

ASSIGNMENT 10 SOLUTION

Nayan Man Singh Pradhan

Excercise 1. (K -means clustering)

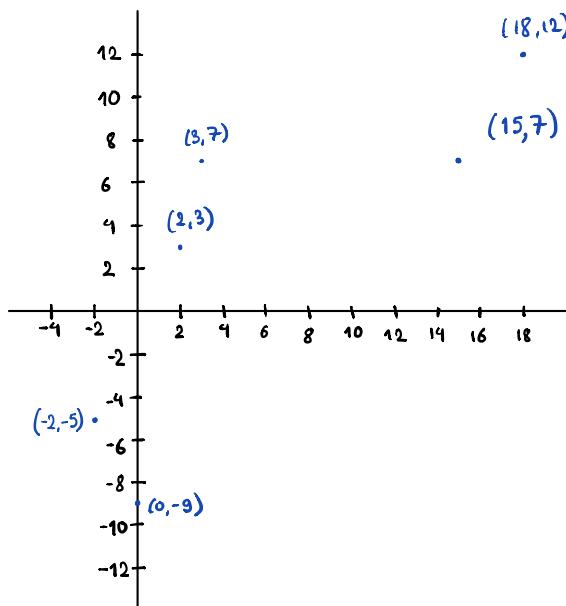
In this task, you carry out K -means clustering with $K = 3$ by paper and pencil. To this end, you are given the following data

i	x_i	$C^{(0)}(i)$
1	$(15, 7)^\top$	2
2	$(0, -9)^\top$	1
3	$(-2, -5)^\top$	1
4	$(2, 3)^\top$	0
5	$(3, 7)^\top$	2
6	$(18, 12)^\top$	0

with its initial cluster assignment. Let the clustering algorithm “run” until it finalized its assignment. Finally, draw the just computed clusters in a two-dimensional scatter plot.

(4 Points)

1)



Iteration 1

i) $N_0 = 2, N_1 = 2, N_2 = 2$

ii) $m_0' = \frac{1}{N_0} \left(\begin{pmatrix} 2 \\ 3 \end{pmatrix} + \begin{pmatrix} 18 \\ 12 \end{pmatrix} \right) = \frac{1}{2} \begin{pmatrix} 20 \\ 15 \end{pmatrix} = \begin{pmatrix} 10 \\ 7.5 \end{pmatrix}$

$$m_1' = \frac{1}{N_1} \left(\begin{pmatrix} 0 \\ -9 \end{pmatrix} + \begin{pmatrix} -2 \\ -5 \end{pmatrix} \right) = \frac{1}{2} \begin{pmatrix} -2 \\ -14 \end{pmatrix} = \begin{pmatrix} -1 \\ -7 \end{pmatrix}$$

$$m_2' = \frac{1}{N_2} \left(\begin{pmatrix} 15 \\ 7 \end{pmatrix} + \begin{pmatrix} 3 \\ 7 \end{pmatrix} \right) = \frac{1}{2} \begin{pmatrix} 18 \\ 14 \end{pmatrix} = \begin{pmatrix} 9 \\ 7 \end{pmatrix}$$

iii) $C^1(1) = \underset{1 \leq k \leq K}{\operatorname{argmin}} \|x_1 - m_k'\|^2$

$$k=1, \left\| \begin{pmatrix} 15 \\ 7 \end{pmatrix} - \begin{pmatrix} 10 \\ 7.5 \end{pmatrix} \right\|^2 = \left\| \begin{pmatrix} 5 \\ 0.5 \end{pmatrix} \right\|^2 = 5^2 + 0.5^2 = 25.25 //$$

$$k=2, \left\| \begin{pmatrix} 15 \\ 7 \end{pmatrix} - \begin{pmatrix} -1 \\ -7 \end{pmatrix} \right\|^2 = \left\| \begin{pmatrix} 16 \\ 14 \end{pmatrix} \right\|^2 = 16^2 + 14^2 = 452$$

