SVM classifier

```
In [ ]: #Name : Vaibhav Laxman Karale
         #Roll no. 58
         #sub:E.T.1
         #Section :3A
         #Date:05/10/2024
In [82]:
          # Aim: to perform operation on svm classifier
In [45]: 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 [46]: import os
 In [5]:
         os.getcwd()
Out[5]: 'C:\\Users\\DELL'
 In [6]: | os.chdir("C:\\Users\\DELL\\OneDrive\\Desktop")
In [48]: df=pd.read csv("C:\\Users\\DELL\\OneDrive\\Desktop\\framingham - Copy.csv")
```

In [49]: df

Out[49]:

	male	age	education	currentSmoker	cigsPerDay	BPMeds	prevalentStroke	prevalentHyp	diab
C	1	39	4.0	0	0.0	0.0	0	0	
1	0	46	2.0	0	0.0	0.0	0	0	
2	1	48	1.0	1	20.0	0.0	0	0	
3	0	61	3.0	1	30.0	0.0	0	1	
4	0	46	3.0	1	23.0	0.0	0	0	
4233	1	50	1.0	1	1.0	0.0	0	1	
4234	1	51	3.0	1	43.0	0.0	0	0	
4235	0	48	2.0	1	20.0	NaN	0	0	
4236	0	44	1.0	1	15.0	0.0	0	0	
4237	0	52	2.0	0	0.0	0.0	0	0	

4238 rows × 16 columns

In [50]: df.head()

Out[50]:

	male	age	education	currentSmoker	cigsPerDay	BPMeds	prevalentStroke	prevalentHyp	diabete
0	1	39	4.0	0	0.0	0.0	0	0	
1	0	46	2.0	0	0.0	0.0	0	0	
2	1	48	1.0	1	20.0	0.0	0	0	
3	0	61	3.0	1	30.0	0.0	0	1	
4	0	46	3.0	1	23.0	0.0	0	0	

In [51]: df.tail()

Out[51]:

	male	age	education	currentSmoker	cigsPerDay	BPMeds	prevalentStroke	prevalentHyp	diab
4233	1	50	1.0	1	1.0	0.0	0	1	
4234	1	51	3.0	1	43.0	0.0	0	0	
4235	0	48	2.0	1	20.0	NaN	0	0	
4236	0	44	1.0	1	15.0	0.0	0	0	
4237	0	52	2.0	0	0.0	0.0	0	0	
4 6	-	-	_						

In [52]: df.shape

Out[52]: (4238, 16)

```
In [53]: df.size
Out[53]: 67808
In [54]: | df.ndim
Out[54]: 2
In [55]: df.info()
         <class 'pandas.core.frame.DataFrame'>
         RangeIndex: 4238 entries, 0 to 4237
         Data columns (total 16 columns):
                               Non-Null Count Dtype
              Column
              _____
                               -----
          0
              male
                               4238 non-null
                                               int64
          1
                               4238 non-null
                                               int64
              age
          2
              education
                               4133 non-null
                                               float64
                                               int64
          3
              currentSmoker
                               4238 non-null
          4
              cigsPerDay
                               4209 non-null
                                               float64
          5
              BPMeds
                               4185 non-null
                                               float64
          6
              prevalentStroke 4238 non-null
                                               int64
          7
              prevalentHyp
                               4238 non-null
                                               int64
                               4238 non-null
          8
              diabetes
                                               int64
          9
              totChol
                               4188 non-null
                                               float64
          10 sysBP
                               4238 non-null
                                               float64
```

dtypes: float64(9), int64(7)

memory usage: 529.9 KB

In [56]: df.describe()

11 diaBP

13 heartRate

14 glucose15 TenYearCHD

12 BMI

Out[56]:

	male	age	education	currentSmoker	cigsPerDay	BPMeds	prevalentStrc
coun	t 4238.000000	4238.000000	4133.000000	4238.000000	4209.000000	4185.000000	4238.0000
meai	0.429212	49.584946	1.978950	0.494101	9.003089	0.029630	0.0058
sto	0.495022	8.572160	1.019791	0.500024	11.920094	0.169584	0.0765
miı	0.000000	32.000000	1.000000	0.000000	0.000000	0.000000	0.0000
25%	0.000000	42.000000	1.000000	0.000000	0.000000	0.000000	0.0000
50%	0.000000	49.000000	2.000000	0.000000	0.000000	0.000000	0.0000
75%	1.000000	56.000000	3.000000	1.000000	20.000000	0.000000	0.0000
max	1.000000	70.000000	4.000000	1.000000	70.000000	1.000000	1.0000

