

# Chapter 1 - Introduction

## 2. Probability Theory

### The Rules of Probability

$$\text{sum rule} \quad p(X) = \sum_Y p(X, Y)$$

$$\text{product rule} \quad p(X, Y) = p(Y|X)p(X).$$

$$\text{Bayes' Theorem: } p(Y|X) = \frac{p(X|Y)p(Y)}{p(X)}$$

### 2.1. Probability Densities

$$\begin{aligned} p(x) &\geq 0 \\ \int_{-\infty}^{\infty} p(x) dx &= 1. \end{aligned} \quad p(x) = \int p(x, y) dy$$

### 2.2. Expectations and Covariances

$$\mathbb{E}[f] = \sum_x p(x) f(x) \quad \mathbb{E}[f] = \int p(x) f(x) dx.$$

$$\begin{aligned} \text{var}[x] &= \mathbb{E}[x^2] - \mathbb{E}[x]^2. \quad \text{cov}[x, y] = \mathbb{E}_{x,y} [\{x - \mathbb{E}[x]\} \{y - \mathbb{E}[y]\}] \\ &= \mathbb{E}_{x,y} [xy] - \mathbb{E}[x]\mathbb{E}[y] \end{aligned}$$

## 5. Decision Theory

### 5.1. Minimising the Misclassification Rate

decision regions - subdivisions of the input space

$R_1 \rightarrow C_1$ ;  $R_2 \rightarrow C_2$

decision boundary/surface - border between decision regions

$$\begin{aligned}
 p(\text{mistake}) &= p(\mathbf{x} \in \mathcal{R}_1, C_2) + p(\mathbf{x} \in \mathcal{R}_2, C_1) \\
 &= \int_{\mathcal{R}_1} p(\mathbf{x}, C_2) d\mathbf{x} + \int_{\mathcal{R}_2} p(\mathbf{x}, C_1) d\mathbf{x}.
 \end{aligned}
 \quad
 \begin{aligned}
 p(\text{correct}) &= \sum_{k=1}^K p(\mathbf{x} \in \mathcal{R}_k, C_k) \\
 &= \sum_{k=1}^K \int_{\mathcal{R}_k} p(\mathbf{x}, C_k) d\mathbf{x}
 \end{aligned}$$

red  $\rightarrow$  misclassified as  $C_1$

green  $\rightarrow$  misclassified as  $C_1$

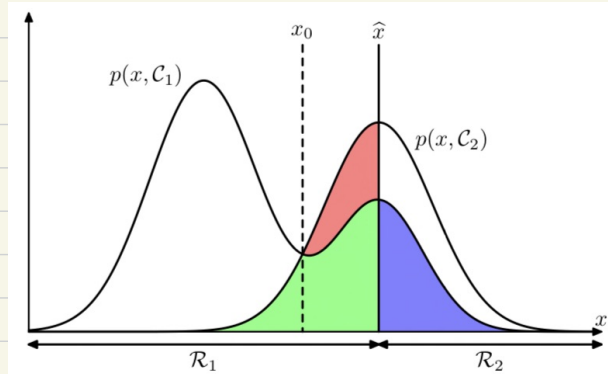
blue  $\rightarrow$  misclassified as  $C_2$

green + blue = const.

red - varies

minimise error=red

$\Rightarrow$  boundary  $x_0$  (red = 0)



### 5.2. Minimising the Expected Loss

loss/cost function - overall measure of loss due to taking a decision

utility function - negative of the loss

$L_{kj}$  - loss matrix

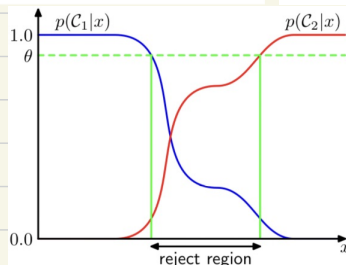
$$\mathbb{E}[L] = \sum_k \sum_j \int_{\mathcal{R}_j} L_{kj} p(\mathbf{x}, C_k) d\mathbf{x}. \quad \sum_k L_{kj} p(C_k | \mathbf{x})$$

### 5.3. The Reject Option

avoid making decisions

$K$  - number of classes

$1 > \theta > 1/K$



## 5.4. Inference and Decision

discrimination function - inference + decision

outlier/novelty detection - low accuracy predictions

generative models - model input as well as output

discriminative models - model posterior probabilities  $p(C|x)$

