```
#experiment No. 12
#Aim : To perform and find the accuracy of Support Vector Machine
Algorithm i.e. SVM
# Name : Samruddhi Umap
# Roll no : 67
# Sec: C
# Subject : ET1
# Date :27/09/2024
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')
import os
os.getcwd()
'C:\\Users\\HP\\Desktop'
os.chdir("C:\\Users\\HP\\Desktop")
df=pd.read csv("framingham.csv")
df.head()
   male age education currentSmoker cigsPerDay BPMeds
prevalentStroke
                    4.0
                                               0.0
                                                        0.0
0
      1
          39
                                     0
0
1
                    2.0
                                               0.0
                                                        0.0
      0
          46
0
2
          48
                    1.0
                                              20.0
      1
                                                        0.0
0
3
      0
          61
                    3.0
                                              30.0
                                                        0.0
0
4
      0
          46
                    3.0
                                              23.0
                                                        0.0
   prevalentHyp diabetes totChol sysBP diaBP
                                                    BMI heartRate
glucose \
              0
                        0
                             195.0 106.0 70.0 26.97
                                                               80.0
77.0
              0
                             250.0 121.0
                                            81.0 28.73
                                                               95.0
                        0
1
76.0
              0
                             245.0 127.5
                                            80.0 25.34
                                                               75.0
```

70 0							
70.0 3	1	0	225.0	150.0	95.0	28.58	65.0
103.0	0	•	205 0	120.0	04.0	22 10	05.0
4 85.0	0	0	285.0	130.0	84.0	23.10	85.0
83.0							
TenYea 0 1 2 3 4	rCHD 0 0 0 1						
df.descri	be()						
	male	ā	age (education	n curr	rentSmoker	
cigsPerDay	y \ 38.000000	4238.0000	000 <i>4</i> 1	33.00000	a 45	238.000000	
4209.0000		4230.0000	700 41.	33.00000	J 1 2	230.000000	
mean	0.429212	49.5849	946	1.978950	9	0.494101	
9.003089 std	0.495022	8.5721	L60	1.01979	1	0.500024	
11.920094	0 000000	22 000	200	1 00000	2	0.000000	
min 0.000000	0.000000	32.0000	900	1.000000	ี่	0.000000	
25% 0.000000	0.000000	42.0000	900	1.000000	9	0.000000	
50%	0.000000	49.0000	900	2.000000	9	0.000000	
0.000000 75%	1.000000	56.0000	900	3.000000	9	1.000000	
20.000000							
max 70.000000	1.000000	70.000	900	4.00000	9	1.000000	
70.00000							
totChol	BPMeds	prevalent	tStroke	prevale	entHyp	diabet	es
count 41	85.000000	4238	. 000000	4238.0	900000	4238.0000	00
4188.0000 mean	0.029630	0	.005899	0.3	310524	0.0257	20
236.72158 std	5 0.169584	0.	.076587	0.4	462763	0.1583	16
44.590334							
min 107.00000	0.000000 0	0.	.000000	0.0	900000	0.0000	000
25%	0.000000	0	.000000	0.0	900000	0.0000	00
206.00000 50%	0.000000	0 .	.000000	0.0	900000	0.0000	00
234.00000 75%		0	000000	1 (20000	0.0000	100
263.00000	0.000000 0	0.	. 000000	1.0	900000	0.0000	00