$$k=3, \left\| \begin{pmatrix} 15 \\ 7 \end{pmatrix} - \begin{pmatrix} 9 \\ 7 \end{pmatrix} \right\|^2 = \left\| \begin{pmatrix} 6 \\ 0 \end{pmatrix} \right\|^2 = 6^2 = 36$$

$$\therefore C^*(1) = 1$$

$$C^*(2) = \operatorname{argmin}_{1 \leq k \leq K} \|\alpha_2 - m_k\|^2$$

$$k=1, \left\| \begin{pmatrix} 0 \\ -9 \end{pmatrix} - \begin{pmatrix} 10 \\ 7.5 \end{pmatrix} \right\|^2 = \left\| \begin{pmatrix} -10 \\ -16.5 \end{pmatrix} \right\|^2 = 372.25$$

$$k=2, \left\| \begin{pmatrix} 0 \\ -9 \end{pmatrix} - \begin{pmatrix} -1 \\ -7 \end{pmatrix} \right\|^2 = \left\| \begin{pmatrix} 1 \\ -2 \end{pmatrix} \right\|^2 = 5 //$$

$$k=3, \left\| \begin{pmatrix} 0 \\ -9 \end{pmatrix} - \begin{pmatrix} 9 \\ 7 \end{pmatrix} \right\|^2 = \left\| \begin{pmatrix} -9 \\ -16 \end{pmatrix} \right\|^2 = 337$$

$$\therefore C^*(2) = 2$$

$$C^*(3) = \operatorname{argmin}_{1 \leq k \leq K} \|\alpha_3 - m_k\|^2$$

$$k=1, \left\| \begin{pmatrix} -2 \\ -5 \end{pmatrix} - \begin{pmatrix} 10 \\ 7.5 \end{pmatrix} \right\|^2 = \left\| \begin{pmatrix} -12 \\ -12.5 \end{pmatrix} \right\|^2 = 300.25$$

$$k=2, \left\| \begin{pmatrix} -2 \\ -5 \end{pmatrix} - \begin{pmatrix} -1 \\ -7 \end{pmatrix} \right\|^2 = \left\| \begin{pmatrix} -1 \\ 2 \end{pmatrix} \right\|^2 = 5 //$$

$$k=3, \left\| \begin{pmatrix} -2 \\ -5 \end{pmatrix} - \begin{pmatrix} 9 \\ 7 \end{pmatrix} \right\|^2 = \left\| \begin{pmatrix} -11 \\ -12 \end{pmatrix} \right\|^2 = 265$$

$$\therefore C^*(3) = 2$$

$$C^*(4) = \operatorname{argmin}_{1 \leq k \leq K} \|\alpha_4 - m_k\|^2$$

$$k=1, \left\| \begin{pmatrix} 2 \\ 3 \end{pmatrix} - \begin{pmatrix} 10 \\ 7.5 \end{pmatrix} \right\|^2 = \left\| \begin{pmatrix} -8 \\ -4.5 \end{pmatrix} \right\|^2 = 84.25$$

$$k=2, \left\| \begin{pmatrix} 2 \\ 3 \end{pmatrix} - \begin{pmatrix} -1 \\ -7 \end{pmatrix} \right\|^2 = \left\| \begin{pmatrix} 3 \\ 10 \end{pmatrix} \right\|^2 = 109$$

$$k=3, \left\| \begin{pmatrix} 2 \\ 3 \end{pmatrix} - \begin{pmatrix} 9 \\ 7 \end{pmatrix} \right\|^2 = \left\| \begin{pmatrix} -7 \\ -4 \end{pmatrix} \right\|^2 = 65 //$$

$$\therefore C^1(4) = 3$$

$$C^1(5) = \underset{1 \leq k \leq K}{\operatorname{argmin}} \|x_5 - m_k\|^2$$

$$k=1, \left\| \begin{pmatrix} 3 \\ 7 \end{pmatrix} - \begin{pmatrix} 10 \\ 7.5 \end{pmatrix} \right\|^2 = \left\| \begin{pmatrix} 7 \\ -0.5 \end{pmatrix} \right\|^2 = 49.25$$

