# CLASSIFICATION: DISCRIMINANT ANALYSIS STEP BY STEP

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### 1. Exercises on Discriminant Analysis

#### 1.1-1 Probabilities $\pi k$ :

$$\pi_G = \frac{Number\ of\ G}{Total} = \frac{10}{20} = 0.5$$
 
$$\pi_R = \frac{Number\ of\ R}{Total} = \frac{10}{20} = 0.5$$
 
$$\text{Mean:}\ \mu_G = \frac{(-1.6) + (-0.8) + (-1.8) + 0.6 + (-0.7) + (-1.8) + (-0.5) + (-0.3) + (-0.4) + (-1.3)}{10} = \frac{-8.6}{10} = -0.86$$
 
$$\mu_R = \frac{2.5 + 1.4 + 0.4 + (-1.2) + 2.1 + 1.0 + 1.0 + 1.9 + 1.8 + 1.6}{10} = \frac{12.5}{10} = 1.25$$
 
$$\text{Var:}\ \sigma^2 = \frac{1}{20 - 2} \sum_{k=1}^2 \sum_{i:y_i = k} (x_i - \mu_k)^2 = \frac{1}{20 - 2} \left( \sum_{i:y_i = G} (x_i - \mu_G)^2 + \sum_{i:y_i = R} (x_i - \mu_R)^2 \right)$$
 
$$= \frac{1}{18} \left[ (-1.6 + 0.86)^2 + (-1.3 + 0.86)^2 + (2.5 - 1.25)^2 + ^* + (1.6 - 1.25)^2 \right]$$
 
$$= \frac{1}{18} (5.7 + 11.7) = 0.9667$$

$$\sigma = \sqrt{0.9667} = 0.9832$$

The density of X: 
$$f_G(x) = \frac{1}{\sqrt{2\pi}*\sigma_G} e^{-\frac{1}{2}(\frac{x-\mu_G}{\sigma_G})^2} = \frac{1}{\sqrt{2\pi}*0.9832} e^{-\frac{1}{2}(\frac{x+0.86}{0.9832})^2}$$

$$f_R(x) = \frac{1}{\sqrt{2\pi} * \sigma_R} e^{-\frac{1}{2} (\frac{x - \mu_R}{\sigma_R})^2} = \frac{1}{\sqrt{2\pi} * 0.9832} e^{-\frac{1}{2} (\frac{x - 1.23}{0.9832})^2}$$

#### $1.1-2 p^k(x)$

$$P(Y = G|X = x) = p_G(X = x) = \frac{0.5 * f_G(x)}{0.5 * f_G(x) + 0.5 f_R(x)} = \frac{f_G(x)}{f_G(x) + f_R(x)}$$

$$P(Y = R|X = x) = p_R(X = x) = \frac{0.5 * f_R(x)}{0.5 * f_R(x) + 0.5 f_G(x)} = \frac{f_R(x)}{f_R(x) + f_G(x)}$$

Give x=0.5:

$$f_G(0.5) = \frac{1}{\sqrt{2\pi} * 0.9832} e^{-\frac{1}{2} (\frac{0.5 + 0.86}{0.9832})^2} = 0.156$$

$$f_R(0.5) = \frac{1}{\sqrt{2\pi} * 0.9832} e^{-\frac{1}{2} (\frac{0.5 - 1.23}{0.9832})^2} = 0.304$$

Then:

$$P(Y = G|X = 0.5) = \frac{0.156}{0.156 + 0.304} = 0.339$$
$$P(Y = R|X = 0.5) = \frac{0.304}{0.304 + 0.156} = 0.661$$

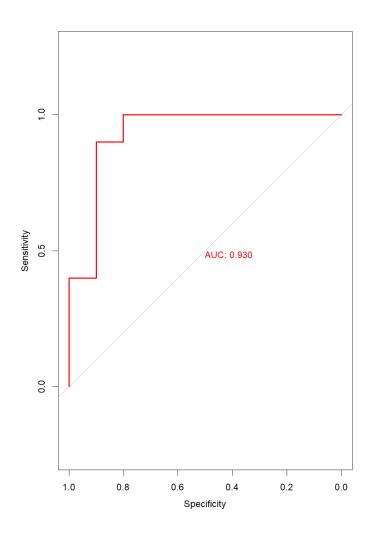
### 1.1-3 The function Ida() in R:

For x=0.5(the  $4^{th}$  number):

	R	CALCULATE
P(Y=G X=0.5)	0.26826346	0.339
P(Y=R X=0.5)	0.731736540	0.661

This may be due to rounding during calculation.

## 1.2 Measuring the accuracy with a ROC curve



- The vertical axis represents Sensitivity: shows the classifier's ability to correctly identify positive cases.
- 2) The horizontal axis represents Specificity (False Positive Rate): shows the proportion of negative cases incorrectly classified as positive.
- 3) The shape of the ROC curve indicates the trade-off between sensitivity and specificity at different thresholds.
- 4) The AUC is 0.930, which means the classifier has excellent overall performance.