float64

float64

float64

float64

int64

4238 non-null

4219 non-null

4237 non-null

3850 non-null

4238 non-null

```
In [57]: | df.isnull().sum()
Out[57]: male
                                0
                                0
         age
                              105
         education
         currentSmoker
                               0
         cigsPerDay
                               29
         BPMeds
                               53
         prevalentStroke
                               0
         prevalentHyp
                               0
         diabetes
                               0
         totChol
                               50
          sysBP
                               0
         diaBP
                               0
         BMI
                               19
         heartRate
                               1
                              388
         glucose
         TenYearCHD
                               0
         dtype: int64
In [59]: print(x)
                    39.
                                                          ]
          1.
                            4.
                                       26.97
                                              80.
                                                      77.
                                                           1
                            2.
                                       28.73
                                              95.
                                                      76.
              0.
                    46.
                                                           ]
           [
              1.
                    48.
                            1.
                                       25.34
                                              75.
                                                      70.
                            2.
                    48.
                                       22.
                                              84.
                                                          1
           [
              0.
                                                      86.
           Γ
              0.
                    44.
                            1.
                                       19.16
                                              86.
                                                        nan]
              0.
                    52.
                            2.
                                       21.47 80.
                                                     107. ]]
In [60]: print(y)
          [000...000]
In [61]: print(y)
          [0 0 0 ... 0 0 0]
In [62]:
          import matplotlib.pyplot as plt
           import seaborn as sns
           import numpy as np
In [63]:
         #Splitting testdata into X_train, X_test, y_train, y_test
         from sklearn.model_selection import train_test_split
         x_train,x_test,y_train,y_test = train_test_split(x,y,test_size=.3,random_state=42)
In [64]: |print(x_train)
                               ... 22.19 70.
          [[ 0.
                  42.
                         1.
                                                  nan]
           [ 0.
                                               87.
                  64.
                         1.
                               ... 29.97 65.
           [ 0.
                  39.
                         2.
                               ... 20.55 68.
                                                  nan]
                         2.
           [ 0.
                  60.
                               ... 25.09 83.
                                               81.
                                                     1
           [ 1.
                  39.
                         2.
                               ... 24.5 68.
                                               62.
                                                     ]
           [ 0.
                  35.
                         2.
                               ... 20.64 90.
                                               80.
                                                     ]]
```

```
In [65]: print(y_train)
      [0 0 0 ... 0 0 0]
```

missing value treatment

```
In [66]:
         df['glucose'].fillna(value = df['glucose'].mean(),inplace=True)
In [67]: df['education'].fillna(value = df['education'].mean(),inplace=True)
In [68]: df['heartRate'].fillna(value = df['heartRate'].mean(),inplace=True)
In [69]: df['BMI'].fillna(value = df['BMI'].mean(),inplace=True)
In [70]: | df['cigsPerDay'].fillna(value = df['cigsPerDay'].mean(),inplace=True)
          df['totChol'].fillna(value = df['totChol'].mean(),inplace=True)
In [71]:
In [72]: |df['BPMeds'].fillna(value = df['BPMeds'].mean(),inplace=True)
In [73]: df.isna().sum()
Out[73]: male
                             0
                             0
         age
         education
                             0
         currentSmoker
                             0
         cigsPerDay
         BPMeds
                             0
         prevalentStroke
                             0
         prevalentHyp
                             0
         diabetes
                             0
         totChol
                             0
         sysBP
                             0
         diaBP
                             0
         BMI
                             0
         heartRate
                             0
         glucose
                             0
         TenYearCHD
         dtype: int64
In [76]: #Splitting the dependent and independent variables.
         x = df.drop("TenYearCHD",axis=1)
         y = df['TenYearCHD']
```

```
In [77]: x
Out[77]:
                               education currentSmoker cigsPerDay BPMeds prevalentStroke prevalentHyp diab
                   male age
                0
                      1
                           39
                                      4.0
                                                                  0.0
                                                                        0.00000
                                                                                               0
                                                                                                             0
                1
                      0
                           46
                                      2.0
                                                       0
                                                                  0.0
                                                                        0.00000
                                                                                               0
                                                                                                             0
                2
                                                                 20.0
                                                                        0.00000
                                                                                               0
                                                                                                             0
                      1
                           48
                                      1.0
                                                       1
                3
                                                                                               0
                      0
                                      3.0
                                                                  30.0
                                                                        0.00000
                                                                                                              1
                           61
                                                       1
                                                                  23.0
                                                                        0.00000
                                                                                               0
                                                                                                              0
                4
                           46
                                      3.0
             4233
                                                                        0.00000
                           50
                                      1.0
                                                       1
                                                                  1.0
                                                                                               0
                                                                                                              1
                      1
                                                                                               0
                                                                                                             0
             4234
                           51
                                                       1
                                                                        0.00000
                                      3.0
                                                                 43.0
             4235
                           48
                                     2.0
                                                                 20.0
                                                                        0.02963
                                                                                               0
                                                                                                              0
             4236
                           44
                                      1.0
                                                       1
                                                                  15.0
                                                                        0.00000
                                                                                               0
                                                                                                              0
            4237
                                      2.0
                                                                  0.0
                                                                        0.00000
                                                                                               0
                                                                                                              0
                      0
                           52
           4238 rows × 15 columns
```

train test split

```
x_train,x_test,y_train,y_test = train_test_split(x,y,test_size=0.2,random_state=42
In [78]:
In [79]: y_train
Out[79]:
         3252
                  0
         3946
                  0
         1261
                  0
         2536
                  0
         4089
                  0
         3444
         466
         3092
                  0
         3772
                  0
         Name: TenYearCHD, Length: 3390, dtype: int64
```

SVM Classifier

```
In [83]: from sklearn.svm import SVC
         from sklearn.metrics import accuracy_score
         svc=SVC()
         svc.fit(x_test,y_test)
         acc = svc.score(x_test,y_test)*100
         print(acc)
         85.37735849056604
```

In []: