

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
from sklearn.datasets import load_breast_cancer
```

```
df=pd.read_csv('/content/2.2 dataset breast cancer.csv')
print(df)
```

```

id diagnosis radius_mean texture_mean perimeter_mean area_mean \
0 842302 M 17.99 10.38 122.80 1001.0
1 842517 M 20.57 17.77 132.90 1326.0
2 84300903 M 19.69 21.25 130.00 1203.0
3 84348301 M 11.42 20.38 77.58 386.1
4 84358402 M 20.29 14.34 135.10 1297.0
.. ... ..
564 926424 M 21.56 22.39 142.00 1479.0
565 926682 M 20.13 28.25 131.20 1261.0
566 926954 M 16.60 28.08 108.30 858.1
567 927241 M 20.60 29.33 140.10 1265.0
568 92751 B 7.76 24.54 47.92 181.0

smoothness_mean compactness_mean concavity_mean concave points_mean \
0 0.11840 0.27760 0.30010 0.14710
1 0.08474 0.07864 0.08690 0.07017
2 0.10960 0.15990 0.19740 0.12790
3 0.14250 0.28390 0.24140 0.10520
4 0.10030 0.13280 0.19800 0.10430
.. ... ..
564 0.11100 0.11590 0.24390 0.13890
565 0.09780 0.10340 0.14400 0.09791
566 0.08455 0.10230 0.09251 0.05302
567 0.11780 0.27700 0.35140 0.15200
568 0.05263 0.04362 0.00000 0.00000

... texture_worst perimeter_worst area_worst smoothness_worst \
0 ... 17.33 184.60 2019.0 0.16220
1 ... 23.41 158.80 1956.0 0.12380
2 ... 25.53 152.50 1709.0 0.14440
3 ... 26.50 98.87 567.7 0.20980
4 ... 16.67 152.20 1575.0 0.13740
.. ... ..
564 ... 26.40 166.10 2027.0 0.14100
565 ... 38.25 155.00 1731.0 0.11660
566 ... 34.12 126.70 1124.0 0.11390
567 ... 39.42 184.60 1821.0 0.16500
568 ... 30.37 59.16 268.6 0.08996

compactness_worst concavity_worst concave points_worst symmetry_worst \
0 0.66560 0.7119 0.2654 0.4601
1 0.18660 0.2416 0.1860 0.2750
2 0.42450 0.4504 0.2430 0.3613
3 0.86630 0.6869 0.2575 0.6638
4 0.20500 0.4000 0.1625 0.2364
.. ... ..
564 0.21130 0.4107 0.2216 0.2060
565 0.19220 0.3215 0.1628 0.2572
566 0.30940 0.3403 0.1418 0.2218
567 0.86810 0.9387 0.2650 0.4087
568 0.06444 0.0000 0.0000 0.2871

fractal_dimension_worst Unnamed: 32
0 0.11890 NaN
1 0.08902 NaN
2 0.08758 NaN
3 0.17300 NaN
4 0.07678 NaN
```

```
breast=load_breast_cancer()
breast_data=breast.data
print(breast_data)
print(breast_data.shape)
```

```
[[1.799e+01 1.038e+01 1.228e+02 ... 2.654e-01 4.601e-01 1.189e-01]
 [2.057e+01 1.777e+01 1.329e+02 ... 1.860e-01 2.750e-01 8.902e-02]
 [1.969e+01 2.125e+01 1.300e+02 ... 2.430e-01 3.613e-01 8.758e-02]
 ...
 [1.660e+01 2.808e+01 1.083e+02 ... 1.418e-01 2.218e-01 7.820e-02]
 [2.060e+01 2.933e+01 1.401e+02 ... 2.650e-01 4.087e-01 1.240e-01]
 [7.760e+00 2.454e+01 4.792e+01 ... 0.000e+00 2.871e-01 7.039e-02]]
(569, 30)
```

[illegible]

(569, 31)

[illegible]

```
[ 'mean radius' 'mean texture' 'mean perimeter' 'mean area'
'mean smoothness' 'mean compactness' 'mean concavity'
'mean concave points' 'mean symmetry' 'mean fractal dimension'
'radius error' 'texture error' 'perimeter error' 'area error'
'smoothness error' 'compactness error' 'concavity error'
'concave points error' 'symmetry error' 'fractal dimension error'
'worst radius' 'worst texture' 'worst perimeter' 'worst area'
'worst smoothness' 'worst compactness' 'worst concavity'
'worst concave points' 'worst symmetry' 'worst fractal dimension']
```

