import necessary libraries

n [1]:			numpy a						
[2]:	df=	pd.	read_csv	'("I:\AI	ML\Featur	eSele	ection.cs	/")	
[3]:	df								
t[3]:		age	weight	height	cholestrol	sugar	Target		
,	0	35	70	150	233	250	1		
	1	56	75	100	250	300	0		
	2	67	68	180	204	260	0		
	3	72	60	170	236	450	1		
	4	39	77	190	354	220	1		
	5	48	68	110	192	360	1		
	6	55	62	200	294	190	0		
	7	42	59	175	263	350	0		
	8	30	58	198	199	280	1		
	9	41	57	100	168	100	1		
4]:	df.	sha	ре						
4]:	(10	, 6)						
6]:	df.	des	cribe()						
[6]:			age	weig	jht he	eight	cholestrol	sugar	Target
	cou	nt	10.000000	10.0000	000 10.00	0000	10.00000	10.000000	10.000000
	me	an	48.500000	65.4000	000 157.30	0000	239.30000	276.000000	0.600000
	s	td	13.769935	7.2141	84 39.99	4583	54.83926	97.661547	0.516398
	m	in	30.000000	57.0000	000 100.00	0000	168.00000	100.000000	0.000000
	25	5%	39.500000	59.2500	000 120.00	0000	200.25000	227.500000	0.000000
	50)%	45.000000	65.0000	000 172.50	0000	234.50000	270.000000	1.000000
	75	%	55.750000	69.5000	000 187.50	0000	259.75000	337.500000	1.000000
		21/	72 000000	77 0000	000 200 00	0000	354 00000	450.000000	1.000000

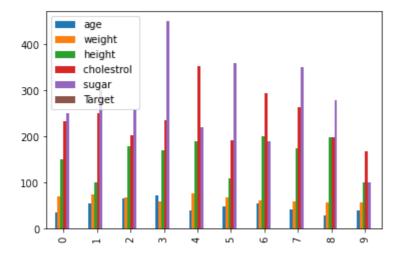
feature scaling using min max method

new value=(old value-min)/(max-min)(new_maxnew_min)+new_min

new_max =1, new_min=0

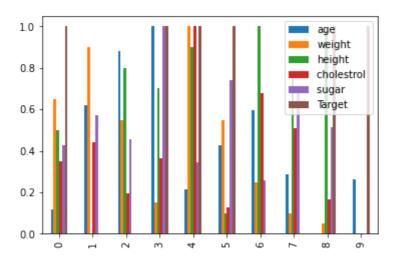
new_value=(old_value - min of the column)/(comlum_max - column_min)(1-0)+0

```
df1=df.copy()
 In [7]:
           df1
 In [8]:
 Out[8]:
                   weight height cholestrol sugar
                                                    Target
              age
           0
               35
                        70
                              150
                                         233
                                                250
                                                          1
           1
                              100
                                         250
                                                300
                                                          0
               56
                        75
           2
                                                          0
               67
                        68
                              180
                                         204
                                                260
           3
               72
                        60
                              170
                                         236
                                                450
                                                          1
           4
               39
                        77
                              190
                                         354
                                                220
                                                          1
                                         192
                                                360
           5
               48
                        68
                              110
                                                          1
           6
               55
                        62
                              200
                                         294
                                                190
                                                          0
                              175
           7
               42
                        59
                                         263
                                                350
                                                          0
           8
               30
                        58
                              198
                                         199
                                                280
                                                          1
               41
                        57
                              100
                                         168
                                                100
                                                          1
In [12]:
           for column in df1.columns:
                df1[column]=(df1[column]-df1[column].min())/(df1[column].max()-df1[column].min()
In [13]:
           df1.head()
Out[13]:
                        weight height cholestrol
                  age
                                                     sugar Target
           0 0.119048
                          0.65
                                   0.5
                                         0.349462 0.428571
                                                                1.0
           1 0.619048
                          0.90
                                   0.0
                                         0.440860 0.571429
                                                                0.0
           2 0.880952
                          0.55
                                   8.0
                                         0.193548 0.457143
                                                                0.0
             1.000000
                          0.15
                                   0.7
                                         0.365591 1.000000
                                                                1.0
           4 0.214286
                          1.00
                                   0.9
                                         1.000000 0.342857
                                                                1.0
           import matplotlib.pyplot as plt
In [14]:
           df.plot(kind='bar')
In [15]:
           <AxesSubplot:>
Out[15]:
```



