In [1]:	-				nbors Algorith	nm i.e. KNN Clas	sifier.										
	#Roll no #Section #Year:3r	o.: 55 n:B	Kirandas Satņ 23	oute													
In [2]:	<pre>import o import r import s from skl import w</pre>	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 varnings.filterwarnings('ignore')															
In [3]:	os.getcw																
Out[3]: In [4]:			ers\\HP\\Desk	<top')< th=""><th></th><th></th><th></th><th></th><th></th><th></th><th></th><th></th><th></th><th></th><th></th><th></th><th></th></top')<>													
In [5]:	df=pd.re	ead_csv('	framingham.cs	SV')													
In [6]:	df.head(		ion currentSmo	ker cinsPerl	Day RPMeds	prevalentStroke	nrevalentHvn	diahetes tot(	thal sysBP r	liaRP RMI	heartRate n	lucose TenVe	earCHD				
Out[6]:	0 1 1 0 2 1 3 0 4 0	39 46 48 61	4.0 2.0 1.0 3.0 3.0	0 0 1 2 1 3	0.0 0.0 0.0 0.0 20.0 0.0 80.0 0.0 23.0 0.0	0 0	0 0 0 1	0 19 0 29 0 24 0 22	95.0 106.0 50.0 121.0 45.0 127.5 25.0 150.0	70.0 26.97 81.0 28.73 80.0 25.34 95.0 28.58 84.0 23.10	80.0 95.0 75.0 65.0 85.0	77.0 76.0 70.0 103.0 85.0	0 0 0 0 1				
In [7]:	df.tail(	()															
Out[7]:	4235 0 4236 0 4237 0 4238 1	e age edu	2.0 1.0 2.0 3.0 3.0	1 1 0 0 1	20.0 I 15.0 0.0 0.0	eds prevalentStr  NaN  0.0  0.0  0.0  0.0  0.0	oke prevalenti  0  0  0  0  0  0	0 0 0 0 0 0	248.0 131.0 210.0 126.5 269.0 133.5 185.0 141.0	72.0 22.0 5 87.0 19.1 5 83.0 21.4 0 98.0 25.6	84.0 86.0 7 80.0 60 67.0	86.0 NaN 107.0 72.0	0 0 0 0 0				
In [8]: Out[8]:	df.info   df.info  0 1 2 3		aFrame.info ( 4.0 2.0 1.0 3.0	of m	0 0 1	education cu 0.0 0.0 0.0 0.0 20.0 0.0 30.0 0.0	rrentSmoker	cigsPerDay	BPMeds \								
	4235 4236 4237 4238 4239	0 46  0 48 0 44 0 52 1 40 0 39 evalentSt	3.0  2.0 1.0 2.0 3.0 3.0		1 1 1 0 0 0 1 1 abetes to 0 0 0 0	23.0 0.0 20.0 NaN 15.0 0.0 0.0 0.0 30.0 0.0  tChol sysBP 195.0 106.0 250.0 121.0 245.0 127.5 225.0 150.0 285.0 130.0	diaBP BN 70.0 26.9 81.0 28.7 80.0 25.3 95.0 28.5 84.0 23.1	73 34 58									
	4235 4236 4237 4238 4239	artRate 80.0	0 0 0 0 0 0 glucose Ten	 0 0 1 0 YearCHD	0 0 0	248.0 131.0 210.0 126.5 269.0 133.5 185.0 141.0 196.0 133.0	72.0 22.6 87.0 19.1 83.0 21.4 98.0 25.6 86.0 20.9	00 L6 17 60									
	1 2 3 4  4235 4236 4237 4238 4239 [4240 row	95.0 75.0 65.0 85.0  84.0 86.0 80.0 67.0 85.0	76.0 70.0 103.0 85.0  86.0 NaN 107.0 72.0 80.0	0 0 1 0  0 0 0													
In [9]: Out[9]:	mean std	male 0.000000 4 0.429245 0.495027	4240.000000 413 49.580189 8.572942	1.979444 1.019791	4240.000000 0.494104 0.500024	cigsPerDay  4211.000000 41  9.005937  11.922462	87.000000 0.029615 0.169544	0.005896 0.076569	0.310613 0.462799	0.025708 2 0.158280	44.591284	132.354599 22.033300	82.897759 11.910394	4221.000000 4239 25.800801 75 4.079840 12	0.000000 3852. 6.878981 81. 2.025348 23.	.963655	0.151887 0.358953