```
breast_dataset.columns=features_labels # embedding column names
breast_dataset.head()
```

	mean radius	mean texture	mean perimeter	mean area	mean smoothness	mean compactness	mean concavity	mean concave points	mean symmetry	mean fractal dimension	...	worst texture	wo perime
0	17.99	10.38	122.80	1001.0	0.11840	0.27760	0.3001	0.14710	0.2419	0.07871	...	17.33	184
1	20.57	17.77	132.90	1326.0	0.08474	0.07864	0.0869	0.07017	0.1812	0.05667	...	23.41	158
2	19.69	21.25	130.00	1203.0	0.10960	0.15990	0.1974	0.12790	0.2069	0.05999	...	25.53	152
3	11.42	20.38	77.58	386.1	0.14250	0.28390	0.2414	0.10520	0.2597	0.09744	...	26.50	98
4	20.29	14.34	135.10	1297.0	0.10030	0.13280	0.1980	0.10430	0.1809	0.05883	...	16.67	152

5 rows × 31 columns

```
#replace the target values
breast_dataset['label'].replace(0, 'Benign', inplace=True)
breast_dataset['label'].replace(1, 'Malignant', inplace=True)
breast_dataset.tail()
```

	mean radius	mean texture	mean perimeter	mean area	mean smoothness	mean compactness	mean concavity	mean concave points	mean symmetry	mean fractal dimension	...	worst texture	peri
564	21.56	22.39	142.00	1479.0	0.11100	0.11590	0.24390	0.13890	0.1726	0.05623	...	26.40	1
565	20.13	28.25	131.20	1261.0	0.09780	0.10340	0.14400	0.09791	0.1752	0.05533	...	38.25	1
566	16.60	28.08	108.30	858.1	0.08455	0.10230	0.09251	0.05302	0.1590	0.05648	...	34.12	1
567	20.60	29.33	140.10	1265.0	0.11780	0.27700	0.35140	0.15200	0.2397	0.07016	...	39.42	1
568	7.76	24.54	47.92	181.0	0.05263	0.04362	0.00000	0.00000	0.1587	0.05884	...	30.37	

5 rows × 31 columns

```
from sklearn.preprocessing import StandardScaler
x=breast_dataset.loc[:,features].values
x=StandardScaler().fit_transform(x) #normalizing the features
print(x.shape)
```

(569, 30)

```
np.mean(x),np.std(x)
```

(-6.118909323768877e-16, 1.0)

```
#normalized features into tabular form
feat_cols=['feature'+str(i) for i in range(x.shape[1])]
normalised_breast=pd.DataFrame(x,columns=feat_cols)
print(normalised_breast)
```

	feature0	feature1	feature2	feature3	feature4	feature5	feature6	\
0	1.097064	-2.073335	1.269934	0.984375	1.568466	3.283515	2.652874	
1	1.829821	-0.353632	1.685955	1.908708	-0.826962	-0.487072	-0.023846	
2	1.579888	0.456187	1.566503	1.558884	0.942210	1.052926	1.363478	
3	-0.768909	0.253732	-0.592687	-0.764464	3.283553	3.402909	1.915897	
4	1.750297	-1.151816	1.776573	1.826229	0.280372	0.539340	1.371011	
..	
564	2.110995	0.721473	2.060786	2.343856	1.041842	0.219060	1.947285	
565	1.704854	2.085134	1.615931	1.723842	0.102458	-0.017833	0.693043	
566	0.702284	2.045574	0.672676	0.577953	-0.840484	-0.038680	0.046588	
567	1.838341	2.336457	1.982524	1.735218	1.525767	3.272144	3.296944	
568	-1.808401	1.221792	-1.814389	-1.347789	-3.112085	-1.150752	-1.114873	
	feature7	feature8	feature9	...	feature20	feature21	feature22	\
0	2.532475	2.217515	2.255747	...	1.886690	-1.359293	2.303601	
1	0.548144	0.001392	-0.868652	...	1.805927	-0.369203	1.535126	
2	2.037231	0.939685	-0.398008	...	1.511870	-0.023974	1.347475	
3	1.451707	2.867383	4.910919	...	-0.281464	0.133984	-0.249939	
4	1.428493	-0.009560	-0.562450	...	1.298575	-1.466770	1.338539	
..	
564	2.320965	-0.312589	-0.931027	...	1.901185	0.117700	1.752563	
565	1.263669	-0.217664	-1.058611	...	1.536720	2.047399	1.421940	
566	0.105777	-0.809117	-0.895587	...	0.561361	1.374854	0.579001	
567	2.658866	2.137194	1.043695	...	1.961239	2.237926	2.303601	
568	-1.261820	-0.820070	-0.561032	...	-1.410893	0.764190	-1.432735	

