

objective: use principal component analysis

```
In [1]: #step1 : import necessary library
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn import datasets
```

```
In [2]: #step2: Load dataset
cancer_data=datasets.load_breast_cancer()
```

```
In [3]: # step3: convert data into dataframe
cancer_df=pd.DataFrame(cancer_data.data,columns=cancer_data.feature_names)
```

```
In [4]: cancer_df.head()
```

Out[4]:

	an re	mean perimeter	mean area	mean smoothness	mean compactness	mean concavity	mean concave points	mean symmetry	mean fractal dimension	...	v ra
38		122.80	1001.0	0.11840	0.27760	0.3001	0.14710	0.2419	0.07871	...	;
77		132.90	1326.0	0.08474	0.07864	0.0869	0.07017	0.1812	0.05667	...	;
25		130.00	1203.0	0.10960	0.15990	0.1974	0.12790	0.2069	0.05999	...	;
38		77.58	386.1	0.14250	0.28390	0.2414	0.10520	0.2597	0.09744	...	;
34		135.10	1297.0	0.10030	0.13280	0.1980	0.10430	0.1809	0.05883	...	;

imns



```
In [5]: cancer_df.info()
```

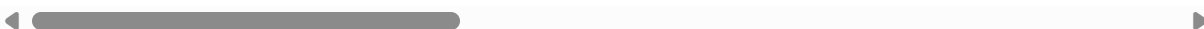
```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 569 entries, 0 to 568
Data columns (total 30 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   mean radius                          569 non-null    float64
1   mean texture                         569 non-null    float64
2   mean perimeter                      569 non-null    float64
3   mean area                          569 non-null    float64
4   mean smoothness                    569 non-null    float64
5   mean compactness                   569 non-null    float64
6   mean concavity                     569 non-null    float64
7   mean concave points                569 non-null    float64
8   mean symmetry                      569 non-null    float64
9   mean fractal dimension             569 non-null    float64
10  radius error                       569 non-null    float64
11  texture error                      569 non-null    float64
12  perimeter error                    569 non-null    float64
13  area error                        569 non-null    float64
14  smoothness error                   569 non-null    float64
15  compactness error                  569 non-null    float64
16  concavity error                    569 non-null    float64
17  concave points error               569 non-null    float64
18  symmetry error                     569 non-null    float64
19  fractal dimension error            569 non-null    float64
20  worst radius                      569 non-null    float64
21  worst texture                      569 non-null    float64
22  worst perimeter                    569 non-null    float64
23  worst area                        569 non-null    float64
24  worst smoothness                   569 non-null    float64
25  worst compactness                  569 non-null    float64
26  worst concavity                     569 non-null    float64
27  worst concave points               569 non-null    float64
28  worst symmetry                      569 non-null    float64
29  worst fractal dimension            569 non-null    float64
dtypes: float64(30)
memory usage: 133.5 KB
```

```
In [6]: # perform statistical analysis
cancer_df.describe()
```

Out[6]:

	mean radius	mean texture	mean perimeter	mean area	mean smoothness	mean compactness	mean concavity
<b>count</b>	569.000000	569.000000	569.000000	569.000000	569.000000	569.000000	569.000000
<b>mean</b>	14.127292	19.289649	91.969033	654.889104	0.096360	0.104341	0.088799
<b>std</b>	3.524049	4.301036	24.298981	351.914129	0.014064	0.052813	0.079720
<b>min</b>	6.981000	9.710000	43.790000	143.500000	0.052630	0.019380	0.000000
<b>25%</b>	11.700000	16.170000	75.170000	420.300000	0.086370	0.064920	0.029560
<b>50%</b>	13.370000	18.840000	86.240000	551.100000	0.095870	0.092630	0.061540
<b>75%</b>	15.780000	21.800000	104.100000	782.700000	0.105300	0.130400	0.130700
<b>max</b>	28.110000	39.280000	188.500000	2501.000000	0.163400	0.345400	0.426800

8 rows × 30 columns



```
In [7]: # standardize the data
from sklearn.preprocessing import StandardScaler
ss=StandardScaler()
ss.fit(cancer_df)
```

Out[7]: StandardScaler()

```
In [8]: data_transform=ss.transform(cancer_df)
```

```
In [10]: data_transform[0]
```

Out[10]: array([ 1.09706398, -2.07333501, 1.26993369, 0.9843749 , 1.56846633,  
3.28351467, 2.65287398, 2.53247522, 2.21751501, 2.25574689,  
2.48973393, -0.56526506, 2.83303087, 2.48757756, -0.21400165,  
1.31686157, 0.72402616, 0.66081994, 1.14875667, 0.90708308,  
1.88668963, -1.35929347, 2.30360062, 2.00123749, 1.30768627,  
2.61666502, 2.10952635, 2.29607613, 2.75062224, 1.93701461])

