

Machine Learning Worksheet 4

Multivariate Normal

1 Basic Gaussian

Problem 1: We decomposed the covariance matrix Σ into $U\Lambda U^T$ with the orthonormal matrix U and the diagonal matrix of eigenvalues Λ . Show that:

$$U\Lambda^{-1}U^T = \sum_{i=1}^d \frac{1}{\lambda_i} \mathbf{u}_i \mathbf{u}_i^T$$

Problem 2: Let L be an invertible n by n matrix over \mathbb{R} . Show that the linear transformation $Y = LX$ of a Gaussian is again a Gaussian. Use the moment parametrisation and the change of variable theorem to get there.

2 Conditioned Gaussian

You got a dataset with pairs \mathbf{x} and \mathbf{y} . You computed the covariance of the concatenated datapoints $\begin{pmatrix} \mathbf{x} \\ \mathbf{y} \end{pmatrix}$. The covariance matrix looks like this: $\begin{pmatrix} \Sigma_{X,X} & \Sigma_{X,Y} \\ \Sigma_{Y,X} & \Sigma_{Y,Y} \end{pmatrix}$. For simplicity we assume that μ_X and μ_Y are $\mathbf{0}$.

Problem 3: The underlying process of the probability distribution $p(\mathbf{y} \mid \mathbf{x})$ can be expressed as $Y = \mathbf{T}X + Z$ where $Z \sim \mathcal{N}(\mathbf{0}, \Sigma_{Y|X})$. What is the value of the matrix \mathbf{T} ?

Problem 4: What is the covariance of Z ?