In [10]:	25% 50% 75%	0.000000 0.000000 0.000000 1.000000	42.000000 49.000000 56.000000	1.000000 1.000000 2.000000 3.000000 4.000000	0.000000 0.000000 0.000000 1.000000	0.000000 0.000000 0.000000 20.000000 70.000000	0.000000 0.000000 0.000000 0.000000 1.000000	0.000000 0.000000 0.000000 0.000000 1.000000	0.000000 0.000000 0.000000 1.000000	0.000000 2 0.000000 2 0.000000 2	234.000000 263.000000	83.500000 117.000000 128.000000 144.000000 295.000000	48.000000 75.000000 82.000000 90.000000 142.500000	23.070000 68 25.400000 75 28.040000 83	3.000000 71. 5.000000 78. 3.000000 87.	.000000 .000000 .000000	0.000000 0.000000 0.000000 1.000000
Out[10]:	male age education currentSr cigsPerDa BPMeds prevalent prevalent diabetes totChol sysBP diaBP BMI heartRate glucose	n moker ay tStroke tHyp	0 0 105 0 29 53 0 0 0 0 50 0														
In [11]:	TenYearCH dtype: in	nt64	388 0	1551.3			,										
In [12]:						),inplace= <b>Tru</b> e ean(),inplace:											
In [13]:						ean(),inplace											
In [14]:	df['BMI'	<mark>'</mark> ].fillna	(value = df[	<mark>'BMI</mark> '].mea	n(),inplac	e=True)											
In [15]:	df['cigs	<pre>df['BMI'].fillna(value = df['BMI'].mean(),inplace=True)  df['cigsPerDay'].fillna(value = df['cigsPerDay'].mean(),inplace=True)</pre>															
In [16]:		<pre>df['totChol'].fillna(value = df['totChol'].mean(),inplace=True)</pre>															
In [17]: In [18]:			lna(value = c	df['BPMeds	'].mean(),	inplace= <b>True</b> )											
Out[18]:	male age education currentSr cigsPerDa BPMeds prevalent diabetes totChol sysBP	n moker ay tStroke tHyp	0 0 0 0 0 0 0														
In [19]:	diaBP BMI heartRate glucose TenYearCH dtype: in	HD nt64	0 0 0 0														
Out[19]:	male age education currentSr cigsPerDa BPMeds prevalent diabetes totChol sysBP diaBP BMI heartRate glucose TenYearCh dtype: in	moker ay tStroke tHyp e	0 0 0 0 0 0 0 0 0 0														
In [20]:	#Splitti x = df.c	ing the d	ependent and YearCHD",axis		nt variabl	es.											
In [21]:		king the															
Out[21]:	male 0 1		ucation currentS	Smoker cigsl	PerDay BPN 0.0 0.000	leds prevalentSt	oke prevalent	Hyp diabetes 0 0			MI heartRate						
	1 0 2 1		2.0 1.0	0	0.0 0.000 20.0 0.000	0000	0	0 0	250.0 121. 245.0 127.	0 81.0 28.7 5 80.0 25.3	73 95.0 34 75.0	76.000000 70.000000					
	<b>4</b> 0		3.0 3.0 	1 1 	30.0 0.000 23.0 0.000		0 0	1 0 0 0 	285.0 130. 	0 84.0 23.2	85.0						
	<b>4237</b> 0	<ul><li>44</li><li>52</li><li>40</li><li>39</li></ul>	2.0 1.0 2.0 3.0 3.0	1 1 0 0	20.0 0.029 15.0 0.000 0.0 0.000 30.0 0.000	0000	0 0 0 0	0 0	248.0 131. 210.0 126. 269.0 133. 185.0 141. 196.0 133.	5 87.0 19.2 5 83.0 21.4 0 98.0 25.6	16 86.0 47 80.0 60 67.0	81.963655 0 107.000000 0 72.000000					
	Train	Test	Split														
In [22]:			•	t = train_	test_split	(x,y,test_size	e=0.2, random	n_state=42)									
In [23]:	y_train	0															
Out[23]:	3257 3822 1263 3575 3444 466 3092 3772	0 0 0 0 0 0															
In [24]:	KNN from skl	Class	ghbors <b>import</b>	t KNeighbo	rsClassifi												
In [ ]:	knn = KN knn.fit(	Neighbors (x_train, nn.score( cc)	Classifier(n_ y_train) x_test,y_test	_neighbors		etric='minkows	ski')										