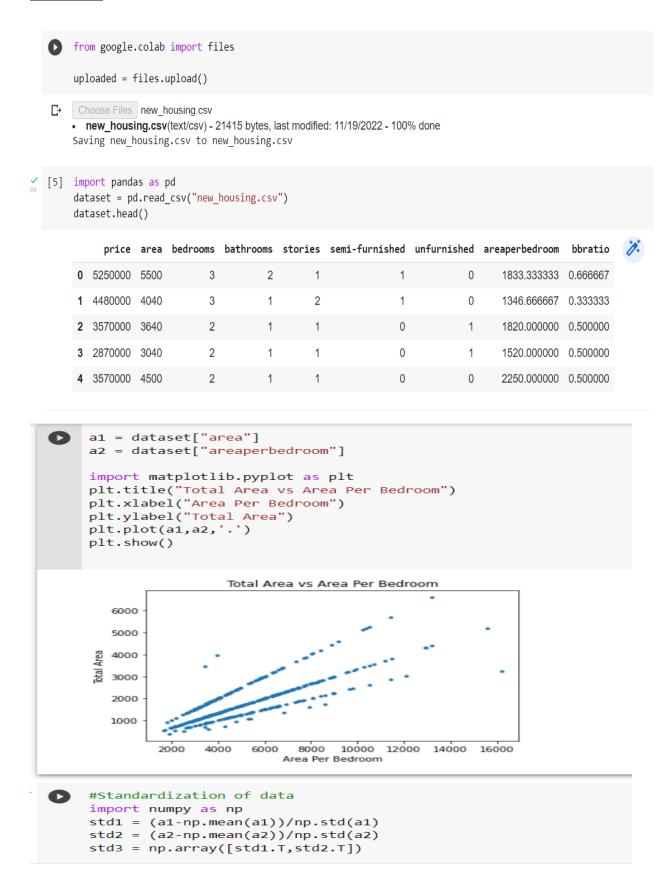
PCA without using Scikitlearn Framework (without using PCA in built function)



```
✓ [13] #Covariance
        covariance = np.cov(np.array([std1,std2]))
✓ [16] #EigenValue and EigenVector
        from numpy import linalg as LA
        eigen_val, eigen_vectors = LA.eig(covariance)
        print(eigen_val)
        print(eigen_vectors)
        [1.80892956 0.19474691]
       [[ 0.70710678 -0.70710678]
        [ 0.70710678  0.70710678]]
   #forming new feature set along PCA axes
        new feature = np.matmul(std3.T,eigen_vectors[1:].T)
        new_feature[:10, :] #printing only 10 rows
   □→ array([[ 0.12534103],
              [-0.76130751],
              [-0.49251428],
               [-0.9412383],
               [ 0.15065681],
               [ 1.19417762],
               [-1.26536113],
              [-0.19336493],
               [ 0.48971714],
              [ 0.82372857]])
```

PCA using Scikitlearn Framework (using PCA in built function)

```
[7] from sklearn.decomposition import PCA
     pca = PCA(n\_components = 1)
     decomposed pcs = pca.fit transform(std3.T)
     decomposed_pcs[:10, :]
     array([[ 0.12534103],
            [-0.76130751],
            [-0.49251428],
            [-0.9412383],
            [ 0.15065681],
            [ 1.19417762],
            [-1.26536113],
            [-0.19336493],
            [ 0.48971714],
            [ 0.82372857]])
     from sklearn.decomposition import PCA
     pca = PCA(n components = 2)
     decomposed pcs = pca.fit transform(std3.T)
     decomposed pcs
    array([[ 0.12534103, 0.10259959],
            [-0.76130751, 0.03693735],
            [-0.49251428, -0.49276294],
            [ 1.7953982 , 0.22627846],
            [ 5.31846177, -1.23561931],
            [-0.80503265, 0.03369922]])
```