$$k=2, \left\| \begin{pmatrix} 3 \\ 7 \end{pmatrix} - \begin{pmatrix} -1 \\ -7 \end{pmatrix} \right\|^2 = \left\| \begin{pmatrix} 4 \\ 14 \end{pmatrix} \right\|^2 = 212$$

$$k=3, \left\| \begin{pmatrix} 3 \\ 7 \end{pmatrix} - \begin{pmatrix} 9 \\ 7 \end{pmatrix} \right\|^2 = \left\| \begin{pmatrix} -6 \\ 0 \end{pmatrix} \right\|^2 = 36 //$$

$$\therefore C^1(5) = 3$$

$$C^1(6) = \underset{1 \leq k \leq K}{\operatorname{argmin}} \|x_6 - m_k\|^2$$

$$k=1, \left\| \begin{pmatrix} 18 \\ 12 \end{pmatrix} - \begin{pmatrix} 10 \\ 7.5 \end{pmatrix} \right\|^2 = \left\| \begin{pmatrix} 8 \\ 4.5 \end{pmatrix} \right\|^2 = 84.25 //$$

$$k=2, \left\| \begin{pmatrix} 18 \\ 12 \end{pmatrix} - \begin{pmatrix} -1 \\ -7 \end{pmatrix} \right\|^2 = \left\| \begin{pmatrix} 19 \\ 19 \end{pmatrix} \right\|^2 = 722$$

$$k=3, \left\| \begin{pmatrix} 18 \\ 12 \end{pmatrix} - \begin{pmatrix} 9 \\ 7 \end{pmatrix} \right\|^2 = \left\| \begin{pmatrix} 9 \\ 5 \end{pmatrix} \right\|^2 = 106$$

$$\therefore C^1(6) = 1$$

After iteration = 1,

i	x_i	$C^{(1)}(i)$
1	$(15, 7)^\top$	0
2	$(0, -9)^\top$	1
3	$(-2, -5)^\top$	1
4	$(2, 3)^\top$	2
5	$(3, 7)^\top$	2
6	$(18, 12)^\top$	0

Since $C^0(i) \neq C^1(i)$, we need to iterate again!

Iteration 2

$$\text{i) } N_0 = 2, N_1 = 2, N_2 = 2$$

$$\text{ii) } m_0^1 = \frac{1}{N_0} \left(\begin{pmatrix} 15 \\ 7 \end{pmatrix} + \begin{pmatrix} 18 \\ 12 \end{pmatrix} \right) = \frac{1}{2} \begin{pmatrix} 33 \\ 19 \end{pmatrix} = \begin{pmatrix} 16.5 \\ 9.5 \end{pmatrix}$$

$$m_1^1 = \frac{1}{N_1} \left(\begin{pmatrix} 0 \\ -9 \end{pmatrix} + \begin{pmatrix} -2 \\ -5 \end{pmatrix} \right) = \frac{1}{2} \begin{pmatrix} -2 \\ -14 \end{pmatrix} = \begin{pmatrix} -1 \\ -7 \end{pmatrix}$$

$$m_2^1 = \frac{1}{N_2} \left(\begin{pmatrix} 2 \\ 3 \end{pmatrix} + \begin{pmatrix} 3 \\ 7 \end{pmatrix} \right) = \frac{1}{2} \begin{pmatrix} 5 \\ 10 \end{pmatrix} = \begin{pmatrix} 2.5 \\ 5 \end{pmatrix}$$

$$\text{iii) } C^2(1) = \underset{1 \leq k \leq K}{\operatorname{argmin}} \|x_1 - m_k\|^2$$

$$k=1, \left\| \begin{pmatrix} 15 \\ 7 \end{pmatrix} - \begin{pmatrix} 16.5 \\ 9.5 \end{pmatrix} \right\|^2 = \left\| \begin{pmatrix} -1.5 \\ -2.5 \end{pmatrix} \right\|^2 = 8.5 //$$

