Assignment 1

Name _____

Reading Assignment:

Johnson and Wichern, Chapter 1, Sections 2.5 and 2.6, Chapter 4, and Chapter 3. Review matrix operations in Chapter 2 and Supplement 2A.

Written Assignment: Due Monday, January 24, in class.

1. Consider a random vector $\tilde{X} = (X_1 X_2 X_3)'$ with mean vector $\mu = (3, 2, 1)'$ and covariance matrix Σ . The eigenvalues of Σ are $\lambda_1 = 12$, $\lambda_2 = 6$, $\lambda_3 = 2$ and the corresponding eigenvectors are

	$\left[1/\sqrt{3}\right]$		$\left[2/\sqrt{6} \right]$		0]
$e_1 = $	$1/\sqrt{3}$	e ₂ =	$-1/\sqrt{6}$	e ₃ =	$1/\sqrt{2}$
	$1/\sqrt{3}$		$-1/\sqrt{6}$	~	$\begin{bmatrix} 0\\ 1/\sqrt{2}\\ -1/\sqrt{2} \end{bmatrix}$

Evaluate the following quantities. Parts (a), (b), (c), (d) and (g) should be done without using a computer.

(a) $|\Sigma| =$ (b) trace $(\Sigma) =$ (c) $V(\underset{\sim}{e'_1 X}) =$ (d) $\Sigma =$ (e) $\Sigma^{-1} =$ $\begin{bmatrix} \\ \\ \\ \end{bmatrix}$

(f)
$$\Sigma^{-1/2} = \begin{bmatrix} & & \\ & & \\ & & \end{bmatrix}$$

(g) Define $Y_1 = e'_1 X$, $Y_2 = e'_2 X$, and $Y_3 = e'_3 X$. Find the mean vector and covariance matrix for $Y = (Y_1 Y_2 Y_3)' = \Gamma' X$, where the i-th column of Γ is the i-th eigenvector for Σ . Evaluate

$$E(\underline{\tilde{Y}}) = \begin{bmatrix} & & \\ & & \end{bmatrix} \qquad V(\underline{\tilde{Y}}) = \Gamma' \Sigma \Gamma = \begin{bmatrix} & & \\ & & \end{bmatrix}$$

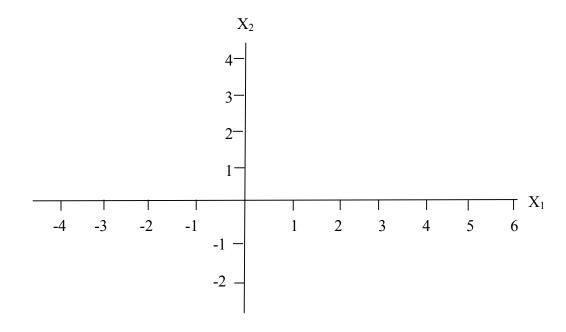
- (h) The linear transformation, $\underline{Y} = \Gamma' \underline{X}$, examined in part (g) is often called a rotation because it corresponds to simply rotating the coordinate axes. Use the definition of eigenvectors to show that lengths of vectors are not changed by this transformation, i.e., show $\underline{Y}' \underline{Y} = \underline{X}' \underline{X}$ when $\underline{Y} = \Gamma' \underline{X}$.
- (i) Let Γ be a matrix for which the i-th column of Γ is the i-th eigenvector for Σ , and let $\underline{Y} = \Gamma' \underline{X}$, $\mu_{\underline{X}} = E(\underline{X})$, $\Sigma_{\underline{X}} = V(\underline{X})$, $\mu_{\underline{Y}} = \Gamma' \mu_{\underline{X}}$, and $\Sigma_{\underline{Y}} = V(\underline{Y}) = \Gamma' \Sigma_{\underline{X}} \Gamma$. Determine which of the following measures of variability or distance are unaffected by rotations. Justify your answers by either giving a counter example or a proof using the properties of eigenvalues, eigenvectors, and matrix operations given in chapter 2 of Johnson and Wichern,
 - (a) Is $|\Sigma_X| = |\Sigma_Y|$?
 - (b) Is trace (Σ_X) = trace (Σ_Y) ?
 - (c) Is $\Sigma_X = \Sigma_Y$?
 - (d) Is $(X \mu_X)' \Sigma_X^{-1} (X \mu_X) = (Y \mu_Y)' \Sigma_Y^{-1} (Y \mu_Y)$?
 - (e) Is $(\underline{X} \mu_X)'(\underline{X} \mu_X) = (\underline{Y} \mu_Y)'(\underline{Y} \mu_Y)$?
- 2. Consider a bivariate normal population with mean vector $\mu_{\tilde{\nu}}$ and covariance matrix Σ , where