In [16]: df1.plot(kind='bar')

Out[16]: <AxesSubplot:>



z score normalization

new_value=(old_value - mean of the column) / (standard deviation of column)

Out[19]:		age	weight	height	cholestrol	sugar	Target		
	0	-0.980397	0.637633	-0.182525	-0.114881	-0.266226	0.774597		
	1	0.544665	1.330712	-1.432694	0.195116	0.245747	-1.161895		
	2	1.343507	0.360401	0.567577	-0.643699	-0.163831	-1.161895		
	3	1.706617	-0.748525	0.317543	-0.060176	1.781663	0.774597		
	4	-0.689909	1.607944	0.817611	2.091567	-0.573409	0.774597		
	5	-0.036311	0.360401	-1.182660	-0.862521	0.860113	0.774597		
	6	0.472043	-0.471294	1.067645	0.997461	-0.880592	-1.161895		
	7	-0.472043	-0.887141	0.442560	0.432172	0.757719	-1.161895		
	8	-1.343507	-1.025757	1.017638	-0.734875	0.040958	0.774597		
	9	-0.544665	-1.164373	-1.432694	-1.300163	-1.802142	0.774597		

PCA

In [20]:	<pre>import pandas as pd</pre>														
In [21]:	<pre>df=pd.read_csv("I:\AIML\heart.csv")</pre>														
In [22]:	df														
Out[22]:		age	gender	ср	trtbps	chol	fbs	restecg	thalachh	exng	oldpeak	slp	caa	thall	outpu
	0	63	1	3	145	233	1	0	150	0	2.3	0	0	1	
	1	37	1	2	130	250	0	1	187	0	3.5	0	0	2	
	2	41	0	1	130	204	0	0	172	0	1.4	2	0	2	
	3	56	1	1	120	236	0	1	178	0	0.8	2	0	2	
	4	57	0	0	120	354	0	1	163	1	0.6	2	0	2	
	•••														
	298	57	0	0	140	241	0	1	123	1	0.2	1	0	3	
	299	45	1	3	110	264	0	1	132	0	1.2	1	0	3	
	300	68	1	0	144	193	1	1	141	0	3.4	1	2	3	
	301	57	1	0	130	131	0	1	115	1	1.2	1	1	3	
	302	57	0	1	130	236	0	0	174	0	0.0	1	1	2	
	303 r	ows ×	: 14 colu	mns	;										
4															•
In [24]:	df.s	hape													
Out[24]:	(303	, 14))												
In [25]:	inpu	its =	df.drop	0('0	utput',	axis	col"	.umns')							

```
target =df['output']
In [26]:
           inputs.head()
In [27]:
                            cp trtbps chol fbs restecg thalachh exng oldpeak slp
Out[27]:
                                                                                         caa
           0
                             3
                                                                150
                                                                                           0
               63
                         1
                                  145
                                        233
                                               1
                                                        0
                                                                        0
                                                                                2.3
                                                                                      0
                                                                                                 1
               37
                                  130
                                        250
                                               0
                                                                187
                                                                                3.5
                                                                                           0
                                                                                                 2
           2
               41
                         0
                             1
                                  130
                                        204
                                               0
                                                        0
                                                                172
                                                                        0
                                                                                1.4
                                                                                      2
                                                                                           0
                                                                                                 2
           3
               56
                                   120
                                        236
                                               0
                                                                178
                                                                                0.8
                                                                                           0
                                                                                                 2
               57
                         0
                             0
                                                        1
                                                                                      2
                                                                                           0
                                                                                                 2
                                  120
                                        354
                                               0
                                                                163
                                                                        1
                                                                                0.6
           target.head()
In [28]:
Out[28]:
           1
                1
           2
                1
           3
                 1
           4
           Name: output, dtype: int64
           x=inputs
In [29]:
           y=target
```

scale the input data

```
In [30]:
         from sklearn.preprocessing import StandardScaler
         scaler=StandardScaler()
         x_scaled=scaler.fit_transform(x)
In [31]:
         x_scaled
In [32]:
         array([[ 0.9521966 , 0.68100522, 1.97312292, ..., -2.27457861,
Out[32]:
                 -0.71442887, -2.14887271],
                [-1.91531289, 0.68100522, 1.00257707, ..., -2.27457861,
                 -0.71442887, -0.51292188],
                [-1.47415758, -1.46841752, 0.03203122, ..., 0.97635214,
                 -0.71442887, -0.51292188],
                [ 1.50364073, 0.68100522, -0.93851463, ..., -0.64911323, ]
                  1.24459328, 1.12302895],
                [0.29046364, 0.68100522, -0.93851463, ..., -0.64911323,
                  0.26508221, 1.12302895],
                [0.29046364, -1.46841752, 0.03203122, ..., -0.64911323,
                  0.26508221, -0.51292188]])
         from sklearn.model_selection import train_test_split
         x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.2)
In [34]:
```

