

$$1. P_{x_j|w_i} = \frac{1}{P(P_i)} \lambda_j^{P_i} x_j^{P_i-1} e^{-\lambda_j x_j} \quad P_i, \lambda_j > 0.$$

$$(a) P(w_i | x_1=x_1, \dots, x_k=x_k) = \frac{P(x_1=x_1, \dots, x_k=x_k | w_i) P(w_i)}{P(x_1, x_2, \dots, x_k)}$$

$$\propto P(x_1|w_i) \cdot P(x_2|w_i) \cdots P(x_k|w_i) P(w_i).$$

$$= \prod_{j=1}^k \frac{1}{P(P_i)} \lambda_j^{P_i} x_j^{P_i-1} e^{-\lambda_j x_j} P(w_i)$$

$$= \frac{P(w_i)}{P(P_i)} \cdot \prod_{j=1}^k \lambda_j^{P_i} x_j^{P_i-1} e^{-\lambda_j x_j}$$

Decide w_i if.

$$\frac{P(w_1)}{P(P_1)} \cdot \prod_{j=1}^k (\lambda_j)^{P_1} (x_j)^{P_1-1} e^{-\lambda_j x_j} > \frac{P(w_2)}{P(P_2)} \prod_{j=1}^k (\lambda_j)^{P_2} (x_j)^{P_2-1} e^{-\lambda_j x_j}$$

$$\Rightarrow \frac{P(w_1)}{P(P_1)} \prod_{j=1}^k (\lambda_j)^{P_1} (x_j)^{P_1-1} > \frac{P(w_2)}{P(P_2)} \prod_{j=1}^k (\lambda_j)^{P_2} (x_j)^{P_2-1}$$

(b) when $|P_2 - P_1| = 1$, it's linear.

(c) when $P_1=4, P_2=2, C=2, k=4, \lambda_1=\lambda_3=1, \lambda_2=\lambda_4=2$.

$$\frac{P(w_1)}{P(P_1)} \cdot \prod_{j=1}^4 (\lambda_j)^{P_1} (x_j)^{P_1-1} > \frac{P(w_2)}{P(P_2)} \prod_{j=1}^4 (\lambda_j)^{P_2} (x_j)^{P_2-1}$$

$$\Rightarrow \frac{1}{3 \times 2 \times 1} \cdot \prod_{j=1}^4 (\lambda_j)^4 (x_j)^3 > \frac{1}{1} \prod_{j=1}^4 (\lambda_j)^2 (x_j)^1$$

$$\Rightarrow \frac{1}{6} \cdot 2^2 \cdot (0.1 \times 0.2 \times 0.3 \times 4)^3 > (0.1) \times 0.2 \times 0.3 \times 4.$$

$$\Rightarrow \frac{8}{3} \times (0.024)^2 > 1 \Rightarrow 0.015 < 1$$

$\Rightarrow w_2$ is the decision $C(x) = 2$.

$$c) P_1 = 3.2 \quad P_2 = 8, \quad C = 2, \quad \lambda_1 = 1 \quad k = 1.$$

$$\frac{P(W_1)}{P(P_1)} \cdot \prod_{j=1}^k (\lambda_j)^{P_1} (x_j)^{P_1-1} > \frac{P(W_2)}{P(P_2)} \prod_{j=1}^k (\lambda_j)^{P_2} (x_j)^{P_2-1}$$

$$\Rightarrow \frac{1}{P(3.2)} x \cdot x \cdot x^{2.2} > \frac{1}{P(8)} x \cdot x \cdot x^7.$$

$$\Rightarrow x^{4.8} = \frac{P(8)}{P(3.2)}$$

$$\Rightarrow x = \sqrt[4.8]{\frac{5040}{2 \cdot 423}} = 4.912.$$

$$\Rightarrow x^* = 4.912.$$

Type-1 error (false positive).