$$\mu = \begin{bmatrix} 2 \\ 1 \end{bmatrix} \quad \text{and} \quad \Sigma = \begin{bmatrix} 5 & -3 \\ -3 & 9 \end{bmatrix}$$

(a) Write down a formula for the joint density function.

(b) Find the eigenvalues and eigenvectors for Σ .

- (d) Write down the formula for the boundary of the smallest region such that there is probability 0.5 that a randomly selected observation will be inside the boundary.
- (e) Sketch the boundary of the region you identified in part (d).



- (f) Determine the area of the region described in parts (d) and (e).
- (g) Determine the area of the smallest region such that there is probability 0.95 that a randomly selected observation will be in that region.
- 3. The joint density function for X_1 and X_2 is

$$f(x_1, x_2) = 3\pi^{-1} \exp\{-4.5x_1^2 + 3x_1x_2 - 2.5x_2^2 - 3x_1 + 17x_2 - 32.5\}$$

Find the mean vector and the covariance matrix for this bivariate distribution.

4. Let X be a normally distributed random vector with

$$\mu = \begin{pmatrix} -3 \\ 1 \\ -2 \end{pmatrix} \text{ and } \Sigma = \begin{pmatrix} 4 & 0 & -1 \\ 0 & 5 & 0 \\ -1 & 0 & 2 \end{pmatrix}$$

- (a) Which of the following are pairs of independent random variables?
 - X_1 and X_2 (iv) (X_1, X_3) and X_2 (i) (v) $X_1 + X_3$ and $X_1 - 2X_2$ (ii) X_1 and X_3 (iii) X_1 and $X_1 + 3X_2 - 2X_3$ (vi) $X_1 + X_2 + X_3$ and $4X_1 - 2X_2 + 3X_3$
- What is the distribution of $Y = (X_1, X_2)'$? (b)
- What is the conditional distribution of $Y = (X_1, X_2)'$ given $X_3 = x_3$? (c)
- Find the correlation between X_1 and X_2 and a formula for the partial correlation (d) between X_1 and X_2 given $X_3 = x_3$.

$$\rho_{12} = \dots \qquad \rho_{12}$$

$$\rho_{12}$$

$$\rho_{12\bullet3} =$$

What is the conditional distribution of X_2 given $X_1 = x_1$ and $X_3 = x_3$? (e)

5. Suppose
$$X \sim N_4 \left(\begin{array}{c} \mu \\ \gamma \end{array}, \begin{array}{c} \Sigma \\ \end{array} \right)$$
 where

$$\mu = \begin{pmatrix} -4 \\ 3 \\ 0 \\ 1 \end{pmatrix} \qquad \Sigma = \begin{pmatrix} 4 & 0 & 2 & 2 \\ 0 & 1 & 1 & 0 \\ 2 & 1 & 9 & -2 \\ 2 & 0 & -2 & 4 \end{pmatrix}$$

- What is the distribution of $Z_1 = 3X_1 5X_2 + X_4$? (a)
- What is the joint distribution of Z_1 in part (a) and $Z_2 = 2X_1 X_3 + 3X_4$. (b)

(c) Find the conditional distribution of
$$\begin{pmatrix} X_1 \\ X_4 \end{pmatrix}$$
 given $X_3 = 1$.