$$k=2, \left\| \begin{pmatrix} 15 \\ 7 \end{pmatrix} - \begin{pmatrix} -1 \\ -7 \end{pmatrix} \right\|^2 = \left\| \begin{pmatrix} 16 \\ 14 \end{pmatrix} \right\|^2 = 452$$

$$k=3, \left\| \begin{pmatrix} 15 \\ 7 \end{pmatrix} - \begin{pmatrix} 2.5 \\ 5 \end{pmatrix} \right\|^2 = \left\| \begin{pmatrix} 12.5 \\ 2 \end{pmatrix} \right\|^2 = 160.25$$

$$\therefore C^2(1) = 1$$

$$C^2(2) = \underset{1 \leq k \leq K}{\operatorname{argmin}} \|x_2 - m_k\|^2$$

$$k=1, \left\| \begin{pmatrix} 0 \\ -9 \end{pmatrix} - \begin{pmatrix} 16.5 \\ 9.5 \end{pmatrix} \right\|^2 = \left\| \begin{pmatrix} -16.5 \\ -18.5 \end{pmatrix} \right\|^2 = 614.5$$

$$k=2, \left\| \begin{pmatrix} 0 \\ -9 \end{pmatrix} - \begin{pmatrix} -1 \\ -7 \end{pmatrix} \right\|^2 = \left\| \begin{pmatrix} 1 \\ -2 \end{pmatrix} \right\|^2 = 5 //$$

$$k=3, \left\| \begin{pmatrix} 0 \\ -9 \end{pmatrix} - \begin{pmatrix} 2.5 \\ 5 \end{pmatrix} \right\|^2 = \left\| \begin{pmatrix} -2.5 \\ -14 \end{pmatrix} \right\|^2 = 202.25$$

$$\therefore C^2(2) = 2$$

$$C^2(3) = \underset{1 \leq k \leq K}{\operatorname{argmin}} \|x_3 - m_k\|^2$$

$$k=1, \left\| \begin{pmatrix} -2 \\ -5 \end{pmatrix} - \begin{pmatrix} 16.5 \\ 9.5 \end{pmatrix} \right\|^2 = \left\| \begin{pmatrix} -18.5 \\ -14.5 \end{pmatrix} \right\|^2 = 552.5$$

$$k=2, \left\| \begin{pmatrix} -2 \\ -5 \end{pmatrix} - \begin{pmatrix} -1 \\ -7 \end{pmatrix} \right\|^2 = \left\| \begin{pmatrix} 1 \\ 2 \end{pmatrix} \right\|^2 = 5 //$$

$$k=3, \left\| \begin{pmatrix} -2 \\ -5 \end{pmatrix} - \begin{pmatrix} 2.5 \\ 5 \end{pmatrix} \right\|^2 = \left\| \begin{pmatrix} -4.5 \\ -10 \end{pmatrix} \right\|^2 = 120.25$$

$$\therefore C^2(3) = 2$$

$$C^2(4) = \operatorname{argmin}_{1 \leq k \leq K} \|x_4 - m_k\|^2$$

$$k=1, \left\| \begin{pmatrix} 2 \\ 3 \end{pmatrix} - \begin{pmatrix} 16.5 \\ 9.5 \end{pmatrix} \right\|^2 = \left\| \begin{pmatrix} -14.5 \\ -6.5 \end{pmatrix} \right\|^2 = 252.5$$

$$k=2, \left\| \begin{pmatrix} 2 \\ 3 \end{pmatrix} - \begin{pmatrix} -1 \\ -7 \end{pmatrix} \right\|^2 = \left\| \begin{pmatrix} 3 \\ 10 \end{pmatrix} \right\|^2 = 109$$

$$k=3, \left\| \begin{pmatrix} 2 \\ 3 \end{pmatrix} - \begin{pmatrix} 2.5 \\ 5 \end{pmatrix} \right\|^2 = \left\| \begin{pmatrix} -0.5 \\ -2 \end{pmatrix} \right\|^2 = 4.25 //$$