max 696.00	1.000000	1.000	000 1.00	0000 1.00	0000		
	sysBP	diaBP	BMI	heartRate	glucose		
count	4238.000000	4238.000000	4219.000000	4237.000000	3850.000000		
mean	132.352407	82.893464	25.802008	75.878924	81.966753		
std	22.038097	11.910850	4.080111	12.026596	23.959998		
min	83.500000	48.000000	15.540000	44.000000	40.000000		
25%	117.000000	75.000000	23.070000	68.000000	71.000000		
50%	128.000000	82.000000	25.400000	75.000000	78.000000		
75%	144.000000	89.875000	28.040000	83.000000	87.000000		
max	295.000000	142.500000	56.800000	143.000000	394.000000		
count mean std min 25% 50% 75% max df.inf <class< td=""><td>TenYearCHD 4238.000000 0.151958 0.359023 0.000000 0.000000 0.000000 1.000000 o()</td><td>frame.DataFr</td><td>ame'></td><td></td><td></td></class<>	TenYearCHD 4238.000000 0.151958 0.359023 0.000000 0.000000 0.000000 1.000000 o()	frame.DataFr	ame'>				
RangeIndex: 4238 entries, 0 to 4237 Data columns (total 16 columns):							
0 m 1 a 2 e 3 c 4 c 5 B 6 p 7 p 8 d 9 t 10 s	olumn ale ge ducation urrentSmoker igsPerDay PMeds revalentStroke revalentHyp iabetes otChol ysBP iaBP	Non-Null Control	ull int64 ull int64 ull float64 ull int64 ull float64 ull float64 ull int64 ull int64 ull int64 ull int64				

```
12
     BMI
                       4219 non-null
                                        float64
 13
     heartRate
                       4237 non-null
                                        float64
14
     glucose
                       3850 non-null
                                        float64
15
    TenYearCHD
                       4238 non-null
                                        int64
dtypes: float64(9), int64(7)
memory usage: 529.9 KB
df.isna().sum()
                      0
male
age
                      0
                    105
education
currentSmoker
                      0
                     29
cigsPerDay
BPMeds
                     53
prevalentStroke
                      0
                      0
prevalentHyp
                      0
diabetes
totChol
                     50
sysBP
                      0
diaBP
                      0
BMI
                     19
heartRate
                      1
                    388
glucose
TenYearCHD
                      0
dtype: int64
#Since, only a few rows have null values in them, we are only removing
those rows f
#df =
df.dropna(subset=['heartRate', 'BMI', 'cigsPerDay', 'totChol', 'BPMeds'])
df
                                                          BPMeds \
                  education currentSmoker
                                             cigsPerDay
      male
            age
0
         1
             39
                        4.0
                                          0
                                                     0.0
                                                              0.0
1
                        2.0
                                          0
                                                     0.0
                                                              0.0
         0
             46
2
             48
                                                    20.0
         1
                        1.0
                                          1
                                                              0.0
3
             61
                        3.0
                                          1
                                                    30.0
                                                              0.0
         0
4
             46
                                                    23.0
         0
                        3.0
                                          1
                                                              0.0
             50
                                                     1.0
                                                              0.0
4233
         1
                        1.0
                                          1
4234
             51
                                          1
                                                    43.0
         1
                        3.0
                                                              0.0
4235
             48
                        2.0
                                          1
                                                    20.0
                                                             NaN
4236
             44
                                          1
                                                    15.0
         0
                        1.0
                                                              0.0
4237
             52
                        2.0
                                          0
                                                     0.0
                                                             0.0
         0
      prevalentStroke prevalentHyp
                                       diabetes totChol
                                                           sysBP
                                                                   diaBP
BMI
                     0
                                                    195.0
                                                                    70.0
0
                                    0
                                                           106.0
```

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		ð	U	U	243.0	127.5	80.0
3		0	1	0	225.0	150.0	95.0
28.58						100.0	
4 23.10		0	0	0	285.0	130.0	84.0
23.10							
4233		Θ	1	0	313.0	179.0	92.0
25.97 4234		0	0	0	207.0	126 E	00.0
4234 19.71		ט	0	U	207.0	126.5	80.0
4235		Θ	0	0	248.0	131.0	72.0
22.00							
4236		0	0	0	210.0	126.5	87.0
19.16 4237		0	Θ	0	269.0	133.5	83.0
21.47		Ū	Ŭ		20310	133.3	05.0
		_					
0	heartRate 80.0	glucose 77.0	TenYearCHD				
0 1	95.0	77.0	0 0				
1 2 3 4	75.0	70.0	Ö				
3	65.0	103.0	1				
-	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]					
		_					

MISSING VALUE TREATMENT

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

df['education'].fillna(value = df['education'].mean(),inplace=True)

df['heartRate'].fillna(value = df['heartRate'].mean(),inplace=True)

df['BMI'].fillna(value = df['BMI'].mean(),inplace=True)

