```
In [2]: #Name : Shreya Sharma
         #Roll no. : 46
         #Sectin : 3B
         #Date : 08/10/2024
 In [2]: #Aim : To perform operation on knn classifier
 In [1]: import pandas as pd
         import matplotlib.pyplot as plt
         import numpy as np
         import seaborn as sns
         from sklearn.model_selection import train_test_split
         import warnings
         warnings.filterwarnings('ignore')
 In [2]: import os
 In [3]: os.getcwd()
 Out[3]: 'C:\\Users\\pravi'
 In [4]: os.chdir("C:\\Users\\pravi\\Desktop")
 In [5]: df=pd.read csv("framingham.csv")
 In [6]: df.head()
                                                                                                             sysBP diaBP
 Out[6]:
                      education currentSmoker cigsPerDay BPMeds
                                                                 prevalentStroke
                                                                                prevalentHyp diabetes totChol
            male age
         0
                            4 0
                                           0
                                                                             0
                                                                                          0
                                                                                                       195.0
                                                                                                              106.0
                                                                                                                     70.0 26
               1
                  39
                                                     0.0
                                                             0.0
                                                                                                  0
         1
               0
                  46
                            20
                                           0
                                                     0.0
                                                             0.0
                                                                             0
                                                                                          0
                                                                                                  0
                                                                                                       250.0
                                                                                                              121 0
                                                                                                                     81.0 28
         2
                                                    20.0
                                                                             0
                                                                                          0
                                                                                                       245.0
                                                                                                              127.5
                                                                                                                     80.0 25
               1
                  48
                            1.0
                                                             0.0
                                                                                                  0
         3
               0
                  61
                            3.0
                                                    30.0
                                                             0.0
                                                                             0
                                                                                                  0
                                                                                                       225.0
                                                                                                              150.0
                                                                                                                     95.0
                                                                                                                         28
         4
                            3.0
                                                    23.0
                                                             0.0
                                                                             0
                                                                                          0
                                                                                                  0
                                                                                                       285.0
                                                                                                              130.0
               0
                  46
                                                                                                                     84.0 23
         4
                                                                                                                         | b
In [12]: df.info()
        <class 'pandas.core.frame.DataFrame'>
        RangeIndex: 4238 entries, 0 to 4237
        Data columns (total 16 columns):
         # Column
                             Non-Null Count Dtype
        - - -
             -----
                              -----
         0
                              4238 non-null
             male
                                               int64
                              4238 non-null
         1
             age
                                               int64
         2
             education
                              4133 non-null
                                               float64
         3
             currentSmoker
                              4238 non-null
                                               int64
             cigsPerDay
                              4209 non-null
                                               float64
                                               float64
                              4185 non-null
         5
             BPMeds
             prevalentStroke 4238 non-null
         6
                                               int64
         7
             prevalentHyp
                              4238 non-null
                                               int64
                                               int64
         8
                              4238 non-null
             diabetes
         9
             totChol
                              4188 non-null
                                               float64
         10 sysBP
                              4238 non-null
                                               float64
                                               float64
         11 diaBP
                              4238 non-null
         12
             BMI
                              4219 non-null
                                               float64
         13 heartRate
                              4237 non-null
                                               float64
                              3850 non-null
                                               float64
         14 glucose
                              4238 non-null
         15 TenYearCHD
                                               int64
        dtypes: float64(9), int64(7)
        memory usage: 529.9 KB
In [15]: df.isna().sum()
```

```
0
          age
                                105
          education
          currentSmoker
                                  0
          cigsPerDay
                                29
          BPMeds
                                53
          prevalentStroke
                                  0
          prevalentHyp
                                  0
                                  0
          diabetes
          totChol
                                50
          sysBP
                                  0
          diaBP
                                  0
          BMI
                                19
          heartRate
                                 1
          glucose
                                388
           {\sf TenYearCHD}
                                  0
          dtype: int64
In [17]: df
Out[17]:
                male
                      age education currentSmoker cigsPerDay BPMeds prevalentStroke prevalentHyp diabetes totChol sysBP diaBP
             0
                                 4.0
                                                  0
                                                            0.0
                                                                                      0
                                                                                                    0
                                                                                                                  195.0
                                                                                                                                  70.0
                   1
                       39
                                                                                                                          106.0
             1
                   0
                       46
