

# CA3

Evangelos Batsis, Christopher Eason, Yastika Motilal

## Question 1:

Let  $X \sim N_p(\mu, \Sigma)$  with  $|\Sigma| > 0$ . Using the spectral decomposition of the covariance matrix, prove that

$$(X - \mu)' \Sigma^{-1} (X - \mu) \sim \chi_p^2$$

1. Start with  $\Sigma e_i = \lambda_i e_i$  and note that  $\Sigma$  is positive definite.

$$\Sigma e_p = \lambda_p e_p \quad (1)$$

$$\Sigma e_p = [\Sigma e_1 \dots \Sigma e_p]$$

$$[\lambda_1 e_1 \dots \lambda_p e_p]$$

$$[\lambda_1 e_1 \dots \lambda_p e_p] = e_p D_p$$

$$\Sigma e_p = e_p D_p$$

$$\Sigma^{-1} = e_p D_p^{-1} e_p^{-1}$$

$$\Sigma^{-1} = e_p D_p^{-1} e_p'$$

$$\Sigma^{-1} = \sum_{i=1}^p e_i \lambda_i^{-1} e_i' \quad (2)$$

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2. Using this, rewrite  $(X - \mu)' \Sigma^{-1} (X - \mu)$  as the sum of p squared variables

$$\begin{aligned}
& (X - \mu)' \Sigma^{-1} (X - \mu) \\
&= (X - \mu)' \left( \sum_{i=1}^p e_i \lambda_i^{-1} e_i' \right) (X - \mu) \quad (2) \\
&= \sum_{i=1}^p \lambda_i^{-1} (X - \mu)' e_i e_i' (X - \mu) \\
&= \sum_{i=1}^p \lambda_i^{-1} [e_i' (X - \mu)]^2 \\
&= \sum_{i=1}^p \left[ \frac{1}{\sqrt{\lambda_i}} e_i' (X - \mu) \right]^2
\end{aligned}$$


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3. Showing that this vector i.e.  $\left[ \frac{1}{\sqrt{\lambda_i}} e_i' (X - \mu) \right]$  is a multivariate normal with zero mean and identity covariance

$$A X \sim N_q(A\mu, A\Sigma A')$$

$$\left[ \frac{1}{\sqrt{\lambda_i}} e_i' \right] [X - \mu] \sim N\left[0, \left[ \left( \frac{1}{\sqrt{\lambda_i}} e_i' \right) \Sigma \left( \frac{1}{\sqrt{\lambda_i}} e_i' \right)' \right] \right]$$

Firstly, since  $X - \mu$  is centered, we know that the Mean of this vector is equal to zero.  
Secondly, to show that the Variance of this vector is equal to the identity matrix ...

$$\begin{aligned}
& \left( \frac{1}{\sqrt{\lambda_i}} e_i' \right) \Sigma \left( \frac{1}{\sqrt{\lambda_i}} e_i' \right)' \\
&= \frac{1}{\sqrt{\lambda_i}} e_i' \Sigma \frac{1}{\sqrt{\lambda_i}} e_i \\
&= \frac{1}{\lambda_i} e_i' \Sigma e_i \\
&= \frac{1}{\lambda_i} e_i' \lambda_i e_i \quad (1) \\
&= I
\end{aligned}$$

Remembering that  $e_i' e_i = I$  because the eigen vectors are orthogonal to each other

Therefore

$$\left[ \frac{1}{\sqrt{\lambda_i}} e_i' \right] [X - \mu] = Z_i \sim N[0, I]$$

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4. Finally showing Q.E.D

$$(X - \mu)' \Sigma^{-1} (X - \mu) = \sum_{i=1}^p \left[ \frac{1}{\sqrt{\lambda_i}} e_i' (X - \mu) \right]^2 = \sum_{i=1}^p Z_i^2 \sim \chi_p^2$$

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## Question 2

$$\begin{bmatrix} x1 \\ x2 \\ x3 \\ x4 \\ x5 \end{bmatrix} \sim N_5 \left( \begin{bmatrix} 5 \\ 0 \\ -2 \\ 6 \\ 2 \end{bmatrix}, \begin{bmatrix} 8 & 3 & -1 & 0 & 5 \\ 3 & 12 & 2 & 2 & -2 \\ -1 & 2 & 9 & 0 & 1 \\ 0 & 2 & 0 & 8 & 2 \\ 5 & -2 & 1 & 2 & 10 \end{bmatrix} \right)$$

Now define  $X1 = \begin{bmatrix} x1 \\ x2 \\ x4 \end{bmatrix}$  and  $X2 = \begin{bmatrix} x3 \\ x5 \end{bmatrix}$ . Find the conditional (joint) distribution of  $X1|X2 = \begin{bmatrix} -1 \\ 2 \end{bmatrix}$

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1. First change the matrix around

$$\begin{bmatrix} x1 \\ x2 \\ x4 \\ x3 \\ x5 \end{bmatrix} \sim N_5 \left( \begin{bmatrix} 5 \\ 0 \\ 6 \\ -2 \\ 2 \end{bmatrix}, \begin{bmatrix} 8 & 3 & 0 & | & -1 & 5 \\ 3 & 12 & 2 & | & 2 & -2 \\ 0 & 2 & 8 & | & 0 & 2 \\ \hline -1 & 2 & 0 & | & 9 & 1 \\ 5 & -2 & 2 & | & 1 & 10 \end{bmatrix} \right)$$

$$X1|X2 = \begin{bmatrix} -1 \\ 2 \end{bmatrix} \sim N(E[X1|X2], cov(X1, x2))$$

$$\mu_1 = \begin{bmatrix} 5 \\ 0 \\ 6 \end{bmatrix}$$

$$\mu_2 = \begin{bmatrix} -2 \\ 2 \end{bmatrix}$$

$$\Sigma_{11} = \begin{bmatrix} 8 & 3 & 0 \\ 3 & 12 & 2 \\ 0 & 2 & 8 \end{bmatrix}$$

$$\Sigma_{12} = \begin{bmatrix} -1 & 5 \\ 2 & -2 \\ 0 & 2 \end{bmatrix}$$

$$\Sigma_{21} = \begin{bmatrix} -1 & 2 & 0 \\ 5 & -2 & 2 \end{bmatrix}$$

$$\Sigma_{22} = \begin{bmatrix} 9 & 1 \\ 1 & 10 \end{bmatrix}$$

$$\Sigma_{21} = \begin{bmatrix} -1 & 2 & 0 \\ 5 & -2 & 2 \end{bmatrix}$$

$$X_2 = \begin{bmatrix} -1 \\ 2 \end{bmatrix}$$


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2. Calculate the  $E[X1|X2]$

$$E[X1|X2] = \mu_1 + \Sigma_{12} \Sigma_{22}^{-1}(X_2 - \mu_2)$$

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      [,1]
[1,] 4.831
[2,] 0.247
[3,] 5.978

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3. Calculate  $cov[X1, X2]$

$$Cov(X1, X2) = \Sigma_{11} - \Sigma_{12} \Sigma_{22}^{-1} \Sigma_{21}$$

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      [,1]    [,2]    [,3]
[1,]  5.247   4.371  -1.034
[2,]  4.371  11.056   2.449
[3,] -1.034   2.449   7.596

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