

```
In [1]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
from scipy.spatial.distance import cdist
from sklearn.decomposition import PCA
import matplotlib.pyplot as plt
from sklearn.preprocessing import scale

In [2]: var=pd.read_csv('C://Users/Gopi/Desktop/machine learning/csv files/Cars.csv')
print(var.head(1))

      HP      MPG  VOL      SP      WT
0  49   53.700681   89  104.185353  28.762059

In [3]: var1 = scale(var)
print(var1.shape)

(81, 5)

In [4]: model = PCA (n_components = 3)
print(model)

PCA(copy=True, iterated_power='auto', n_components=3, random_state=None,
    svd_solver='auto', tol=0.0, whiten=False)

In [5]: temp = model.fit_transform(var1)
print(temp)
```

[[-2.6123435 -0.64477881 0.80705114]
[-2.16482189 -0.74286674 0.60638348]
[-2.17886412 -0.72254398 0.59843893]
[-1.52045534 -0.37235871 0.50966693]
[-2.27211186 -0.74850319 0.59738183]
[-1.63473818 -0.22846673 0.442958 ]
[-2.17296003 -0.73108874 0.60177924]
[-3.48353714 1.37690391 -0.57665129]
[-3.45723844 1.33884283 -0.5617725 ]
[-1.11853471 -0.28224631 0.3261837 ]
[-1.62537117 -0.24398604 0.29049378]
[-2.32435717 2.1129633 -0.69631507]
[-0.45425584 -0.24231121 0.39523401]
[-1.62459541 -0.24510877 0.29093267]
[-1.84394062 -0.39462765 0.27424893]
[-1.61297613 -0.2619249 0.29750641]
[-1.32809924 -0.16136988 0.25726692]
[-2.30171443 2.08019336 -0.68350468]
[-1.30787075 -0.19064578 0.26871142]
[-0.45605404 -0.5190537 0.29490354]
[-0.52081831 -0.29308336 0.20382739]
[-0.92427762 -1.10004267 0.44632784]
[-0.39968472 -0.30891593 0.26971261]
[-0.97963754 -0.43935199 0.15300412]
[-0.84307983 0.24677206 0.0106483 ]
[-2.35090654 2.00428176 -0.86508382]
[ 0.13898218 -1.1189562 0.58942975]
[-0.49458671 -0.33104732 0.21866821]
[-1.57882026 -1.51404087 0.0477038 ]
[ 0.13859814 -0.5483948 0.21781288]
[-0.32611995 -1.25809464 0.37728814]
[-0.78536201 -0.68664035 0.1085742 ]
[-0.37105829 -0.05915608 0.02201072]
[ 0.27158475 -0.87453434 0.39653797]
[-0.87528053 -0.70164739 0.06171734]
[-0.68137371 -0.3759046 0.01025803]
[-1.04635432 0.08063389 -0.20307431]
[-0.70764594 0.39447045 -0.47675563]
[-0.66907161 0.33864334 -0.45493176]
[ 0.29213008 0.66596462 -0.18960451]
[ 0.1108387 -1.06351306 0.11219377]
[-0.15262856 -0.73238578 -0.04165411]
[-0.44995011 0.06452757 -0.32685704]
[-0.86099603 0.16377615 -0.4547338 ]
[-0.4114596 -0.54737622 -0.13912801]
[-0.27800444 -0.29505713 -0.17308967]
[ 0.16449654 -0.84945107 0.08546878]
[-0.20831598 -0.52815404 -0.0792276 ]
[ 0.065407 0.95108186 -0.32197991]
[ 0.25081348 -0.22964799 -0.49424191]
[ 0.23824718 -0.21146128 -0.50135144]
[ 0.24969516 -0.2280295 -0.49487461]
[ 1.07372462 -1.25574266 -0.01293764]
[ 2.3143954 0.46364636 0.25130471]
[ 1.73500035 0.07795969 0.14636314]
[ 1.5039631 -0.96024313 0.06333135]
[-0.56970194 -0.15420247 -0.84113861]
[ 0.21360791 -0.17580183 -0.51529138]
[-0.34991598 -0.13612732 -0.81882186]
[ 0.54571798 -1.08935378 -0.44480737]
[ 1.36292876 -0.1236054 -0.04543082]
[ 1.05384298 -0.88756317 -0.68699989]
[ 2.09500491 -1.25510199 -0.16348238]
[ 1.79570327 -0.87928136 -0.33828801]
[ 1.89876922 -0.62312401 -0.24570877]
[-1.4876562 2.96226819 -0.11129725]
[ 1.9575568 -0.03267054 -0.23731517]
[ 2.40580247 -0.58821019 0.02517744]
[ 2.26808216 -0.41756644 -0.05547556]
[ 3.46987881 1.50643507 0.60789185]
[ 1.88292918 5.11620807 -0.42822362]
[ 2.63079463 -1.00162822 0.16195721]
[ 2.53605779 -0.88602439 0.10630664]
[ 3.30919283 -2.71376916 -0.05833179]
[ 2.14734614 -1.25449325 -0.73686503]
[ 2.81337421 -0.65338183 -0.30885677]
[ 1.37969232 5.49502645 1.45241711]
[ 3.37796208 1.09224205 0.2107926 ]
[ 0.47370804 4.52162531 0.42923747]
[ 4.50897322 1.85684909 1.07124624]
[ 3.14274433 1.20334346 -0.93224695]]

```
In [6]: ratio = model.explained_variance_ratio_
print(ratio)

[0.59078919 0.36332536 0.04155747]
```

```
In [10]: final = pd.DataFrame(data = temp,columns = ['principal component 1', 'principal component 2','principal component 3'])
print('Initial Data set',var.shape)
print('After applying PCA',final.shape)
final.head(3)

Initial Data set (81, 5)
After applying PCA (81, 3)
```

Out[10]:

	principal component 1	principal component 2	principal component 3
0	-2.612343	-0.644779	0.807051
1	-2.164822	-0.742867	0.606383
2	-2.178864	-0.722544	0.598439