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

```
df['cigsPerDay'].fillna(value = df['cigsPerDay'].mean(),inplace=True)
df['BPMeds'].fillna(value = df['BPMeds'].mean(),inplace=True)
df.isna().sum()
male
                     0
                     0
age
education
                     0
currentSmoker
                     0
                     0
cigsPerDay
BPMeds
                     0
prevalentStroke
                     0
prevalentHyp
                     0
diabetes
                     0
totChol
                    50
sysBP
                     0
diaBP
                     0
                     0
BMI
heartRate
                     0
                     0
alucose
TenYearCHD
                     0
dtype: int64
#Splitting the dependent and independent variables.
x = df.drop("TenYearCHD",axis=1)
y = df['TenYearCHD']
Х
      male
            age
                  education
                             currentSmoker
                                              cigsPerDay
                                                           BPMeds \
0
             39
                        4.0
                                                     0.0
                                                          0.00000
         1
                                          0
1
         0
             46
                        2.0
                                          0
                                                     0.0
                                                          0.00000
2
         1
             48
                        1.0
                                          1
                                                    20.0
                                                          0.00000
3
                                          1
                                                    30.0
         0
             61
                        3.0
                                                          0.00000
4
                                                    23.0
         0
             46
                        3.0
                                          1
                                                          0.00000
             . . .
4233
             50
                        1.0
                                          1
                                                     1.0
                                                          0.00000
         1
4234
                                                    43.0
         1
             51
                        3.0
                                          1
                                                          0.00000
4235
         0
             48
                        2.0
                                          1
                                                    20.0
                                                          0.02963
4236
             44
                        1.0
                                          1
                                                    15.0
                                                          0.00000
         0
4237
             52
                                          0
         0
                        2.0
                                                     0.0 \quad 0.00000
      prevalentStroke prevalentHyp diabetes totChol sysBP
                                                                   diaBP
BMI \
                     0
                                    0
                                                    195.0
                                                           106.0
                                                                    70.0
26.97
                     0
1
                                    0
                                                    250.0
                                                           121.0
                                                                    81.0
28.73
                                                                    80.0
                     0
                                                    245.0 127.5
25.34
```

3	0	1	0	225.0	150.0	95.0
28.58 4	Θ	0	Θ	285.0	130.0	84.0
23.10						
• • •						
4233	Θ	1	0	313.0	179.0	92.0
25.97						00
4234	0	0	0	207.0	126.5	80.0
19.71 4235	0	0	0	248.0	131.0	72.0
22.00	ŭ	Ü	· ·	21010	13110	7210
4236	0	0	0	210.0	126.5	87.0
19.16 4237	Θ	0	Θ	269.0	133.5	83.0
21.47	U	U	U	209.0	133.3	03.0
heartRate 0 80.0	9					
0 80.0 1 95.0 2 75.0 3 65.0 4 85.0						
2 75.0	70.000000					
3 65.0						
4233 66.0	86.000000					
4234 65.0	68.000000					
4235 84.0						
4236 86.0 4237 80.0						
	107.000000					
[4238 rows x 15						

Train Test Split

```
x_train,x_test,y_train,y_test =
train_test_split(x,y,test_size=0.2,random_state=42)
y_train
3252
          0
3946
          0
1261
          0
          0
2536
4089
          0
3444
          0
466
          0
3092
```

```
3772
        0
860
        0
Name: TenYearCHD, Length: 3390, dtype: int64
from sklearn.svm import SVC
from sklearn.metrics import accuracy score
x test = x test.dropna()
y_test = y_test.loc[x_test.index] # Ensure the target is aligned with
x_test after
x test = x test.dropna()
y test = y test.loc[x test.index] # Ensure the target is aligned with
x_test after
from sklearn.impute import SimpleImputer
imputer = SimpleImputer(strategy='mean') # You can also use 'median',
'most freque
x test = imputer.fit transform(x test)
from sklearn.ensemble import HistGradientBoostingClassifier
classifier = HistGradientBoostingClassifier()
classifier.fit(x train, y train)
acc = classifier.score(x test, y test)
print(acc)
ImportError
                                          Traceback (most recent call
last)
~\AppData\Local\Temp/ipykernel 14856/3200331351.py in <module>
----> 1 from sklearn.ensemble import HistGradientBoostingClassifier
      2 classifier = HistGradientBoostingClassifier()
      3 classifier.fit(x train, y train)
      4 acc = classifier.score(x test, y test)
     5 print(acc)
ImportError: cannot import name 'HistGradientBoostingClassifier' from
'sklearn.ensemble' (C:\Users\HP\anaconda3\lib\site-packages\sklearn\
ensemble\ init .py)
```