

Q1

Find first and second principal components. Explain your solution step by step.

```
import numpy as np
import pandas as pd
# inputting the matrix x1 x2 - inserting the data set
A = np.matrix([[2.2,2.1],
               [0.4,0.6],
               [1.6,2.3],
               [2,2.1],
               [2.9,2.8],
               [0.8,1.1],
               [1.3,1.4],
               [1.1,1],
               [2.3,2.6]])

dataf = pd.DataFrame(A,columns = ['x1','x2'])
dataf

# standardize the dataset
df_standardized = (dataf - dataf.mean()) / (dataf.std())
df_standardized

# find the covariance (population) matrix for the given data set
dataf_cov = np.cov(df_standardized.T,bias = 1)
dataf_cov

# verify var(x1)
print('var(f1) (population formula): ',((df_standardized.x1)**2).sum()/9)

# verify cov(x1,x2)
print('cov(f1,f2) (population formula): ',((df_standardized.x1)*(df_standardized.x2

# calculate evalue and evector
eigen_val, eigen_vectors = np.linalg.eig(dataf_cov)
print("Eig-val:\n",eigen_val)
print("Eig-vec:\n",eigen_vectors)

# sort the e values and corresponding e vectors
number_of_components = 2
top_eigen_vectors = eigen_vectors[:, :number_of_components]
top_eigen_vectors.shape
np.array(df_standardized).shape
transformed_data = np.matmul(np.array(df_standardized),top_eigen_vectors)
pd.DataFrame(data = transformed_data
              , columns = ['Principal component '+ str(i+1) for i in range(number_of
transformed_data.shape
```

```
# using PCA in sklearn
from sklearn.decomposition import PCA
# define number of components
pca = PCA(n_components=number_of_components)
# transform standardized data set
principalComponents = pca.fit_transform(df_standardized)
# define the principal components data frame
principalDf = pd.DataFrame(data = principalComponents
                           , columns = ['Principal Component ' + str(i+1) for i in range(number_of
# print the principal components data frame
principalDf
```

```
var(f1) (population formula): 0.8888888888888888
cov(f1,f2) (population formula): 0.8406973999676008
Eig-val:
[0.04819149 1.72958629]
Eig-vec:
[[-0.70710678 -0.70710678]
 [ 0.70710678 -0.70710678]]
```

	Principal Component 1	Principal Component 2
0	-0.805194	-0.216626
1	2.156428	0.005114
2	-0.457292	0.496593
3	-0.628340	-0.039773
4	-2.063486	-0.196308
5	1.346075	0.108054
6	0.629953	-0.060092
7	1.172124	-0.248556
8	-1.350267	0.151594

Q2

Write a Python script that calculates approximations of an image using Singular Value Decomposition.

```
# import libraries
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import cv2

# import image as image.jpg
img = cv2.imread('image.jpg',0)
```

```
from numpy.linalg import svd
```

```
# k = 1,5,10,20,40
```

```
# create a methof to approximate image with various values for k
```

```
def approxi(img,k):
```

```
    U,s,V = svd(img,full_matrices = False)
```

```
    reconstructedMatrix = np.dot(U[:, :k], np.dot(np.diag(s[:k]), V[:k, :]))
```

```
    return reconstructedMatrix,s
```

```
# k = 1
```

```
print("k=1")
```

```
mat1 = approxi(img,1)
```

```
print(mat1)
```

```
# k = 5
```

```
print("k=5")
```

```
mat5 = approxi(img,5)
```

```
print(mat5)
```

```
# k = 10
```

```
print("k=10")
```

```
mat10 = approxi(img,10)
```

```
print(mat10)
```

```
# k = 20
```

```
print("k=20")
```

```
mat20 = approxi(img,20)
```

```
print(mat20)
```

```
# k = 40
```

```
print("k=40")
```

```
mat40 = approxi(img,40)
```

```
print(mat40)
```



```
[194.23151582, 194.35247381, 194.3993644 , ..., 55.2900777 ,
 55.42310729, 55.58840207],
[193.60001196, 193.71906289, 193.76618699, ..., 55.1705606 ,
 55.3059179 , 55.46935284],
...,
[119.72133701, 119.76887615, 119.77970202, ..., 51.17632091,
 51.25782013, 51.36332498],
[120.0475878 , 120.09518647, 120.10591302, ..., 50.99376397,
 51.07533356, 51.18164575],
[120.55535371, 120.60450333, 120.61573836, ..., 50.94964741,
 51.0307033 , 51.13744593]], array([1.96358642e+05, 3.82897607e+04,
 5.37743999e+00, 5.29537110e+00, 5.11965639e+00]))
```

```
k=10
```

```
(array([[191.86504485, 191.9599801 , 191.99134879, ..., 53.36913961,
 53.27068826, 53.31474643],
[192.22803615, 192.31553133, 192.3398733 , ..., 53.28161536,
 53.16357517, 53.19448081],
[192.03430799, 192.11647829, 192.13591917, ..., 53.43742188,
 53.30474079, 53.32480385],
...,
[131.32637939, 131.39087393, 131.43783377, ..., 58.01641972,
 58.00655212, 58.10478242],
```

```

58.09655212, 58.18478243],
[131.01943631, 131.08308502, 131.12787955, ..., 57.82616177,
 57.90501406, 57.99343717],
[130.5778039, 130.64223459, 130.68605553, ..., 57.42156969,
 57.50152186, 57.5911883 ]]), array([1.96358642e+05, 3.82897607e+04,
5.37743999e+00, 5.29537110e+00, 5.11965639e+00]))

k=20
(array([[186.04556562, 185.96646006, 185.84782509, ..., 40.10911694,
 40.16997758, 40.44031193],
[185.39834214, 185.28243502, 185.12784608, ..., 40.35205811,
 40.40033171, 40.67329799],
[185.06149251, 184.92427185, 184.74732627, ..., 40.60192927,
 40.63807898, 40.910929 ],
...,
[150.22824854, 150.30810668, 150.34828929, ..., 56.35300726,
 56.45438115, 56.50816184],
[150.30960908, 150.38961161, 150.42932713, ..., 55.992642 ,
 56.0950985 , 56.15119544],
[150.64687335, 150.72525883, 150.76293792, ..., 55.6733736 ,
 55.77002731, 55.82680655]]), array([1.96358642e+05, 3.82897607e+04,
5.37743999e+00, 5.29537110e+00, 5.11965639e+00]))

k=40
(array([[174.18713935, 174.09540178, 174.0488977 , ..., 44.69437725,
 44.90535326, 45.34361985],
[173.2170004 , 173.12722337, 173.07138564, ..., 44.66684437,
 44.89504374, 45.38153128],
[173.18547844, 173.10002939, 173.04021173, ..., 44.74191749,
 44.97489825, 45.48709844],
...,
[158.16046399, 158.19906625, 158.28456149, ..., 56.05659885,
 55.954847 , 55.76736066],
[158.45176595, 158.47992734, 158.55049821, ..., 55.93146475,
 55.82549256, 55.63311204],
[158.5253082 , 158.54797782, 158.60354239, ..., 55.77152053,
 55.65654732, 55.45329548]]), array([1.96358642e+05, 3.82897607e+04,
5.37743999e+00, 5.29537110e+00, 5.11965639e+00]))

```

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