Case 3: It is more general case.

$$\Sigma_1 = \begin{bmatrix} 1 & -0.5 \\ -0.5 & 2 \end{bmatrix}$$
 $\Sigma_2 = \begin{bmatrix} 1 & 1.2 \\ 1.2 & 3 \end{bmatrix}$ are not independent

2. The decision surface is hyper quadrank in nothing

3. Covariance matrix is arbitrary

Fram (1); we can't ignore anything here because of \(\) is arbitrary in nature.

From A we can write:

$$= -\frac{1}{2} \left[x^{\frac{1}{2}} x - 2 x^{\frac{1}{2}} x + 4 x^{\frac{1}{2}} x + 4 x^{\frac{1}{2}} x + 4 x^{\frac{1}{2}} x \right] + \ln \rho(x) - \frac{1}{2} \ln |x|$$

we can't ignore this because Zi is arbitrary

$$= x^{t} \left(\frac{1}{2} \sum_{i}^{t} \right) \times + \mu_{i}^{t} \sum_{i}^{t} x_{A} \left(\frac{1}{2} \mu_{i}^{t} \sum_{i}^{t} \mu_{i} \right) + \ln p(\omega_{i}) - \frac{1}{2} \ln \frac{1}{2} \left(\frac{1}{2} \mu_{i}^{t} \sum_{i}^{t} \mu_{i} \right) + \ln p(\omega_{i}) - \frac{1}{2} \ln \frac{1}{2} \left(\frac{1}{2} \mu_{i}^{t} \sum_{i}^{t} \mu_{i} \right) + \ln p(\omega_{i}) - \frac{1}{2} \ln \frac{1}{2} \left(\frac{1}{2} \mu_{i}^{t} \sum_{i}^{t} \mu_{i} \right) + \ln p(\omega_{i}) - \frac{1}{2} \ln \frac{1}{2} \left(\frac{1}{2} \mu_{i}^{t} \sum_{i}^{t} \mu_{i} \right) + \ln p(\omega_{i}) - \frac{1}{2} \ln \frac{1}{2} \left(\frac{1}{2} \mu_{i}^{t} \sum_{i}^{t} \mu_{i} \right) + \ln p(\omega_{i}) - \frac{1}{2} \ln \frac{1}{2} \left(\frac{1}{2} \mu_{i}^{t} \sum_{i}^{t} \mu_{i} \right) + \ln p(\omega_{i}) - \frac{1}{2} \ln \frac{1}{2} \left(\frac{1}{2} \mu_{i}^{t} \sum_{i}^{t} \mu_{i} \right) + \ln p(\omega_{i}) - \frac{1}{2} \ln \frac{1}{2} \left(\frac{1}{2} \mu_{i}^{t} \sum_{i}^{t} \mu_{i} \right) + \ln p(\omega_{i}) - \frac{1}{2} \ln \frac{1}{2} \left(\frac{1}{2} \mu_{i}^{t} \sum_{i}^{t} \mu_{i} \right) + \ln p(\omega_{i}) - \frac{1}{2} \ln \frac{1}{2} \left(\frac{1}{2} \mu_{i}^{t} \sum_{i}^{t} \mu_{i}^{t} \sum_{i}^{t} \mu_{i}^{t} \right) + \ln p(\omega_{i}) - \frac{1}{2} \ln \frac{1}{2} \left(\frac{1}{2} \mu_{i}^{t} \sum_{i}^{t} \mu_{i}^{t} \sum_{i}^{t} \mu_{i}^{t} \right) + \ln p(\omega_{i}) - \frac{1}{2} \ln \frac{1}{2} \left(\frac{1}{2} \mu_{i}^{t} \sum_{i}^{t} \mu_{i}^{t} \sum_{i}^{t} \mu_{i}^{t} \right) + \ln p(\omega_{i}) - \frac{1}{2} \ln \frac{1}{2} \ln$$

Buck

3

where

$$Ai = -\frac{1}{2} \Sigma_{i}^{-1}$$

The decision surface is quadratic hyperplane

Summary: Multivariate cure:

Case 1:
$$\Sigma_1 = \sigma^2 \Gamma$$
; Same for all class

Case 2?
$$\Sigma_i = \Sigma_i'$$
 same for all class

care 3:
$$\Sigma_1 + \Sigma_j$$
; different for different classes.

$$\Sigma_1 \neq \Sigma_2$$

Bivariali cure

Cuse 1:
$$\sigma_1^2 = \sigma_2^2 = \sum_{i=1}^{\infty} \left[\sigma_1^2 \circ \sigma_2^2 \right]$$

case 2:
$$\sigma_1^2 > \sigma_2^2 = \Sigma = \begin{bmatrix} \sigma_1^2 & 0 \\ 0 & \sigma_2^2 \end{bmatrix}$$

Care 3:
$$\Sigma = a_0 s_0 i_1 t_0 s_0 i_1$$
 = $\Sigma = \begin{bmatrix} \sigma_1^2 & \sigma_2^2 \\ \sigma_{21} & \sigma_{22}^2 \end{bmatrix}$