$$\therefore C^2(4) = 3$$

$$C^2(5) = \operatorname{argmin}_{1 \leq k \leq K} \|x_5 - m_k\|^2$$

$$k=1, \left\| \begin{pmatrix} 3 \\ 7 \end{pmatrix} - \begin{pmatrix} 16.5 \\ 9.5 \end{pmatrix} \right\|^2 = \left\| \begin{pmatrix} -13.5 \\ -2.5 \end{pmatrix} \right\|^2 = 188.5$$

$$k=2, \left\| \begin{pmatrix} 3 \\ 7 \end{pmatrix} - \begin{pmatrix} -1 \\ -7 \end{pmatrix} \right\|^2 = \left\| \begin{pmatrix} 4 \\ 14 \end{pmatrix} \right\|^2 = 212$$

$$k=3, \left\| \begin{pmatrix} 3 \\ 7 \end{pmatrix} - \begin{pmatrix} 2.5 \\ 5 \end{pmatrix} \right\|^2 = \left\| \begin{pmatrix} 0.5 \\ 2 \end{pmatrix} \right\|^2 = 4.25 //$$

$$\therefore C^2(5) = 3$$

$$C^2(6) = \operatorname{argmin}_{1 \leq k \leq K} \|x_6 - m_k\|^2$$

$$k=1, \left\| \begin{pmatrix} 18 \\ 12 \end{pmatrix} - \begin{pmatrix} 16.5 \\ 9.5 \end{pmatrix} \right\|^2 = \left\| \begin{pmatrix} 1.5 \\ 2.5 \end{pmatrix} \right\|^2 = 8.5 //$$

$$k=2, \left\| \begin{pmatrix} 18 \\ 12 \end{pmatrix} - \begin{pmatrix} -1 \\ -7 \end{pmatrix} \right\|^2 = \left\| \begin{pmatrix} 19 \\ 19 \end{pmatrix} \right\|^2 = 722$$

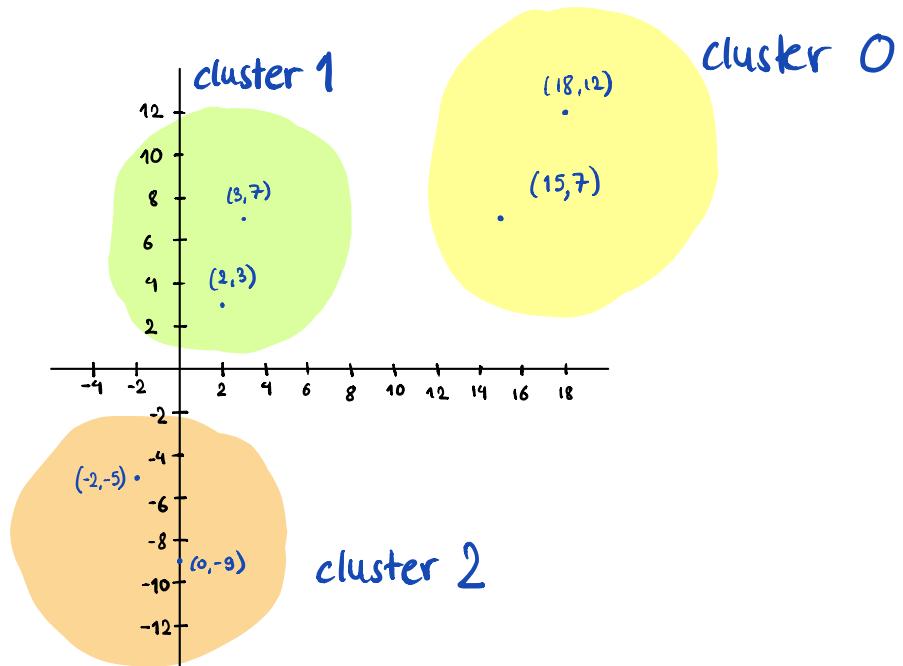
$$k=3, \left\| \begin{pmatrix} 18 \\ 12 \end{pmatrix} - \begin{pmatrix} 2.5 \\ 5 \end{pmatrix} \right\|^2 = \left\| \begin{pmatrix} 15.5 \\ 7 \end{pmatrix} \right\|^2 = 289.25$$

$$\therefore C^2(6) = 1$$

After iteration = 2,

i	x_i	$C^{(2)}(i)$
1	$(15, 7)^\top$	0
2	$(0, -9)^\top$	1
3	$(-2, -5)^\top$	1
4	$(2, 3)^\top$	2
5	$(3, 7)^\top$	2
6	$(18, 12)^\top$	0

