

```
In [1]: #Experiment no 8 To Perform and find the accuracy of KNN Classifier.
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```
In [2]: #Name: Shravani M Karne
#Roll no.: 39
#Sec:A
#Year:3rd Year
#Sub: Big Data Analysis(ET 2 Lab)
```

```
In [3]: import pandas as pd
import os
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 [4]: os.getcwd()
```

```
Out[4]: 'C:\\Users\\rautp'
```

```
In [5]: os.chdir('C:\\Users\\rautp')
```

```
In [6]: df=pd.read_csv('framingham.csv')
```

```
In [7]: df.head()
```

```
Out[7]:
```

| | male | age | education | currentSmoker | cigsPerDay | BPMeds | prevalentStroke | prevalentHyp | diabetes | totChol |
|---|------|-----|-----------|---------------|------------|--------|-----------------|--------------|----------|---------|
| 0 | 1 | 39 | 4.0 | 0 | 0.0 | 0.0 | 0 | 0 | 0 | 195.0 |
| 1 | 0 | 46 | 2.0 | 0 | 0.0 | 0.0 | 0 | 0 | 0 | 250.0 |
| 2 | 1 | 48 | 1.0 | 1 | 20.0 | 0.0 | 0 | 0 | 0 | 245.0 |
| 3 | 0 | 61 | 3.0 | 1 | 30.0 | 0.0 | 0 | 1 | 0 | 225.0 |
| 4 | 0 | 46 | 3.0 | 1 | 23.0 | 0.0 | 0 | 0 | 0 | 285.0 |

```
In [8]: df.tail()
```

```
Out[8]:
```

| | male | age | education | currentSmoker | cigsPerDay | BPMeds | prevalentStroke | prevalentHyp | diabetes | totC |
|------|------|-----|-----------|---------------|------------|--------|-----------------|--------------|----------|------|
| 4233 | 1 | 50 | 1.0 | 1 | 1.0 | 0.0 | 0 | 1 | 0 | 31 |
| 4234 | 1 | 51 | 3.0 | 1 | 43.0 | 0.0 | 0 | 0 | 0 | 20 |
| 4235 | 0 | 48 | 2.0 | 1 | 20.0 | NaN | 0 | 0 | 0 | 24 |
| 4236 | 0 | 44 | 1.0 | 1 | 15.0 | 0.0 | 0 | 0 | 0 | 21 |
| 4237 | 0 | 52 | 2.0 | 0 | 0.0 | 0.0 | 0 | 0 | 0 | 26 |

```
In [9]: df.info
```

```
Out[9]: <bound method DataFrame.info of
PMed5 \
0      1      39      4.0      0      0.0      0.0
1      0      46      2.0      0      0.0      0.0
2      1      48      1.0      1      20.0      0.0
3      0      61      3.0      1      30.0      0.0
4      0      46      3.0      1      23.0      0.0
...
4233    1      50      1.0      1      1.0      0.0
4234    1      51      3.0      1      43.0      0.0
4235    0      48      2.0      1      20.0      NaN
4236    0      44      1.0      1      15.0      0.0
4237    0      52      2.0      0      0.0      0.0

prevalentStroke  prevalentHyp  diabetes  totChol  sysBP  diaBP  BMI \
0                0            0          0    195.0  106.0   70.0  26.97
1                0            0          0    250.0  121.0   81.0  28.73
2                0            0          0    245.0  127.5   80.0  25.34
3                0            1          0    225.0  150.0   95.0  28.58
4                0            0          0    285.0  130.0   84.0  23.10
...
4233              0            1          0    313.0  179.0   92.0  25.97
4234              0            0          0    207.0  126.5   80.0  19.71
4235              0            0          0    248.0  131.0   72.0  22.00
4236              0            0          0    210.0  126.5   87.0  19.16
4237              0            0          0    269.0  133.5   83.0  21.47

heartRate  glucose  TenYearCHD
0      80.0      77.0          0
1      95.0      76.0          0
2      75.0      70.0          0
3      65.0     103.0          1
4      85.0      85.0          0
...
4233      66.0      86.0          1
4234      65.0      68.0          0
4235      84.0      86.0          0
4236      86.0      NaN          0
4237      80.0     107.0          0

[4238 rows x 16 columns]>
```

```
In [10]: df.describe()
```

```
Out[10]:
```