                                 2.0
                                                  0
                                                            0.0
                                                                     0.0
                                                                                      0
                                                                                                    0
                                                                                                             0
                                                                                                                  250.0
                                                                                                                          121.0
                                                                                                                                 81.0
             2
                                                  1
                   1
                       48
                                 1.0
                                                           20.0
                                                                     0.0
                                                                                      0
                                                                                                    0
                                                                                                             0
                                                                                                                  245.0
                                                                                                                                 80.0
                                                                                                                          127.5
             3
                   0
                       61
                                 3.0
                                                           30.0
                                                                     0.0
                                                                                      0
                                                                                                    1
                                                                                                             0
                                                                                                                  225.0
                                                                                                                          150.0
                                                                                                                                 95.0
             4
                       46
                                 3.0
                                                  1
                                                           23.0
                                                                                      0
                                                                                                    0
                                                                                                                  285.0
                                                                                                                          130.0
                                                                                                                                  84.0
             ...
          4233
                                                                                      0
                                                                                                                         179.0
                       50
                                 1.0
                                                  1
                                                            1.0
                                                                     0.0
                                                                                                    1
                                                                                                                                 92.0
                   1
                                                                                                             0
                                                                                                                  313.0
          4234
                       51
                                 3.0
                                                           43.0
                                                                     0.0
                                                                                      0
                                                                                                    0
                                                                                                                  207.0
                                                                                                                          126.5
                                                                                                                                 80.0
          4235
                       48
                                 2.0
                                                  1
                                                           20.0
                                                                    NaN
                                                                                      0
                                                                                                    0
                                                                                                                  248.0
                                                                                                                          131.0
                                                                                                                                  72.0
          4236
                   0
                       44
                                  1.0
                                                           15.0
                                                                     0.0
                                                                                      0
                                                                                                    0
                                                                                                             0
                                                                                                                  210.0
                                                                                                                          126.5
                                                                                                                                 87.0
                                                  0
                       52
                                 2.0
                                                                                      0
                                                                                                    0
                                                                                                                         133.5
          4237
                   0
                                                            0.0
                                                                     0.0
                                                                                                             0
                                                                                                                  269.0
                                                                                                                                 83.0
         4238 rows × 16 columns
In [19]: df['glucose'].fillna(value = df['glucose'].mean(),inplace=True)
In [21]: df['education'].fillna(value = df['education'].mean(),inplace=True)
In [23]: df['heartRate'].fillna(value = df['heartRate'].mean(),inplace=True)
In [25]: df['BMI'].fillna(value = df['BMI'].mean(),inplace=True)
In [27]: df['cigsPerDay'].fillna(value = df['cigsPerDay'].mean(),inplace=True)
In [29]: df['totChol'].fillna(value = df['totChol'].mean(),inplace=True)
In [31]: df['BPMeds'].fillna(value = df['BPMeds'].mean(),inplace=True)
In [33]: df.isna().sum()
Out[33]: male
                               0
                               0
          age
                               0
          education
          {\tt currentSmoker}
                               0
          cigsPerDay
                               0
          BPMeds
                               0
          prevalentStroke
                               0
          prevalentHyp
                               0
                               0
          diabetes
          totChol
                               0
          sysBP
                               0
          diaBP
                               0
          BMI
                               0
                               0
          heartRate
                                0
          glucose
          TenYearCHD
                               0
          dtype: int64
In [35]: x = df.drop("TenYearCHD",axis=1)
          y = df['TenYearCHD']
```

