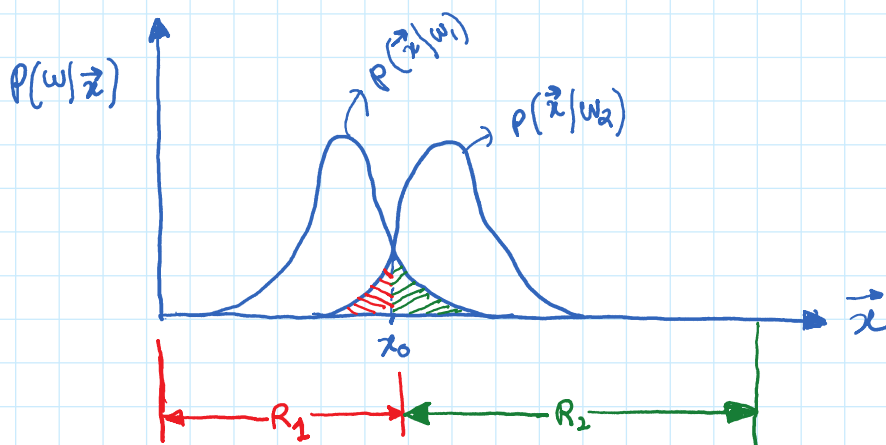


Bayesian Classification Part-3

• Decision Error:

Let P_e be the total probability of committing a decision error.



For two equiprobable classes, i.e., $P(w_1) = P(w_2) = \frac{1}{2}$,

$$P_e = \frac{1}{2} \int_{-\infty}^{x_0} P(\vec{x}|w_2) d\vec{x} + \frac{1}{2} \int_{x_0}^{+\infty} P(\vec{x}|w_1) d\vec{x} \quad (6)$$

P_e is equal to the total shaded area in the above figure.

For any values of $P(w_1)$ and $P(w_2)$, the above expression can be written as:

$$P_e = \int_{R_2} P(w_1) P(\vec{x}|w_1) d\vec{x} + \int_{R_1} P(w_2) P(\vec{x}|w_2) d\vec{x} \quad (7)$$

• Minimizing the Average Risk:

In practice, there are many applications where falsely classifying a feature vector that belongs to class-A to class-B needs to be penalized differently than falsely classifying a feature vector of class-B to class-A.

For example let w_1 and w_2 represent malignant tumor and benign tumor classes respectively. In such cases, falsely classifying a malignant tumor as a benign tumor has to be penalized more compared to falsely classifying a benign tumor as malignant tumor.

Let λ_{ij} represent the penalty associated with classifying a feature vector of i th class as j th class.

Then the risk function for $M=2$ is given by:

$$\gamma = \lambda_{12} P(\omega_1) \int_{R_2} p(\vec{x}|\omega_1) d\vec{x} + \lambda_{21} P(\omega_2) \int_{R_1} p(\vec{x}|\omega_2) d\vec{x} \quad (8)$$

For the above mentioned scenario of malignant and benign tumors, in order to minimize the risk of falsely classifying a malignant tumor, we can choose $\lambda_{12} > \lambda_{21}$

Generalizing the risk function to M -classes:

let γ_k be the risk (penalty) associated with wrongly assigning feature vectors of k^{th} class to other classes.

$$\gamma_k = P(\omega_k) \sum_{i=1}^M \int_{R_i} p(\vec{x}|\omega_k) d\vec{x} \quad (9)$$

Total penalty across all classes is given by:
 $\lambda_{ki} = 0$, for $k=i$

$$\gamma = \sum_{k=1}^M \gamma_k = \sum_{k=1}^M P(\omega_k) \left(\sum_{i=1}^M \lambda_{ki} \int_{R_i} p(\vec{x}|\omega_k) d\vec{x} \right)$$

Interchanging the summations in the above equation \Rightarrow

$$\gamma = \sum_{i=1}^M \int_{R_i} \left(\sum_{k=1}^M \lambda_{ki} P(\vec{x}|\omega_k) P(\omega_k) \right) d\vec{x} \quad (10)$$

The goal here is to choose partitionings (R_i) such that the penalty function: γ is minimized.

let l_i be defined as: $l_i \triangleq \sum_{k=1}^M \lambda_{ki} P(\vec{x}|\omega_k) P(\omega_k) \quad (11)$

Minimizing Eq (10) is equivalent to selecting partitioning region such that

$$\vec{x} \in R_i \text{ if } l_i < l_j, \forall j \neq i \quad (12)$$