```
import numpy as np
import pandas as pd
import matplotlib as plt
from sklearn.datasets import load_breast_cancer
df=pd.read_csv("/content/2.2 dataset breast cancer.csv")
   <class 'pandas.core.frame.DataFrame'>
   RangeIndex: 569 entries, 0 to 568
   Data columns (total 33 columns):
       Column
                        Non-Null Count Dtype
   ---
    0
      id
                        569 non-null
                                    int64
    1
       diagnosis
                        569 non-null
                                    object
      radius_mean
                        569 non-null
                                    float64
                        569 non-null
                                    float64
    3
       texture mean
      perimeter mean
                        569 non-null
                                    float64
                        569 non-null
    5
                                    float64
       area mean
                        569 non-null
                                    float64
    6
      smoothness mean
                        569 non-null
                                    float64
       compactness_mean
    8
       concavity_mean
                        569 non-null
                                    float64
    9
       concave points_mean
                        569 non-null
                                    float64
    10 symmetry_mean
                        569 non-null
                                    float64
    11 fractal_dimension_mean
                        569 non-null
                                    float64
                        569 non-null
      radius_se
                                    float64
                        569 non-null
    13 texture_se
                                    float64
    14 perimeter se
                        569 non-null
                                    float64
                        569 non-null
                                    float64
    15 area_se
    16 smoothness se
                        569 non-null
                                    float64
                        569 non-null
                                    float64
    17 compactness se
                        569 non-null
                                    float64
    18 concavity se
    19 concave points_se
                        569 non-null
                                    float64
    20 symmetry_se
                        569 non-null
                                    float64
    21
      fractal_dimension_se
                        569 non-null
                                    float64
                        569 non-null
    22 radius_worst
                                    float64
    23 texture_worst
                        569 non-null
                                    float64
                        569 non-null
    24 perimeter_worst
                                    float64
    25
                        569 non-null
                                    float64
      area worst
    26 smoothness_worst
                        569 non-null
                                    float64
                        569 non-null
                                    float64
    27
      compactness worst
                        569 non-null
                                    float64
    28 concavity worst
    29 concave points_worst
                        569 non-null
                                    float64
    30 symmetry_worst
                        569 non-null
                                    float64
    31 fractal_dimension_worst 569 non-null
                                    float64
    32 Unnamed: 32
                        0 non-null
                                    float64
   dtypes: float64(31), int64(1), object(1)
   memory usage: 146.8+ KB
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)
breast_labels=breast.target
print(breast_labels)
print(breast labels.shape)
   10000000010111110010011110010011110000
    1 0 1 1 1 0 1 1 0 0 1 0 0 0 0 0 1 0 0 0 1 0 1 0 1 1 0 1 0 0 0 0 1 1 0 0 1 1
    1011010111111111111111101111010111100011
    1 1 1 1 1 1 1 0 0 0 0 0 0 1]
```

reshape the dataset by adding the label to it

```
labels=np.reshape(breast_labels,(569,1))
final_breast_data=np.concatenate([breast_data,labels],axis=1)
print(final_breast_data.shape)
     (569, 31)
breast_dataset=pd.DataFrame(final_breast_data)
print(breast_dataset.head())
                 1
    0 17.99 10.38 122.80 1001.0 0.11840 0.27760 0.3001 0.14710 0.2419
       20.57 17.77 132.90 1326.0 0.08474 0.07864
                                                      0.0869
                                                              0.07017
    2 19.69 21.25 130.00 1203.0 0.10960 0.15990 0.1974
                                                             0.12790 0.2069
                     77.58
                             386.1 0.14250 0.28390 0.2414 0.10520 0.2597
    3 11.42 20.38
    4 20.29 14.34 135.10 1297.0 0.10030 0.13280 0.1980 0.10430 0.1809
                                22
            9
                        21
                                        23
                                                24
                                                       25
                                                               26
                                                                       27 \
    0 \quad 0.07871 \quad \dots \quad 17.33 \quad 184.60 \quad 2019.0 \quad 0.1622 \quad 0.6656 \quad 0.7119 \quad 0.2654
       0.05667 ...
                     23.41 158.80 1956.0 0.1238 0.1866 0.2416
                                                                   0.1860
       0.05999 \ \dots \ 25.53 \ 152.50 \ 1709.0 \ 0.1444 \ 0.4245 \ 0.4504 \ 0.2430
    3 0.09744 ... 26.50
                            98.87 567.7 0.2098 0.8663 0.6869 0.2575
    4 0.05883 ... 16.67 152.20 1575.0 0.1374 0.2050 0.4000 0.1625
           28
                    29
                        30
    0 0.4601 0.11890 0.0
    1 0.2750 0.08902 0.0
     2 0.3613 0.08758 0.0
    3 0.6638 0.17300 0.0
    4 0.2364 0.07678 0.0
     [5 rows x 31 columns]
```

print the features that are there in the breast cancer dataset

'worst concave points' 'worst symmetry' 'worst fractal dimension']

```
features = breast.feature_names
print(features)

['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'
```

here the labl feild is missing so add to it

```
features_labels=np.append(features,'label')
```

