



Agenda Detection and classification



- ▶ Motivation
- Assumptions
- Definitions
- ► Classification with LDA and QDA

Motivation Detection and Classification



Purpose:

► Sort new observations into one of two or more populations

Two-population Classification Detection and Classification



- \blacktriangleright π_1 : pdf $f_1(x)$, μ_1 , Σ_1
- \blacktriangleright π_2 : pdf $f_2(x)$, μ_2 , Σ_2
- $ightharpoonup X_0 \in \mathbb{R}^p$

$$d(X_0) = \begin{cases} d1, & X_0 \in R_1 \\ d2, & X_0 \in R_2 \end{cases}$$



► Cost Matrix:

Classify as \rightarrow	d_1	d_2
True π_1	0	$c[d_2 \pi_1]$
π_2	$c[d_1 \pi_2]$	0

- ► Prior Probabilities:
 - $p_1 = P[X_0 \in \pi_1]$

 - $ightharpoonup p_1 + p_2 = 1$

General Criteria Two population Classification



- ► General Criteria (using ECM)

General Criteria

Two population Classification



► General Criteria (using ECM)

$$p_1 f_1(x_0) c[d_2|\pi_1] \underset{d_2}{\overset{d_1}{\gtrless}} p_2 f_2(x_0) c[d_1|\pi_2]$$

$$\blacktriangleright \quad \frac{f_1(x_0)}{f_2(x_0)} \stackrel{d_1}{\underset{d_2}{\gtrless}} \frac{p_2}{p_1} \frac{c[d_1|\pi_2]}{c[d_2|\pi_1]}$$

ightharpoonup Equal Cost (min(TPM))

$$\blacktriangleright \quad \frac{f_1(X_0)}{f_2(x_0)} \overset{d_1}{\underset{d_2}{\gtrless}} \frac{p_2}{p_1}$$

General Criteria

Two population Classification



► General Criteria (using ECM)

$$p_1 f_1(x_0) c[d_2|\pi_1] \underset{d_2}{\overset{d_1}{\gtrless}} p_2 f_2(x_0) c[d_1|\pi_2]$$

$$\blacktriangleright \quad \frac{f_1(x_0)}{f_2(x_0)} \stackrel{d_1}{\underset{d_2}{\gtrless}} \frac{p_2}{p_1} \frac{c[d_1|\pi_2]}{c[d_2|\pi_1]}$$

ightharpoonup Equal Cost (min(TPM))

$$\blacktriangleright \quad \frac{f_1(X_0)}{f_2(x_0)} \underset{d_2}{\overset{d_1}{\geqslant}} \frac{p_2}{p_1}$$

- ► Equal Cost & Priors
 - ► $f_1(x_0) \overset{d_1}{\underset{d_2}{\gtrless}} f_2(x_0)$

General Criteria

Two population Classification



► General Criteria (using ECM)

ightharpoonup Equal Cost (min(TPM))

$$\blacktriangleright \quad \frac{f_1(X_0)}{f_2(x_0)} \overset{d_1}{\underset{d}{\gtrless}} \quad \frac{p_2}{p_1}$$

► Equal Cost & Priors

$$ightharpoonup f_1(x_0) \overset{d_1}{\underset{d_2}{\gtrless}} f_2(x_0)$$

► MAP

$$P[\pi_1|x_0] = \frac{p_1 f_1(x_0)}{p_1 f_1(x_0) + p_2 f_2(x_0)}$$

$$P[\pi_2|x_0] = \frac{p_1 f_1(x_0) + p_2 f_2(x_0)}{p_1 f_1(x_0) + p_2 f_2(x_0)}$$

Classification for 2 MVN populations



Having $X|_{\pi_i} \sim \mathcal{N}_p(\mu_i, \Sigma_i)$ for i = 1, 2

If $\Sigma_1 = \Sigma_2 \stackrel{def}{=} \Sigma$:

Classification for 2 MVN populations



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:

► LDA

$$(\mu_1 - \mu_2)^T \Sigma^{-1} x_0 - \frac{1}{2} (\mu_1 - \mu_2)^T \Sigma^{-1} (\mu_1 - \mu_2) \stackrel{d_1}{\underset{d_2}{\gtrless}} \log(\frac{p_2}{p_1} \frac{c(d_1 | \pi_2)}{c(d_2 | \pi_1)})$$

Classification for 2 MVN populations



Having $X|_{\pi_i} \sim \mathcal{N}_p(\mu_i, \Sigma_i)$ for i = 1, 2

If
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:

- ► LDA
 - $(\mu_1 \mu_2)^T \Sigma^{-1} x_0 \frac{1}{2} (\mu_1 \mu_2)^T \Sigma^{-1} (\mu_1 \mu_2) \underset{d_2}{\overset{d_1}{\gtrless}} \log(\frac{p_2}{p_1} \frac{c(d_1 | \pi_2)}{c(d_2 | \pi_1)})$

If $\Sigma_1 \neq \Sigma_2$:

- ► QDA

 - $k = \frac{1}{2} \log \frac{|\Sigma_1|}{|\Sigma_2|} + \frac{1}{2} \left(\mu_1^T \Sigma_1^{-1} \mu_1 \mu_2^T \Sigma_2^{-1} \mu_2 \right)$

Performance
Classification for 2 MVN populations



π_1	π_2
$n_{1C} \\ n_{1M}$	$n_{2M} \\ n_{2C}$
	n_{1C}

Performance
Classification for 2 MVN populations

	π_1	π_2
d_1	n_{1C}	n_{2M}
d_2	n_{1M}	n_{2C}

$$APER = \frac{n_{1M} + n_{2M}}{n_{1M} + n_{1C} + n_{2M} + n_{2C}}$$

$$ightharpoonup$$
 $A\hat{E}R$