$$\Rightarrow T_1 = 1 - P(W_1 | X).$$

$$= 1 - \frac{P(X_1 | W_1) - P(X_2 | W_1) P(W_1)}{P(X)}.$$

$$= 1 - \frac{\frac{P(W_1)}{P(3.2)} x^{2.2}}{\frac{P(W_1)}{P(3.2)} x^{2.2} + \frac{P(W_2)}{P(8)} x^7}$$

$$= 1 - \frac{1}{1 + \frac{P(3.2)}{P(8)} x^{4.8}} = 1 - \frac{1}{2} = \frac{1}{2}$$

$$\Rightarrow \text{Type-1 error} = \frac{1}{2}$$

$$T_2 = 1 - P(W_2 | X)$$

$$= 1 - \frac{\frac{P(W_1)}{P(8)} x^7}{\frac{P(W_1)}{P(3,2)} x^{2,2} + \frac{P(W_2)}{P(8)} x^7}$$

$$= 1 - \frac{1}{\frac{P(8)}{P(3,2)} x^{-4,8}}$$

$$= 1 - \frac{1}{2}$$

$$= \frac{1}{2}$$

(e)

$$P_1 = P_2 = 4 \quad C = 2, \quad k = 2 \quad \lambda_1 = 8 \quad \lambda_2 = 0.3.$$

$$P(W_1) = \frac{1}{4}, \quad P(W_2) = \frac{3}{4}.$$

$$\frac{P(W_1)}{P(P_1)} \prod_{j=1}^2 (\lambda_j)^{P_1} (x_j)^{P_1-1} = \frac{P(W_2)}{P(P_2)} \prod_{j=1}^2 (\lambda_j)^{P_2} (x_j)^{P_2-1}$$

$$\Rightarrow \frac{1}{4} \cdot (x_1 x_2)^3 = \frac{3}{4} (x_1 x_2)$$

$$\Rightarrow 2(x_1 x_2)^3 = 0.$$

$$\therefore f(x_1, x_2) = 0 \quad \text{when} \quad x_1 x_2 = 0.$$

2. As we know $x_i | w_j \sim \text{lap}(m_{ij}, \lambda_i)$.

$$p_{x_i | w_j}(x_i | w_j) = \frac{\lambda_i}{2} e^{-\lambda_i |x_i - m_{ij}|}$$

$$p(w_1) = p(w_2) = \dots = p(w_j) \quad (j \in \{1, 2, \dots, C\})$$

$$\Rightarrow p(w_i | x) > p(w_j | x) \quad \forall j \neq i.$$

$$\Rightarrow \frac{p(x | w_i) p(w_i)}{p(x)} > \frac{p(x | w_j) p(w_j)}{p(x)}$$

$$\Rightarrow p(w_i) \cdot \prod_{j=1}^K p(x_j | w_i) > p(w_j) \cdot \prod_{j=1}^K p(x_j | w_j)$$

$$\Rightarrow \prod_{j=1}^K \left(\frac{\lambda_j}{2} \cdot e^{-\lambda_j |x_j - m_{ij}|} \right) > \prod_{j=1}^K \frac{\lambda_j}{2} e^{-\lambda_j |x_j - m_{jn}|}$$

$$\Rightarrow \sum_{j=1}^K [\lambda_j (|x_j - m_{ij}| - |x_j - m_{jn}|)] > 0 \quad \forall n \neq i.$$

\Rightarrow it's the weighted Manhattan Distance classifier.

the Min Manhattan Distance classifier when $m_{ji} = m_{jn}$.

(3)

$$(a) R(a_i | x) = \sum_{j=1}^4 \lambda(a_i | w_j) p(w_j | x)$$

$$\Rightarrow R(a_1 | x_1) = \sum_{j=1}^4 \lambda(a_1 | w_j) p(w_j | x_1).$$

$$= \sum_{j=1}^4 \lambda(a_1 | w_j) \frac{p(x_1 | w_j) \cdot p(w_j)}{p(x_1)}.$$

$$\therefore p(x_1) = \sum_{j=1}^4 p(x_1 | w_j) \cdot p(w_j) = 0.2967.$$

$$\begin{aligned} \Rightarrow R(a_1 | x_1) &= \frac{1}{0.2967} \cdot \left(0 \times \frac{1}{3} \times \frac{1}{10} + 2 \times \frac{1}{2} \times \frac{1}{8} + 3 \times \frac{1}{6} \times \frac{1}{2} + 4 \times \frac{2}{3} \times \frac{1}{3} \right) \\ &= \frac{0.77}{0.2967} = 2.5952. \end{aligned}$$

