

Gaussian Random Vectors, Gaussian Discriminant Analysis

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1. If a random vector $\mathbf{x} \sim \mathcal{N}(\boldsymbol{\mu}, \boldsymbol{\Sigma})$ is a Gaussian random vector, then the marginals are Gaussian, i.e. $p(x_j) \sim \mathcal{N}(\mu_j, \sigma_j)$. In this problem, we consider the converse. Let $X_1 \sim \mathcal{N}(0, 1)$ and define the random variable

$$X_2 = \begin{cases} X_1 & \text{with probability } 1/2 \\ -X_1 & \text{with probability } 1/2. \end{cases}$$

- a) What is $p(x_2)$?
- b) Sketch $p(x_1, x_2)$ in the x_1, x_2 plane (or describe the joint distribution of X_1, X_2).
- c) Is $[X_1 \ X_2]^T$ a Gaussian random vector?
2. For binary classification the MAP classification rule can be expressed using the log-likelihood ratio: if

$$\log \left(\frac{p(\mathbf{x}|y=0)}{p(\mathbf{x}|y=1)} \right) > 0 \quad (1)$$

then $\hat{y} = 0$, else $\hat{y} = 1$.

In *Gaussian discriminant analysis*, we assume that the class conditional distributions are Gaussian:

$$p(\mathbf{x}|y=0) = \frac{1}{(2\pi)^{\frac{n}{2}} |\boldsymbol{\Sigma}_0|^{\frac{1}{2}}} \exp \left(-\frac{1}{2} (\mathbf{x} - \boldsymbol{\mu}_0)^T \boldsymbol{\Sigma}_0^{-1} (\mathbf{x} - \boldsymbol{\mu}_0) \right)$$

and

$$p(\mathbf{x}|y=1) = \frac{1}{(2\pi)^{\frac{n}{2}} |\boldsymbol{\Sigma}_1|^{\frac{1}{2}}} \exp \left(-\frac{1}{2} (\mathbf{x} - \boldsymbol{\mu}_1)^T \boldsymbol{\Sigma}_1^{-1} (\mathbf{x} - \boldsymbol{\mu}_1) \right).$$

- a) In linear discriminant analysis, $\boldsymbol{\Sigma}_0 = \boldsymbol{\Sigma}_1$, and (1) is equivalent to the following linear classification rule: if

$$\mathbf{w}^T \mathbf{x} > c$$

then $\hat{y} = 0$, else $\hat{y} = 1$. Find \mathbf{w} and c in terms of $\boldsymbol{\Sigma}_0$, $\boldsymbol{\mu}_0$ and $\boldsymbol{\mu}_1$.

- b) In quadratic discriminant analysis, $\boldsymbol{\Sigma}_0 \neq \boldsymbol{\Sigma}_1$, and (1) is equivalent to the following quadratic rule: if

$$\mathbf{x}^T \mathbf{B} \mathbf{x} + \mathbf{w}^T \mathbf{x} > c$$

then $\hat{y} = 0$, else $\hat{y} = 1$. Find an expression for \mathbf{B} , \mathbf{w} and c in terms of $\boldsymbol{\Sigma}_0$, $\boldsymbol{\Sigma}_1$, $\boldsymbol{\mu}_0$ and $\boldsymbol{\mu}_1$.