Since $C^1 = C^2$, we can stop the algorithm. We get $C^2(i)$ as our clusters with $i = 1, 2, 3, \dots, 6$



Excercise 2. (Principle component analysis)

You are given the data set

$$\{(1, 0)^\top, (0.5, 1)^\top, (1, 0.5)^\top\}.$$

Manually compute the principle components of this data set by solving the associated eigenvalue problem.

(4 Points)

$$\begin{array}{cccc} x & 1 & 0.5 & 1 \\ y & 0 & 1 & 0.5 \end{array}$$

$$\text{Mean of } x = \bar{x} = 5/6$$

$$\text{Mean of } y = \bar{y} = 1/2$$

Ordered pairs: $(x, x), (x, y), (y, x), (y, y)$

Covariance of all ordered pairs

$$\begin{aligned} \text{Cov}(x, x) &= \frac{1}{N-1} \sum_{k=1}^N (x_{ik} - \bar{x}_i)(x_{jk} - \bar{x}_j) \\ &= \frac{1}{3-1} \left[\left(1 - \frac{5}{6}\right)^2 + \left(0.5 - \frac{5}{6}\right)^2 + \left(1 - \frac{5}{6}\right)^2 \right] \\ &= \frac{1}{2} \times \frac{1}{6} \\ &= \frac{1}{12} \end{aligned}$$

$$\begin{aligned} \text{Cov}(x, y) &= \frac{1}{3-1} \left[\left(1 - \frac{5}{6}\right)\left(0 - \frac{1}{2}\right) + \left(0.5 - \frac{5}{6}\right)\left(1 - \frac{1}{2}\right) + \left(1 - \frac{5}{6}\right)\left(0.5 - \frac{1}{2}\right) \right] \\ &= \frac{1}{2} \times -\frac{1}{9} \\ &= -\frac{1}{18} \end{aligned}$$

$$\text{Cov}(y, x) = \text{Cov}(x, y) = -\frac{1}{18}$$

$$\begin{aligned} \text{Cov}(y, y) &= \frac{1}{3-1} \left[\left(0 - \frac{1}{2}\right)^2 + \left(1 - \frac{1}{2}\right)^2 + \left(0.5 - \frac{1}{2}\right)^2 \right] \\ &= \frac{1}{4} \end{aligned}$$

$$\text{Covariance Matrix} = \begin{bmatrix} \text{cov}(x,x) & \text{cov}(x,y) \\ \text{cov}(y,x) & \text{cov}(y,y) \end{bmatrix}$$

$$C = \begin{bmatrix} 1/12 & -1/8 \\ -1/8 & 1/4 \end{bmatrix}$$

Now, we find the eigenvalues of C

$$(C - \lambda I) = \begin{pmatrix} 1/12 & -1/8 \\ -1/8 & 1/4 \end{pmatrix} - \begin{pmatrix} \lambda & 0 \\ 0 & \lambda \end{pmatrix}$$

$$\begin{aligned} \det(C - \lambda I) &= \det \begin{pmatrix} \frac{1}{12} - \lambda & -\frac{1}{8} \\ -\frac{1}{8} & \frac{1}{4} - \lambda \end{pmatrix} \\ &= \left(\frac{1}{12} - \lambda\right)\left(\frac{1}{4} - \lambda\right) - \left(-\frac{1}{8}\right)\left(-\frac{1}{8}\right) \\ &= \frac{1}{12}\left(\frac{1}{4} - \lambda\right) - \lambda\left(\frac{1}{4} - \lambda\right) - \frac{1}{64} \\ &= \frac{1}{48} - \frac{1}{12}\lambda - \frac{1}{4}\lambda + \lambda^2 - \frac{1}{64} \end{aligned}$$