Apply any classification method without PCA: logistic regression

In [36]: from sklearn.linear_model import LogisticRegression
 model=LogisticRegression(max_iter=1000)
 model.fit(x_train,y_train)

Out[36]: LogisticRegression(max_iter=1000)

In [37]: model.score(x_test,y_test)

Out[37]: 0.8688524590163934

use PCA to reduce dimensions

In [38]:	х													
Out[38]:		age	gender	ср	trtbps	chol	fbs	restecg	thalachh	exng	oldpeak	slp	caa	thall
	0	63	1	3	145	233	1	0	150	0	2.3	0	0	1
	1	37	1	2	130	250	0	1	187	0	3.5	0	0	2
	2	41	0	1	130	204	0	0	172	0	1.4	2	0	2
	3	56	1	1	120	236	0	1	178	0	0.8	2	0	2
	4	57	0	0	120	354	0	1	163	1	0.6	2	0	2
	•••									•••			•••	
	298	57	0	0	140	241	0	1	123	1	0.2	1	0	3
	299	45	1	3	110	264	0	1	132	0	1.2	1	0	3
	300	68	1	0	144	193	1	1	141	0	3.4	1	2	3
	301	57	1	0	130	131	0	1	115	1	1.2	1	1	3
	302	57	0	1	130	236	0	0	174	0	0.0	1	1	2

303 rows × 13 columns

import libraries for PCA

In [39]: from sklearn.decomposition import PCA

use components such that 99% similarities retrived

In [40]: pca=PCA(0.99)

```
In [41]:
          x_pca=pca.fit_transform(x)
          x.shape
In [42]:
          (303, 13)
Out[42]:
In [43]:
          x_pca.shape
          (303, 4)
Out[43]:
          pca.explained_variance_ratio_
In [44]:
         array([0.7475642 , 0.15037022, 0.08459685, 0.01621596])
Out[44]:
In [45]:
          x_pca
         array([[ -12.26734484,
                                    2.87383781,
                                                  14.96987876,
                                                                   6.8929401],
Out[45]:
                                  -39.87137362,
                                                                 -10.58359849],
                     2.69013712,
                                                   0.8778823 ,
                 [ -42.95021407, -23.63681988,
                                                   1.75944589,
                                                                  -7.7866551 ],
                 [ -51.96381148,
                                   13.32379836,
                                                  15.48684358,
                                                                 11.6374553 ],
                 [-114.75598084,
                                                  0.12777095,
                                                                   0.51138633],
                                   36.43518423,
                 [ -10.39614198, -23.30240081,
                                                   2.39130354,
                                                                   7.18326419]])
```

how many PC created

```
In [46]: pca.n_components_
Out[46]: 4

In [47]: x_train_pca,x_test_pca,y_train,y_test=train_test_split(x_pca,y,test_size=0.2)

In [48]: from sklearn.linear_model import LogisticRegression model=LogisticRegression(max_iter=1000) model.fit(x_train_pca,y_train)

Out[48]: LogisticRegression(max_iter=1000)

In [49]: model.score(x_test_pca,y_test)

Out[49]: 0.5901639344262295
```

let's cosider 7 componets

```
In [50]: pca=PCA(n_components=7)
In [51]: x_pca=pca.fit_transform(x)
In [52]: x_pca.shape
Out[52]: (303, 7)
In [53]: x_train_pca,x_test_pca,y_train,y_test=train_test_split(x_pca,y,test_size=0.2)
```

```
In [54]: from sklearn.linear_model import LogisticRegression
    model=LogisticRegression(max_iter=1000)
    model.fit(x_train_pca,y_train)

Out[54]: LogisticRegression(max_iter=1000)

In [55]: model.score(x_test_pca,y_test)

Out[55]: 0.7049180327868853

In []:
```