CSCI3320: Homework 4, Spring 2020 Deadline: March 8th, 11:59 pm

Teacher: John C.S. Lui

1. In class, we discussed about the multivariate Gaussian distribution. Assume now we have a bivariate Gaussian distribution with mean vector and covariance matrix as:

mean vector is 
$$\boldsymbol{\mu}^T = [\mu_1, \mu_2]$$
  $\boldsymbol{\Sigma} = \begin{bmatrix} \sigma_1^2 & \rho \sigma_1 \sigma_2 \\ \rho \sigma_1 \sigma_2 & \sigma_2^2 \end{bmatrix}$ 

Show that the joint bivariate density function is:

$$p(x_1, x_2) = \frac{1}{2\pi\sigma_1\sigma_2\sqrt{1-\rho^2}} \exp\left[-\frac{1}{2(1-\rho^2)} \left(z_1^2 - 2\rho z_1 z_2 + z_2^2\right)\right]$$

where  $z_i = (x_i - \mu_i)/\sigma_i$ , i = 1, 2, are standardized variables

## **Answer:**

Given that

$$\Sigma = \left[ \begin{array}{cc} \sigma_1^2 & \rho \sigma_1 \sigma_2 \\ \rho \sigma_1 \sigma_2 & \sigma_2^2 \end{array} \right]$$

we have

$$\begin{split} |\Sigma| &= \sigma_1^2 \sigma_2^2 - \rho^2 \sigma_1^2 \sigma_2 = \sigma_1^2 \sigma_2 (1 - \rho^2) \\ |\Sigma|^{1/2} &= \sigma_1 \sigma_2 \sqrt{1 - \rho^2} \\ \Sigma^{-1} &= \frac{1}{\sigma_1^2 \sigma_2 (1 - \rho^2)} \begin{bmatrix} \sigma_2^2 & -\rho \sigma_1 \sigma_2 \\ -\rho \sigma_1 \sigma_2 & \sigma_1^2 \end{bmatrix} \end{split}$$

and  $(\mathbf{x} - \boldsymbol{\mu})^T \boldsymbol{\Sigma}^{-1} (\mathbf{x} - \boldsymbol{\mu})$  can be expanded as

$$\begin{aligned} \left[ x_1 - \mu_1 \ x_2 - \mu_2 \right] \left[ \begin{array}{cc} \frac{\sigma_2^2}{\sigma_1^2 \sigma_2 (1 - \rho^2)} & -\frac{\rho \sigma_1 \sigma_2}{\sigma_1^2 \sigma_2 (1 - \rho^2)} \\ -\frac{\rho \sigma_1 \sigma_2}{\sigma_1^2 \sigma_2 (1 - \rho^2)} & \frac{\sigma_1^2}{\sigma_1^2 \sigma_2 (1 - \rho^2)} \end{array} \right] \left[ \begin{array}{c} x_1 - \mu_1 \\ x_2 - \mu_2 \end{array} \right] \\ = \frac{1}{1 - \rho^2} \left[ \left( \frac{x_1 - \mu_1}{\sigma_1} \right)^2 - 2\rho \left( \frac{x_1 - \mu_1}{\sigma_1} \right) \left( \frac{x_2 - \mu_2}{\sigma_2} \right) + \left( \frac{x_2 - \mu_2}{\sigma_2} \right)^2 \right] \end{aligned}$$

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## 2.

Let us say we have two variables  $x_1$  and  $x_2$  and we want to make a quadratic fit using them, namely,

$$f(x_1, x_2) = w_0 + w_1 x_1 + w_2 x_2 + w_3 x_1 x_2 + w_4 (x_1)^2 + w_5 (x_2)^2$$

How can we find  $w_i$ , i = 0, ..., 5, given a sample of  $\mathcal{X} = \{x_1^t, x_2^t, r^t\}$ ?

## **Answer:**

We write the fit as

$$f(x_1, x_2) = w_0 + w_1 z_1 + w_2 z_2 + w_3 z_3 + w_4 z_4 + w_5 z_5$$

where  $z_1 = x_1$ ,  $z_2 = x_2$ ,  $z_3 = x_1x_2$ ,  $z_4 = (x_1)^2$ , and  $z_5 = (x_2)^2$ . We can then use linear regression to learn  $w_i$ , i = 0, ..., 5. The linear fit in the five-dimensional  $(z_1, z_2, z_3, z_4, z_5)$  corresponds to a quadratic fit in the two-dimensional  $(x_1, x_2)$  space. We discuss such generalized linear models in more detail (and other nonlinear basis functions) in chapter 10.