_data)
input_data_reshaped = input_data_as_numpy_array.reshape(1 , -1)
prediction = dtmodel.predict(input_data_reshaped)
prediction = le.inverse_transform(prediction)
if prediction == "negative":
    print("The person does not have Heart attack problem.")
else:
    print("The person has Heart attack problem.")
```

The person does not have Heart attack problem.

C:\Users\ASUS\AppData\Roaming\Python\Python311\site-packages\sklearn\base.py:465: UserWarning: X does not have v
alid feature names, but DecisionTreeClassifier was fitted with feature names
warnings.warn(

In []:

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