0

Out[15]: male

In [37]:	х												
Out[37]:		male	age	education	currentSmoker	cigsPerDay	BPMeds	prevalentStroke	prevalentHyp	diabetes	totChol	sysBP	diaBP
	0	1	39	4.0	0	0.0	0.00000	0	0	0	195.0	106.0	70.0
	1	0	46	2.0	0	0.0	0.00000	0	0	0	250.0	121.0	81.0
	2	1	48	1.0	1	20.0	0.00000	0	0	0	245.0	127.5	80.0
	3	0	61	3.0	1	30.0	0.00000	0	1	0	225.0	150.0	95.0
	4	0	46	3.0	1	23.0	0.00000	0	0	0	285.0	130.0	84.0
	4233	1	50	1.0	1	1.0	0.00000	0	1	0	313.0	179.0	92.0
	4234	1	51	3.0	1	43.0	0.00000	0	0	0	207.0	126.5	80.0
	4235	0	48	2.0	1	20.0	0.02963	0	0	0	248.0	131.0	72.0
	4236	0	44	1.0	1	15.0	0.00000	0	0	0	210.0	126.5	87.0
	4237	0	52	2.0	0	0.0	0.00000	0	0	0	269.0	133.5	83.0
	4238 rows × 15 columns												
	4												)
In [39]:	x_tra	ain,x_	test,	y_train,y	_test = train_	_test_split	(x,y,test	t_size=0.2,rand	lom_state=42)				
In [41]:	x train												
Out[41]:		male	age	education	currentSmoker	cigsPerDay	BPMeds	prevalentStroke	prevalentHyp	diabetes	totChol	sysBP	diaBP
	3252	1	40	4.0	1	30.0	0.0	0	0	0	205.0	131.0	81.0
	3946	0	57	2.0	0	0.0	0.0	0	1	0	250.0	152.5	92.5
	1261	0	47	1.0	0	0.0	0.0	0	0	0	230.0	123.0	71.0
	2536	1	41	2.0	1	30.0	0.0	0	0	0	228.0	113.0	82.5
	4089	0	64	1.0	0	0.0	0.0	0	1	0	232.0	149.5	84.0
	3444	0	36	1.0	1	5.0	0.0	0	1	0	222.0	147.0	94.0
	466	0	57	3.0	1	15.0	0.0	0	0	0	250.0	125.0	74.0
	3092	0	60	2.0	0	0.0	0.0	0	1	0	298.0	133.0	89.0
	3772	1	39	2.0	1	10.0	0.0	0	0	0	215.0	102.0	64.5
	860	0	35	2.0	0	0.0	0.0	0	0	0	248.0	107.0	73.0
	3390 r	ows ×	15 col	umns									
	4												Þ
In [43]:	y_tra	ain											
Out[43]:	3252 3946 1261 2536 4089	0 0 0											
	3444 466 3092 3772 860 Name	0 0 0 0	′earCl	HD, Length	ı: 3390, dtype	: int64							

In [45]: x\_test

	3188	1	63	1.0	0	0.0	0.0	0	1	0	190.0	148.0	90.0
	764	1	45	3.0	0	0.0	0.0	0	0	0	162.0	125.0	89.0
	3264	0	51	1.0	1	2.0	0.0	0	0	0	261.0	127.0	81.0
	1967	1	45	3.0	1	30.0	0.0	0	0	0	250.0	126.0	89.5
	2185	0	45	2.0	1	3.0	0.0	0	0	0	250.0	130.0	80.0
	3303	1	47	1.0	0	0.0	0.0	0	0	0	259.0	139.0	79.0
	4056	1	44	2.0	0	0.0	0.0	0	0	0	254.0	130.0	80.0
	4210	1	50	1.0	0	0.0	0.0	0	0	0	282.0	126.5	88.0
	3971	1	64	3.0	0	0.0	0.0	0	1	1	195.0	176.0	78.0
	2540	1	55	3.0	1	20.0	0.0	0	0	0	214.0	110.0	71.0
In [47]: Out[47]:	3188 764 3264 1967 2185 3303 4056 4210 3971 2540	0 0 0 0 0  1 0 0		ength: 848, dty	pe: int64	1							
In [49]:	<pre>from sklearn.neighbors import KNeighborsClassifier knn = KNeighborsClassifier(n_neighbors=5, p=2, metric='minkowski') knn.fit(x_train, y_train) acc = knn score(x_test y_test)*100</pre>												

male age education currentSmoker cigsPerDay BPMeds prevalentStroke prevalentHyp diabetes totChol sysBP diaBP

83.13679245283019

print(acc)

Out[45]:

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acc = knn.score(x\_test,y\_test)\*100