(d) Find the partial correlation between X_1 and X_4 given $X_2 = x_2$ and $X_3 = x_3$. 6. Let X be $N(\mu_1, \Sigma_1)$ and Y be $N(\mu_2, \Sigma_2)$ where X and Y are independent and

$$\mu_1 = \begin{pmatrix} 2 \\ 6 \end{pmatrix} \qquad \Sigma_1 = \begin{pmatrix} 8 & -2 \\ -2 & 4 \end{pmatrix} \qquad \mu_2 = \begin{pmatrix} 3 \\ 4 \end{pmatrix} \qquad \Sigma_2 = \begin{pmatrix} 4 & 2 \\ 2 & 4 \end{pmatrix}$$

- (a) Evaluate $\operatorname{Cov}\left(\begin{array}{ccc} X Y, & X + Y \\ \sim & \sim & \sim & \sim \end{array}\right)$
- (b) Are X Y and X + Y independent random vectors? Explain.
- (c) Show that the joint distribution for the four dimensional random vector

$$\begin{bmatrix} X & -Y \\ \tilde{-} & \tilde{-} \\ X & +Y \\ \tilde{-} & \tilde{-} \end{bmatrix}$$
 is a multivariate normal distribution.

- 7. Let d(p,q) denote a measure of distance between p and q. Johnson and Wichern $\sim \sim$ indicate that any distance measure should possess the following four properties:

 - (ii) d(p,q) > 0 if $p \neq q$
 - (iii) d(p,q) = 0 if p = q
 - (iv) $d(p,q) \le d(p,r) + d(r,q)$, where $r = (r_1, r_2)'$.

These properties are satisfied, for example, by Euclidean distance, i.e., .

$$d(p, q) = [(p-q)'(p-q)]^{1/2}$$

(a) In this class we will often consider distance measures of the form $d(p,q) = [(p-q)'A(p-q)]^{1/2}$. Sometimes A will be the inverse of a covariance

matrix which implies that A is symmetric and positive definite. Show that properties (i) through (iv) are satisfied when A is symmetric and positive definite.

(b) Does A have to be both symmetric and positive definite for $d(p, q) = [(p-q)'A(p-q)]^{1/2}$ to satisfy properties (i) through (iv)?

Present your proofs, counter examples or explanations.

(c) For k-dimensional vectors it is obvious that the measure

$$d(p,q) = \max\{|p_1 - q_1|, |p_2 - q_2|, ..., |p_k - q_k|\}$$

satisfies properties (i), (ii), (iii). Either prove or disprove that it satisfies property (iv), the triangle inequality.

8. (a) By multiplication of partitioned matrices, verify for yourself that

$$\begin{bmatrix} I_{q \times q} & B \\ 0_{r \times q} & I_{r \times r} \end{bmatrix}^{-1} = \begin{bmatrix} I_{q \times q} & -B \\ 0_{r \times q} & I_{r \times r} \end{bmatrix} \text{ and } \begin{bmatrix} I_{r \times r} & 0_{r \times q} \\ B & I_{q \times q} \end{bmatrix}^{-1} = \begin{bmatrix} I_{r \times r} & 0_{r \times q} \\ -B & I_{q \times q} \end{bmatrix}$$

for any $q \times r$ matrix B, where $0_{a \times b}$ denotes a matrix of zeros and $I_{a \times a}$ denotes an identity matrix.

(b) Show that
$$\begin{vmatrix} I_{q \times q} & B \\ 0_{r \times q} & I_{r \times r} \end{vmatrix} = \begin{vmatrix} I_{r \times r} & 0_{r \times q} \\ B & I_{q \times q} \end{vmatrix} = 1$$
 for any $q \times r$ matrix B.