| | male | age | education | currentSmoker | cigsPerDay | BPMeds | prevalentStroke | preva |
|-------|-------------|-------------|-------------|---------------|-------------|-------------|-----------------|-------|
| count | 4238.000000 | 4238.000000 | 4133.000000 | 4238.000000 | 4209.000000 | 4185.000000 | 4238.000000 | 4238 |
| mean | 0.429212 | 49.584946 | 1.978950 | 0.494101 | 9.003089 | 0.029630 | 0.005899 | 0 |
| std | 0.495022 | 8.572160 | 1.019791 | 0.500024 | 11.920094 | 0.169584 | 0.076587 | 0 |
| min | 0.000000 | 32.000000 | 1.000000 | 0.000000 | 0.000000 | 0.000000 | 0.000000 | 0 |
| 25% | 0.000000 | 42.000000 | 1.000000 | 0.000000 | 0.000000 | 0.000000 | 0.000000 | 0 |
| 50% | 0.000000 | 49.000000 | 2.000000 | 0.000000 | 0.000000 | 0.000000 | 0.000000 | 0 |
| 75% | 1.000000 | 56.000000 | 3.000000 | 1.000000 | 20.000000 | 0.000000 | 0.000000 | 1 |
| max | 1.000000 | 70.000000 | 4.000000 | 1.000000 | 70.000000 | 1.000000 | 1.000000 | 1 |

```
In [11]: df.isna().sum()
```

```
Out[11]: male          0
         age           0
         education     105
         currentSmoker  0
         cigsPerDay     29
         BPMeds        53
         prevalentStroke 0
         prevalentHyp   0
         diabetes       0
         totChol        50
         sysBP          0
         diaBP          0
         BMI           19
         heartRate      1
         glucose        388
         TenYearCHD     0
         dtype: int64
```

```
In [12]: df['glucose'].fillna(value = df['glucose'].mean(), inplace=True)
```

```
In [13]: df['education'].fillna(value = df['education'].mean(), inplace=True)
```

```
In [14]: df['heartRate'].fillna(value = df['heartRate'].mean(), inplace=True)
```

```
In [15]: df['BMI'].fillna(value = df['BMI'].mean(), inplace=True)
```

```
In [16]: df['cigsPerDay'].fillna(value = df['cigsPerDay'].mean(), inplace=True)
```

```
In [17]: df['totChol'].fillna(value = df['totChol'].mean(), inplace=True)
```

```
In [18]: df['BPMeds'].fillna(value = df['BPMeds'].mean(), inplace=True)
```

```
In [19]: df.isna().sum()
```

```
Out[19]: male          0
         age           0
         education     0
         currentSmoker  0
         cigsPerDay     0
         BPMeds        0
         prevalentStroke 0
         prevalentHyp   0
         diabetes       0
         totChol        0
         sysBP          0
         diaBP          0
         BMI           0
         heartRate      0
         glucose        0
         TenYearCHD     0
         dtype: int64
```

```
In [20]: df.isna().sum()
```

```
Out[20]: male      0
age      0
education 0
currentSmoker 0
cigsPerDay 0
BPMeds    0
prevalentStroke 0
prevalentHyp 0
diabetes   0
totChol    0
sysBP      0
diaBP      0
BMI         0
heartRate  0
glucose     0
TenYearCHD 0
dtype: int64
```

```
In [21]: #Splitting the dependent and independent variables.
x = df.drop("TenYearCHD",axis=1)
y = df['TenYearCHD']
```

```
In [22]: x #checking the features
```

```
Out[22]:
```

| | male | age | education | currentSmoker | cigsPerDay | BPMeds | prevalentStroke | prevalentHyp | diabetes | totChol |
|------|------|-----|-----------|---------------|------------|---------|-----------------|--------------|----------|---------|
| 0 | 1 | 39 | 4.0 | 0 | 0.0 | 0.00000 | 0 | 0 | 0 | 19 |
| 1 | 0 | 46 | 2.0 | 0 | 0.0 | 0.00000 | 0 | 0 | 0 | 25 |
| 2 | 1 | 48 | 1.0 | 1 | 20.0 | 0.00000 | 0 | 0 | 0 | 24 |
| 3 | 0 | 61 | 3.0 | 1 | 30.0 | 0.00000 | 0 | 1 | 0 | 22 |
| 4 | 0 | 46 | 3.0 | 1 | 23.0 | 0.00000 | 0 | 0 | 0 | 28 |
| ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... |
| 4233 | 1 | 50 | 1.0 | 1 | 1.0 | 0.00000 | 0 | 1 | 0 | 31 |
| 4234 | 1 | 51 | 3.0 | 1 | 43.0 | 0.00000 | 0 | 0 | 0 | 20 |
| 4235 | 0 | 48 | 2.0 | 1 | 20.0 | 0.02963 | 0 | 0 | 0 | 24 |
| 4236 | 0 | 44 | 1.0 | 1 | 15.0 | 0.00000 | 0 | 0 | 0 | 21 |
| 4237 | 0 | 52 | 2.0 | 0 | 0.0 | 0.00000 | 0 | 0 | 0 | 26 |

4238 rows × 15 columns

Train test Split

```
In [23]: x_train,x_test,y_train,y_test = train_test_split(x,y,test_size=0.2,random_state=42)
```

```
In [24]: y_train
```

```
Out[24]: 3252    0
          3946    0
          1261    0
          2536    0
          4089    0
          ..
          3444    0
          466     0
          3092    0
          3772    0
          860     0
Name: TenYearCHD, Length: 3390, dtype: int64
```

KNN Classifier

```
In [25]: 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
print(acc)

83.13679245283019
```