```

      feature23 feature24 feature25 feature26 feature27 feature28 \
0      2.001237  1.307686  2.616665  2.109526  2.296076  2.750622
1      1.890489 -0.375612 -0.430444 -0.146749  1.087084 -0.243890
2      1.456285  0.527407  1.082932  0.854974  1.955000  1.152255
3     -0.550021  3.394275  3.893397  1.989588  2.175786  6.046041
4      1.220724  0.220556 -0.313395  0.613179  0.729259 -0.868353
..      ...      ...      ...      ...      ...      ...
564    2.015301  0.378365 -0.273318  0.664512  1.629151 -1.360158
565    1.494959 -0.691230 -0.394820  0.236573  0.733827 -0.531855
566    0.427906 -0.809587  0.350735  0.326767  0.414069 -1.104549
567    1.653171  1.430427  3.904848  3.197605  2.289985  1.919083
568   -1.075813 -1.859019 -1.207552 -1.305831 -1.745063 -0.048138

      feature29
0      1.937015
1      0.281190
2      0.201391
3      4.935010
4     -0.397100
..      ...
564   -0.709091
565   -0.973978
566   -0.318409
567    2.219635
568   -0.751207
```

[569 rows x 30 columns]

normalised_breast.tail()

	feature0	feature1	feature2	feature3	feature4	feature5	feature6	feat
564	2.110995	0.721473	2.060786	2.343856	1.041842	0.219060	1.947285	2.32
565	1.704854	2.085134	1.615931	1.723842	0.102458	-0.017833	0.693043	1.26
566	0.702284	2.045574	0.672676	0.577953	-0.840484	-0.038680	0.046588	0.10
567	1.838341	2.336457	1.982524	1.735218	1.525767	3.272144	3.296944	2.65
568	-1.808401	1.221792	-1.814389	-1.347789	-3.112085	-1.150752	-1.114873	-1.26

5 rows x 30 columns

```
#3d to 2d
from sklearn.decomposition import PCA
pca_breast=PCA(n_components=2)
principalComponents_breast=pca_breast.fit_transform(x)
principal_breast_Df=pd.DataFrame(data=principalComponents_breast,columns=['principal component 1', 'principal component 2'])
principal_breast_Df.tail()
```

	principal component 1	principal component 2
564	6.439315	-3.576817
565	3.793382	-3.584048
566	1.256179	-1.902297
567	10.374794	1.672010
568	-5.475243	-0.670637

```
#plot the pca
import matplotlib.pyplot as plt
plt.figure()
plt.figure(figsize=(10,10))
plt.xticks(fontsize=12)
plt.yticks(fontsize=14)
plt.xlabel('Principal Component - 1', fontsize=20)
plt.ylabel('Principal Component - 2', fontsize=20)
plt.title("Principal Component Analysis of Breast Cancer Dataset", fontsize=20)
targets=['Benign','Malignant']
colors=['r','g']
for target, color in zip(targets,colors):
    indicesTokeep=breast_dataset['label']==target
    plt.scatter(principal_breast_Df.loc[indicesTokeep,'principal component 1'],
                principal_breast_Df.loc[indicesTokeep,'principal component 2'],c=color, s=5)
plt.legend(targets,prop('size':15))
```

```
File "<ipython-input-1-41a3853caf16>", line 16
    plt.legend(targets,prop('size':15))
                                ^
```

SyntaxError: invalid syntax

Start coding or [generate](#) with AI.

Start coding or [generate](#) with AI.

Start coding or [generate](#) with AI.

Start coding or [generate](#) with AI.

Start coding or [generate](#) with AI.

Start coding or [generate](#) with AI.

Start coding or [generate](#) with AI.

Start coding or [generate](#) with AI.

Start coding or [generate](#) with AI.

Start coding or [generate](#) with AI.