Embedding the column names to the dataframe

```
breast_dataset.columns=features_labels
breast_dataset.head()
```

	mean radius	mean texture	mean perimeter	mean area	mean smoothness	mean compactness	mean concavity	mean concave points	sy
0	17.99	10.38	122.80	1001.0	0.11840	0.27760	0.3001	0.14710	
1	20.57	17.77	132.90	1326.0	0.08474	0.07864	0.0869	0.07017	
2	19.69	21.25	130.00	1203.0	0.10960	0.15990	0.1974	0.12790	
3	11.42	20.38	77.58	386.1	0.14250	0.28390	0.2414	0.10520	
4	20.29	14.34	135.10	1297.0	0.10030	0.13280	0.1980	0.10430	
5 rows × 31 columns									

Replace function

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

```
mean
        mean
                 mean
                             mean
                                     mean
                                                 mean
                                                               mean
                                                                           mean
                                                                                 concave
      radius texture perimeter
                                     area smoothness compactness concavity
                                                                                  points
564
       21.56
                22.39
                           142.00 1479.0
                                               0.11100
                                                            0.11590
                                                                        0.24390
                                                                                 0.13890
                           131.20 1261.0
565
       20.13
                                               0.09780
                                                                        0.14400
                                                                                0.09791
                28.25
                                                            0.10340
566
       16.60
                28.08
                           108.30
                                    858.1
                                               0.08455
                                                            0.10230
                                                                        0.09251
                                                                                 0.05302
567
       20.60
                29.33
                           140.10 1265.0
                                               0.11780
                                                            0.27700
                                                                        0.35140
                                                                                0.15200
        7.76
                            47.92
568
                24.54
                                    181.0
                                               0.05263
                                                            0.04362
                                                                        0.00000 0.00000
5 rows × 31 columns
```

PCA/Normalizing the features

```
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)

```
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
    1.097064 -2.073335 1.269934 0.984375 1.568466 3.283515
    1.829821 -0.353632 1.685955 1.908708 -0.826962 -0.487072
    1.579888 0.456187 1.566503 1.558884 0.942210 1.052926
    -0.768909 0.253732 -0.592687 -0.764464 3.283553
                                                     3.402909
                                                               1.915897
3
    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
    1.704854 2.085134 1.615931 1.723842 0.102458 -0.017833
                                                               0.693043
565
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
                                 . . .
    2.532475 2.217515 2.255747 ...
0
                                       1.886690 -1.359293
                                                             2.303601
    0.548144
              0.001392 -0.868652 ...
                                        1.805927
                                                 -0.369203
                                                             1.535126
1
2
    2.037231
              0.939685 -0.398008 ...
                                        1.511870
                                                 -0.023974
                                                             1.347475
    1.451707
              2.867383 4.910919 ...
                                                  0.133984
                                                            -0.249939
3
                                       -0.281464
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
                                  . . .
                                                  2.047399
    1.263669 -0.217664 -1.058611
                                        1.536720
                                                             1.421940
    0.105777 -0.809117 -0.895587
                                        0.561361
                                                  1.374854
                                                             0.579001
                                  . . .
    2.658866 2.137194 1.043695
                                                   2.237926
                                       1.961239
                                                             2.303601
                                 . . .
   -1.261820 -0.820070 -0.561032 ...
                                       -1.410893
                                                  0.764190
                                                            -1.432735
    feature23 feature24 feature25 feature26 feature27 feature28
0
               1.307686
                                                           2.750622
     2.001237
                          2.616665
                                     2.109526
                                                2.296076
               -0.375612
                          -0.430444
                                                 1.087084
                                                           -0.243890
1
     1.890489
                                     -0.146749
                0.527407
                                                1.955000
2
     1.456285
                          1.082932
                                      0.854974
                                                           1.152255
                3.394275
                           3.893397
3
    -0.550021
                                      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
     1.494959
                         -0.394820
              -0.691230
                                      0.236573
                                                 0.733827
                                                           -0.531855
566
     0.427906
               -0.809587
                           0.350735
                                      0.326767
                                                 0.414069
                                                           -1.104549
     1.653171
               1.430427
                          3.904848
                                      3.197605
                                                2.289985
                                                           1.919083
    -1.075813
              -1.859019 -1.207552 -1.305831 -1.745063
568
                                                           -0.048138
    feature29
0
     1.937015
1
     0.281190
2
     0.201391
3
     4.935010
4
     -0.397100
564
    -0.709091
565
    -0.973978
    -0.318409
```

```
567 2.219635
568 -0.751207
```

[569 rows x 30 columns]

```
normalised_breast.tail()
```

```
feature0 feature1 feature2 feature3 feature4 feature5 feature6 feature7
564 2.110995 0.721473
                         2.060786
                                   2.343856
                                            1.041842
                                                      0.219060 1.947285
                                                                          2.320965
     1.704854 2.085134
                         1.615931
                                  1.723842
                                            0.102458 -0.017833 0.693043
                                                                         1.263669
565
     0.702284 2.045574
                         0.672676
                                  0.577953 -0.840484 -0.038680
                                                                0.046588
                                                                          0.105777
566
567
     1.838341 2.336457
                        1.982524
                                  1.735218
                                            1.525767
                                                     3.272144
                                                               3.296944
                                                                          2.658866
568 -1.808401 1.221792 -1.814389 -1.347789 -3.112085 -1.150752 -1.114873 -1.261820
5 rows × 30 columns
```

```
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 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 Ananlysis of Breast Cancer",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']
plt.legend(targets,prop={'size':15})
plt.show
```

```
matplotlib.pyplot.show
def show(*args, **kwargs)

Display all open figures.

Parameters
------
block : bool, optional
Whether to wait for all figures to be closed before returning.
```

<Figure size 640x480 with 0 Axes>