$$0 = \lambda^2 - \frac{1}{3}\lambda + \frac{1}{192}$$

$$\lambda_1 = 0.316, \lambda_2 = 0.016$$

Eigenvector corresponding to eigenvalue 0.316 is

$$\begin{pmatrix} \frac{1}{12} - \lambda_1 & -\frac{1}{8} \\ -\frac{1}{8} & \frac{1}{4} - \lambda_1 \end{pmatrix} \begin{pmatrix} u_1 \\ u_2 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

$$\begin{pmatrix} -0.232 & -0.925 \\ -0.925 & -0.066 \end{pmatrix} \begin{pmatrix} u_1 \\ u_2 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

\therefore Eigenvector of λ_1 is $\begin{bmatrix} 2 - \sqrt{13} \\ 3 \end{bmatrix}$

Eigenvector corresponding to eigenvalue 0.016 is

$$\begin{pmatrix} \frac{1}{12} - \lambda_2 & -\frac{1}{8} \\ -\frac{1}{8} & \frac{1}{4} - \lambda_2 \end{pmatrix} \begin{pmatrix} u_1 \\ u_2 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

$$\begin{pmatrix} 0.067 & -0.125 \\ -0.125 & 0.234 \end{pmatrix} \begin{pmatrix} u_1 \\ u_2 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

∴ Eigenvector of λ_2 is $\begin{bmatrix} 2 + \sqrt{13} \\ 3 \end{bmatrix}$

∴ Principle components are $\left\{ \begin{bmatrix} 2 - \sqrt{13} \\ 3 \end{bmatrix}, \begin{bmatrix} 2 + \sqrt{13} \\ 3 \end{bmatrix} \right\}$

Excercise 3. (Principle component analysis for data compression)

A well-known application of principle component analysis is lossy data compression. In this application, you are given a large data set $\{\mathbf{x}\}_{i=1}^N$ with $\mathbf{x}_i \in \mathbb{R}^D$ and reduce it to a data set $\{\tilde{\mathbf{x}}\}_{i=1}^N$ with $\tilde{\mathbf{x}}_i \in \mathbb{R}^d$ where $d < D$, while storing a matrix that allows to reconstruct an approximation to the \mathbf{x}_i from the vectors $\tilde{\mathbf{x}}_i$.

Develop and give a compression and a decompression algorithm which carry out the above described data compression and decompression tasks by using principle component analysis. You can either try to develop the idea by yourself or do a research in the internet. In the latter case, please quote the source from which you took the information.

(4 Points)

Input: $\{x\}_{i=1}^N$ where $x_1, x_2, x_3, x_4, \dots, x_N \in \mathbb{R}^D$

Step 1: Preprocessing / mean normalization

$$\mu_j = \frac{1}{m} \sum_{i=1}^m x_j^{(i)}$$

Mean normalized dataset (\bar{x})

$$\bar{x}_j = x_j - \mu_j$$

Step 2: Compute covariance matrix (Σ)

$$\Sigma = \frac{1}{m} \sum_{i=1}^n (\bar{x}^{(i)}) (\bar{x}^{(i)})^\top$$

Step 3: Compute "eigenvectors" of matrix Σ using SVD

$$[U, S, V] = \text{svd}(\Sigma)$$

$$\text{We get } U = \begin{bmatrix} | & | & | & | & \dots & | \\ u_1^{(1)} & u_1^{(2)} & u_1^{(3)} & u_1^{(4)} & \dots & u_1^{(n)} \end{bmatrix} \in \mathbb{R}^{n \times n}$$

Now in order to compress $\{x\}_{i=1}^N$ with $x_i \in \mathbb{R}^D$ and reduce it to $\{\tilde{x}\}_{i=1}^N$, we simply take the first N columns of U and store it into a new vector $U_{\text{reduce}} = U(:, 1:k)$.

$$z = U_{\text{reduce}}^\top \cdot x$$

Reference Video: <https://www.youtube.com/watch?v=rng04VJxUt4>