(c) By multiplication of partitioned matrices verify that if $A = \begin{bmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{bmatrix}$ is a square matrix for which A_{11} is $q \times q$ matrix and A_{22} is an $r \times r$ matrix and the inverse of A_{22}

exists, then

$$\begin{bmatrix} A_{11} - A_{12} A_{22}^{-1} A_{21} & 0_{q \times r} \\ 0_{r \times q} & A_{22} \end{bmatrix} = \begin{bmatrix} I_{q \times q} & -A_{12} A_{22}^{-1} \\ 0_{r \times q} & I_{r \times r} \end{bmatrix} \begin{bmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{bmatrix} \begin{bmatrix} I_{q \times q} & 0_{q \times r} \\ -A_{22}^{-1} A_{21} & I_{r \times r} \end{bmatrix}$$

(d) Use the results from parts (a) through (c) to show that

$$|A| = |A_{22}| |A_{11} - A_{12}A_{22}^{-1}A_{21}|$$
 for any square matrix A with $|A_{22}| \neq 0$.

(e) Use the results from parts (a) through (d) to show that

$$\mathbf{A}^{-1} = \begin{bmatrix} \mathbf{I}_{q \times q} & \mathbf{0}_{q \times r} \\ -\mathbf{A}_{22}^{-1} \mathbf{A}_{21} & \mathbf{I}_{r \times r} \end{bmatrix}^{-1} \begin{bmatrix} \left(\mathbf{A}_{11} - \mathbf{A}_{12} \mathbf{A}_{22}^{-1} \mathbf{A}_{21}\right)^{-1} & \mathbf{0}_{q \times r} \\ \mathbf{0}_{r \times q} & \mathbf{A}_{22}^{-1} \end{bmatrix} \begin{bmatrix} \mathbf{I}_{q \times q} & -\mathbf{A}_{12} \mathbf{A}_{22}^{-1} \\ \mathbf{0}_{r \times q} & \mathbf{I}_{r \times r} \end{bmatrix}^{-1}$$

for any symmetric matrix A with $|A_{22}| \neq 0$.

9. Consider a multivariate normal distribution with positive definite covariance matrix

$$\Sigma = \begin{bmatrix} \Sigma_{11} & \Sigma_{12} \\ \Sigma_{21} & \Sigma_{22} \end{bmatrix}$$
. Note that from part (d) in problem 8, $|\Sigma| = |\Sigma_{22}| |\Sigma_{11} - \Sigma_{12} \Sigma_{22}^{-1} \Sigma_{21}|$,

the product of determinants of covariance matrices for a marginal and a conditional normal distribution. Use the result from part (e) in problem 8 to expand

$$(\underline{\mathbf{x}} - \underline{\mathbf{\mu}})' \Sigma^{-1} (\underline{\mathbf{x}} - \underline{\mathbf{\mu}}) = \begin{pmatrix} \underline{\mathbf{x}}_1 - \underline{\mathbf{\mu}}_1 \\ \underline{\mathbf{x}}_2 - \underline{\mathbf{\mu}}_2 \end{pmatrix}' \begin{bmatrix} \Sigma_{11} & \Sigma_{12} \\ \Sigma_{21} & \Sigma_{22} \end{bmatrix}^{-1} \begin{pmatrix} \underline{\mathbf{x}}_1 - \underline{\mathbf{\mu}}_1 \\ \underline{\mathbf{x}}_2 - \underline{\mathbf{\mu}}_2 \end{pmatrix}$$

in an appropriate way to show that the multivariate normal density function is a product of the density functions for the marginal distribution of X_2 and a conditional distribution.

What happens when $\sum_{12} = 0$, i.e. when X_1 and X_2 are uncorrelated?

For additional practice you could do problems 3.6, 3.8, 3.9, 3.10, 3.14, 3.16 at the end of Chapter 3 and problems 4.1, 4.3, 4.4, 4.5, 4.14, 4.15, 4.16, 4.17 at the end of Chapter 4. Problem 4.8 gives a trivial example of a non-normal bivariate distribution with normal marginal distributions. We will consider a more substantial example on the next assignment. Do not hand in these additional problems; answers will be given on the solution sheet for this assignment.