

Exam I (In class portion 10%)

Show your detailed work for full credit

1. (1 pt) The degrees of morning temperature (X_1) and afternoon temperature (X_2) in Celsius for last three days are organized into the data matrix \mathbf{X} . $\mathbf{X} = \begin{bmatrix} 4 & 1 \\ -1 & 3 \\ 3 & 5 \end{bmatrix}$. Compute $\text{Cov}(\mathbf{X})$, the sample variance-covariance matrix of \mathbf{X} .

$$s_{xy} = \frac{1}{n-1} \sum (x_i - \bar{x})(y_i - \bar{y}) \quad n=3, \quad \bar{x}=2, \quad \bar{y}=3$$

$$\begin{aligned} s_{xx} &= \frac{1}{2} [(4-2)^2 + (-1-2)^2 + (3-2)^2] \\ &= \frac{14}{2} = 7 \end{aligned}$$

$$\begin{aligned} s_{yy} &= \frac{1}{2} [(1-3)^2 + (3-3)^2 + (5-3)^2] \\ &= 4 \end{aligned}$$

$$\begin{aligned} s_{xy} &= \frac{1}{2} [(4-2)(1-3) + (-1-2)(3-3) + (3-2)(5-3)] \\ &= \frac{1}{2} [-4 + 0 + 2] \\ &= -1 \end{aligned}$$

$$\text{cov} = \begin{pmatrix} 7 & -1 \\ -1 & 4 \end{pmatrix}$$

2. Suppose the sample correlation matrix of a bivariate data is $\mathbf{R} = \begin{bmatrix} 1 & r \\ r & 1 \end{bmatrix}$, $r > 0$

a. (1 pt) Find the eigenvalues of \mathbf{R} .

$$\det(\mathbf{R} - \lambda \mathbf{I}) = \begin{vmatrix} 1-\lambda & r \\ r & 1-\lambda \end{vmatrix} = (1-\lambda+r)(1-\lambda-r) = 0$$

$$(1-\lambda+r) = 0 \quad 1-\lambda-r = 0$$

$$\lambda_1 = 1+r > \lambda_2 = 1-r$$

b. (1 pt) Find the normalized eigenvectors of \mathbf{R} .

$$\lambda_1 = 1+r. \quad (\mathbf{R} - \lambda_1 \mathbf{I}) \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{bmatrix} -r & r \\ +r & -r \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

$$-x_1 + x_2 = 0 \quad \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 1 \\ 1 \end{bmatrix} \Rightarrow \text{normalize} \quad \begin{bmatrix} \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} \end{bmatrix}$$

$$\lambda_2 = 1-r \quad (\mathbf{R} - \lambda_2 \mathbf{I}) \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{bmatrix} r & r \\ r & r \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

$$x_1 + x_2 = 0 \quad \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 1 \\ -1 \end{bmatrix} \Rightarrow \text{normalize} \quad \begin{bmatrix} \frac{1}{\sqrt{2}} \\ -\frac{1}{\sqrt{2}} \end{bmatrix}$$

c. (1 pt) What are the two principal components?

$$Y_1 = \frac{1}{\sqrt{2}} x_1 + \frac{1}{\sqrt{2}} x_2 = \left(\frac{1}{\sqrt{2}}, \frac{1}{\sqrt{2}} \right) \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \mathbf{a}^T \mathbf{X} \quad \mathbf{a}^T = \left(\frac{1}{\sqrt{2}}, \frac{1}{\sqrt{2}} \right)$$

$$Y_2 = \frac{1}{\sqrt{2}} x_1 - \frac{1}{\sqrt{2}} x_2 = \left(\frac{1}{\sqrt{2}}, -\frac{1}{\sqrt{2}} \right) \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \mathbf{b}^T \mathbf{X} \quad \mathbf{b}^T = \left(\frac{1}{\sqrt{2}}, -\frac{1}{\sqrt{2}} \right)$$

d. (1 pt) Calculate the sample variances of the two principal components.

