

# ICMU Mini-course on Gaussian Processes

## Homework 1: The Multivariate Normal Distribution

Please upload your completed homework using [this link](#) by the end of the day on Saturday, February 7. The document name should include your last name. Scans or clear photographs of hand-written answers are acceptable if they are uploaded as a single document. If you upload multiple documents with the same name (e.g., if you wish to update a previously submitted document), we will grade the latest version.

1. This question is about linear transformations of multivariate normal random vectors and singularity. Let  $X$  be a  $p$ -dimensional random vector such that  $X \sim N_p(\mu, \Sigma)$ . Consider the linear transformation  $Y = AX + b$ , where  $A$  is an  $m \times p$  constant matrix and  $b$  is an  $m \times 1$  constant vector.
  - (a) Derive the distribution of  $Y$  and clearly state the resulting mean vector and covariance matrix.
  - (b) Suppose  $p = 3$  and the covariance matrix  $\Sigma$  has eigenvalues  $\{4, 2, 0\}$ .
    - i. Describe the geometric shape of this distribution in  $\mathbb{R}^3$ .
    - ii. Explain why the joint probability density function (pdf) for  $X$  cannot be written in its standard form. Refer specifically to the properties of  $\det(\Sigma)$  and the existence of  $\Sigma^{-1}$ .
2. This question concerns the conditional distributions of multivariate normal random vectors and the Schur complement. Suppose the  $(p+q) \times 1$  vector  $X$  is distributed as  $N_{p+q}(\mu, \Sigma)$ . We partition  $X$ ,  $\mu$ , and  $\Sigma$  as follows:

$$X = \begin{bmatrix} X_1 \\ X_2 \end{bmatrix}, \quad \mu = \begin{bmatrix} \mu_1 \\ \mu_2 \end{bmatrix}, \quad \Sigma = \begin{bmatrix} \Sigma_{11} & \Sigma_{12} \\ \Sigma_{21} & \Sigma_{22} \end{bmatrix}$$

where  $X_1$  is  $p \times 1$  and  $X_2$  is  $q \times 1$ . Assume  $\Sigma_{22}$  is non-singular.

- (a) Define the transformation  $Z = X_1 - \Sigma_{12}\Sigma_{22}^{-1}X_2$ . Show that  $Z$  and  $X_2$  are independent.
  - (b) Use the result from part (a) to prove that the conditional distribution of  $X_1$  given  $X_2 = X_2$  is multivariate normal.
  - (c) State the conditional mean  $E[X_1|X_2 = X_2]$  and the conditional covariance  $\text{Var}(X_1|X_2 = X_2)$ . How does the conditional variance change as the value of  $X_2$  changes?
3. A common pitfall is assuming that zero covariance implies independence. This question explores the specific conditions under which that implication holds.
    - (a) Let  $X = (X_1, X_2, \dots, X_p)^\top$  follow a multivariate normal distribution  $N_p(\mu, \Sigma)$ . Prove that if  $\text{Cov}(X_i, X_j) = 0$  for all  $i \neq j$ , then the components  $\{X_i\}$  are mutually independent. *Hint: Compare the joint pdf or moment generating function (mgf) to the product of the marginals.*
    - (b) Provide a counter-example to show that if two random variables  $X$  and  $Y$  are individually normally distributed and  $\text{Cov}(X, Y) = 0$ , they are not necessarily independent.
    - (c) Explain why your counter-example in part (b) does not contradict the result in part (a).