The objective of this assignment is to implement PCA on a given dataset and analyse the results.

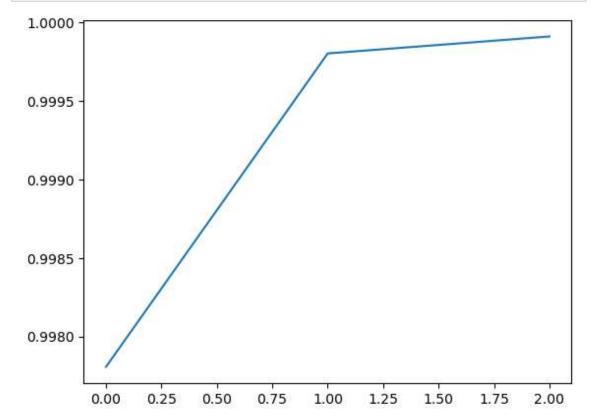
```
import pandas as pd
In [1]:
              df = pd.read csv('wine.csv')
In [2]:
              df
    Out[2]:
                    Wine Alcohol Malic.acid Ash
                                                  Acl
                                                       Mg Phenols Flavanoids Nonflavanoid.phenols
                            14.23
                                       1.71 2.43 15.6
                                                                                               0.28
                 0
                       1
                                                       127
                                                               2.80
                                                                           3.06
                 1
                       1
                            13.20
                                       1.78 2.14 11.2
                                                      100
                                                               2.65
                                                                           2.76
                                                                                                0.26
                                                                           3.24
                 2
                            13.16
                                       2.36 2.67 18.6 101
                                                               2.80
                                                                                               0.30
                       1
                 3
                                       1.95 2.50 16.8
                       1
                            14.37
                                                       113
                                                               3.85
                                                                           3.49
                                                                                                0.24
                 4
                       1
                            13.24
                                       2.59 2.87 21.0
                                                       118
                                                               2.80
                                                                           2.69
                                                                                                0.39
                                                                            ...
                      ...
                       3
                            13.71
                                       5.65 2.45 20.5
                                                                           0.61
                                                                                                0.52
               173
                                                        95
                                                               1.68
                                                       102
                                                                           0.75
               174
                       3
                            13.40
                                       3.91 2.48 23.0
                                                               1.80
                                                                                               0.43
               175
                       3
                            13.27
                                       4.28 2.26 20.0
                                                       120
                                                                1.59
                                                                           0.69
                                                                                                0.43
               176
                       3
                            13.17
                                       2.59 2.37 20.0 120
                                                               1.65
                                                                           0.68
                                                                                               0.53
                                       4.10 2.74 24.5
               177
                       3
                            14.13
                                                        96
                                                               2.05
                                                                           0.76
                                                                                               0.56
              178 rows × 14 columns
In [3]:
             x = df.drop('Wine',axis = 1 )
              y = df['Wine']
              from sklearn.model selection import train test split
In [4]:
              x_train,x_test,y_train,y_test = train_test_split(x,y,test_size = 0.20 , ra
              from sklearn.decomposition import PCA
In [5]:
              pca=PCA(n_components=3)
              рса
    Out[5]: PCA(n_components=3)
```

In a Jupyter environment, please rerun this cell to show the HTML representation or trust the notebook.

On GitHub, the HTML representation is unable to render, please try loading this page with nbviewer.org.

```
x train=pca.fit transform(x train)
In [6]:
            x train
   Out[6]: array([[-7.49349515e+01, -8.62866599e-01, -7.23588298e+00],
                   [-2.19971071e+02, -2.52597701e-01, -4.75918150e+00],
                   [-7.51038966e+01, -1.10615564e+01, 1.58317882e+00],
                   [-1.15087077e+02, -9.17867162e+00, -4.53814043e+00],
                   [ 2.85024471e+02, -7.83445476e+00, 2.82671368e+00],
                   [-4.22971985e+02, -6.02842277e-01, 3.30149652e+00],
                   [-2.84862447e+02, 3.80522761e+00, 3.72465069e+00],
                   [ 1.52899280e+01, 9.20255499e+00, 3.05842076e+00],
                   [-1.85024331e+02, -5.01070474e+00, -3.63174215e-02],
                   [ 4.50038279e+02, -7.81507097e+00, -1.44959905e+00],
                   [-2.49481611e+01, -3.03617323e+00, 2.55125821e+00],
                   [-3.69839061e+02, 8.54780405e+00, -4.03547114e+00],
                   [-9.48291814e+01, 4.35146401e+00, -7.90255258e-01],
                   [ 5.29900044e+02, -1.63311491e+01, -6.09145632e-01],
                   [-3.43079063e+02, -8.13757163e+00, 5.01451248e+00],
                   [ 2.35413393e+02, 1.31272425e+01, 2.01216202e+00],
                   [ 4.59897892e+02, -1.60625383e+01, 9.51760726e-01],
                   [-5.99146817e+01, -1.30757504e+00, -7.63503202e-01],
                   [-1.55099663e+02, -9.56374993e+00, -7.23980592e-02],
In [7]:
         pca.explained_variance_ratio_
```

Out[7]: array([9.97808300e-01, 1.99598478e-03, 1.07916251e-04])



In []: ▶