$$\text{Var}(Y_1) = \left(\frac{1}{\sqrt{2}}, \frac{1}{\sqrt{2}} \right) \text{cov}(\mathbf{X}) \begin{pmatrix} \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} \end{pmatrix} = \frac{1}{2} (1, 1) \begin{pmatrix} 1 & r \\ r & 1 \end{pmatrix} \frac{1}{\sqrt{2}} \begin{pmatrix} 1 \\ 1 \end{pmatrix}$$

$$= 1+r$$

$$\text{Var}(Y_2) = \left(\frac{1}{\sqrt{2}}, -\frac{1}{\sqrt{2}} \right) \text{cov}(\mathbf{X}) \begin{pmatrix} \frac{1}{\sqrt{2}} \\ -\frac{1}{\sqrt{2}} \end{pmatrix} = \frac{1}{2} (1, -1) \begin{pmatrix} 1 & r \\ r & 1 \end{pmatrix} \begin{pmatrix} 1 \\ -1 \end{pmatrix}$$

$$= 1-r$$

3. (3 pts) Explain how to find the first principal component Y_1 and derive that $\text{Var}(Y_1) = \lambda_1$, where λ_1 is the largest eigenvalue of the variance-covariance matrix.

Suppose R is covariance matrix of multivariable X .
 λ_1 is the largest eigenvalue of R and $(a_{11}, a_{12}, \dots, a_{1q})^T$ is a standardized eigenvector corresponding to λ_1 . Then the first principal component $Y_1 = a_{11}X_1 + a_{12}X_2 + \dots + a_{1q}X_q$

$$= \vec{a}_1^T X$$

$$\text{Var}(Y_1) = \vec{a}_1^T \text{cov}(X) \vec{a}_1 \quad \text{cov}(X) \vec{a}_1 = \lambda_1 \vec{a}_1$$

$$= \vec{a}_1^T \lambda_1 \vec{a}_1$$

$$= \lambda_1 \vec{a}_1^T \vec{a}_1$$

$$= \lambda_1$$

4. (2 pts) Let X_1, X_2, \dots, X_q are q multivariate variables with a sample-variance covariance matrix of S . The principal components of the variables, Y_j , $j = 1, 2, \dots, q$, are the linear combinations of the X_i 's, where $i = 1, 2, \dots, q$. Let $\lambda_1 \geq \lambda_2 \geq \dots \geq \lambda_q$ are eigenvalues and $\vec{a}_j = (a_{j1}, a_{j2}, \dots, a_{jq})^T$ are corresponding eigenvectors.

Show that the correlation of variable x_i with component y_j is $r_{x_i, y_j} = \frac{a_{ji} \sqrt{\lambda_j}}{s_{x_i}}$.

$$\begin{aligned} r_{x_i, y_j} &= \frac{\text{cov}(X_i, Y_j)}{s_{x_i} s_{y_j}} = \frac{\text{cov}(X_i, a_{j1}X_1 + a_{j2}X_2 + \dots + a_{jq}X_q)}{s_{x_i} s_{y_j}} \\ &= \frac{(0, \dots, 0, 1, 0, \dots, 0) \text{cov}(X) \begin{pmatrix} a_{j1} \\ a_{j2} \\ \vdots \\ a_{jq} \end{pmatrix}}{s_{x_i} s_{y_j}} \\ &= \frac{(0, \dots, 0, 1, 0, \dots, 0) \lambda_j \vec{a}_j}{s_{x_i} s_{y_j}} \quad \vec{a}_j = \begin{pmatrix} a_{j1} \\ a_{j2} \\ \vdots \\ a_{ji} \\ \vdots \\ a_{jq} \end{pmatrix} \\ &= \frac{\lambda_j a_{ji}}{s_{x_i} \sqrt{\lambda_j}} = \frac{\sqrt{\lambda_j} a_{ji}}{s_{x_i}} \end{aligned}$$