In [11]: `cancer_df.head()`

Out[11]:

	mean radius	mean texture	mean perimeter	mean area	mean smoothness	mean compactness	mean concavity	mean concave points	mean symmetry
0	17.99	10.38	122.80	1001.0	0.11840	0.27760	0.3001	0.14710	0.2419
1	20.57	17.77	132.90	1326.0	0.08474	0.07864	0.0869	0.07017	0.1812
2	19.69	21.25	130.00	1203.0	0.10960	0.15990	0.1974	0.12790	0.2069
3	11.42	20.38	77.58	386.1	0.14250	0.28390	0.2414	0.10520	0.2597
4	20.29	14.34	135.10	1297.0	0.10030	0.13280	0.1980	0.10430	0.1809

5 rows × 30 columns



In [13]: `data_transform.shape`

Out[13]: (569, 30)

```
#create principal components
from sklearn.decomposition import PCA

#create an instance of PCA
pca=PCA(n_components=2)

# apply
x_pca=pca.fit_transform(data_transform)
```

In [15]: `# result is x_pca`  
`x_pca.shape`

Out[15]: (569, 2)

```
# visualize data - create a dataframe
pca_df=pd.DataFrame(x_pca,columns=['PC1', 'PC2'])
```

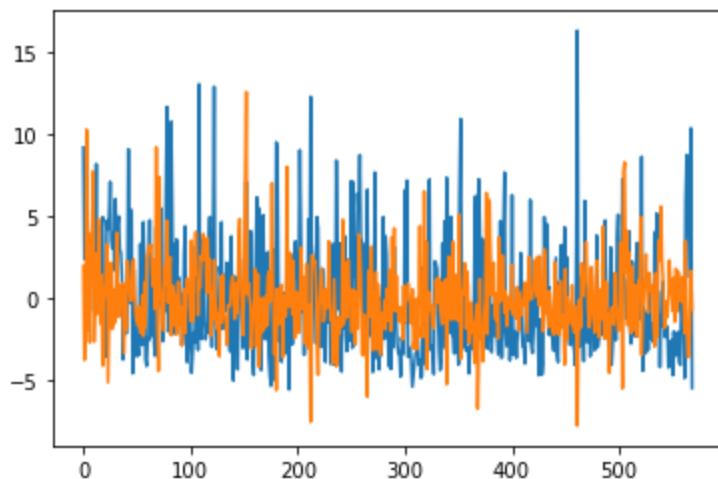
In [19]: `pca_df.head()`

Out[19]:

	PC1	PC2
0	9.192837	1.948583
1	2.387802	-3.768172
2	5.733896	-1.075174
3	7.122953	10.275589
4	3.935302	-1.948072

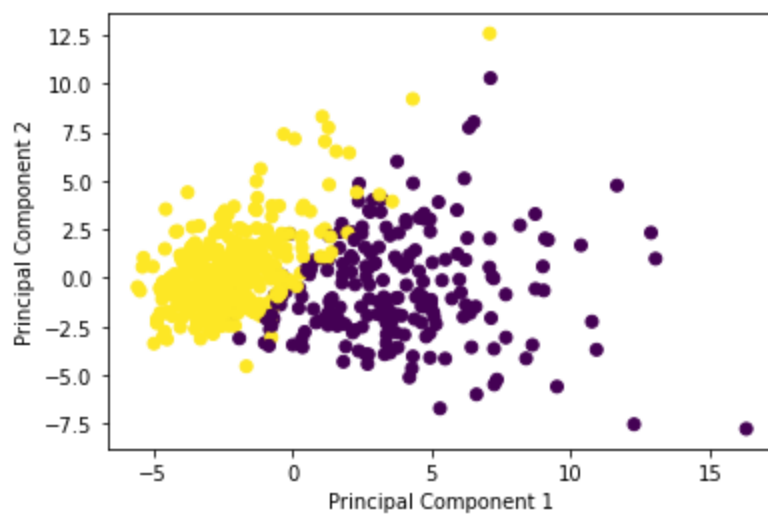
```
In [22]: # draw a scatter plot of principal component  
plt.plot(pca_df)
```

```
Out[22]: [<matplotlib.lines.Line2D at 0x1ccbd74b0a0>,  
<matplotlib.lines.Line2D at 0x1ccbd74b0d0>]
```



```
In [27]: plt.scatter(x='PC1',y='PC2',data=pca_df,c=cancer_data.target)  
plt.xlabel('Principal Component 1')  
plt.ylabel('Principal Component 2')
```

```
Out[27]: Text(0, 0.5, 'Principal Component 2')
```



```
In [ ]:
```