$$\begin{aligned}
 R(a_2|x_1) &= \frac{1}{P(x_1)} \cdot \sum_{j=1}^4 (\lambda(a_2|w_j) \cdot P(x_1|w_j) P(w_j)) \\
 &= \frac{1}{0.2917} \cdot (1 \times \frac{1}{10} \times \frac{1}{3} + 0 \times \frac{1}{2} \times \frac{1}{5} + 1 \times \frac{1}{6} \times \frac{1}{2} + 8 \times \frac{2}{5} \times \frac{1}{5}) \\
 &= 2.5506
 \end{aligned}$$

$$\Rightarrow R(a_3|x_1) = 1.5506$$

$$\Rightarrow R(a_4|x_1) = 1.8539$$

\therefore when $x=x_1$, the minimum is $R(a_3|x_1) = 1.5506$.

$$P(x_2) = \sum_{j=1}^4 P(x_2|w_j) \cdot P(w_j) = 0.33$$

$$\begin{aligned}
 R(a_1|x_2) &= \frac{1}{P(x_2)} \cdot \sum_{j=1}^4 (\lambda(a_1|w_j) \cdot P(x_2|w_j) P(w_j)) \\
 &= \frac{1}{0.33} \cdot (\frac{1}{10} \times \frac{1}{3} \times 0 + \frac{1}{5} \times \frac{1}{6} \times 2 + \frac{1}{2} \times \frac{1}{3} \times 3 + \frac{1}{5} \times \frac{2}{5} \times 0) \\
 &= 2.7879.
 \end{aligned}$$

$$R(a_2|x_2) = 2.5955, \quad R(a_3|x_2) = 1.0909, \quad R(a_4|x_2) = 1.4646$$

\Rightarrow when $x=x_2$, the minimum is $R(a_3|x_2) = 1.0909$.

$$P(x_3) = \sum_{j=1}^4 P(x_3|w_j) \cdot P(w_j) = 0.3733$$

$$\begin{aligned}
 R(a_1|x_3) &= \frac{1}{P(x_3)} \cdot \sum_{j=1}^4 (\lambda(a_1|w_j) \cdot P(x_3|w_j) P(w_j)) \\
 &= \frac{1}{0.3733} \cdot (\frac{1}{10} \times \frac{1}{3} \times 0 + \frac{1}{5} \times \frac{1}{6} \times 2 + \frac{1}{2} \times \frac{1}{2} \times 3 + \frac{1}{5} \times \frac{1}{5} \times 4)
 \end{aligned}$$

$$= 2.2054$$

$$R(a_2|x_3) = 1.6161$$

$$R(a_3|x_3) = 0.7500$$

$$R(a_4|x_3) = 1.5179$$

when $x=x_3$, the minimum is $R(a_3|x_3) = 0.7500$

(b)

$$R = \sum_{i=1}^3 R(a(x_i)|x_i) \cdot P(x_i)$$

$$= R(a(x_1)|x_1) P(x_1) + R(a(x_2)|x_2) P(x_2) + R(a(x_3)|x_3) P(x_3)$$

$$= 0.2967 \times 1.5506 + 0.33 \times 1.0909 + 0.3733 \times 0.75$$

$$\approx 1.1$$

4.

(a) w_1 = evade taxes.

w_2 = x evade taxes.

$$P(w_1) = \frac{3}{10} \quad P(w_2) = \frac{7}{10}$$

(b) conditional Gaussianity.

$$P(x|w_i) = N(\mu_i, \sigma_i^2) = \frac{1}{\sqrt{2\pi}\sigma_i} e^{-\frac{1}{2} \left(\frac{x-\mu_i}{\sigma_i}\right)^2}$$

$$\mu_1 = \frac{88+10+85}{3} = 87.67 \quad \sigma_1^2 = \frac{\sum (x-\mu_1)^2}{n-1} = 4.22 \quad \sigma_1 = 2.5166$$

$$\mu_2 = \frac{122+77+106+20+72+117+60}{7} = 109.143$$

$$\sigma_z^2 = 2539.5 \quad b_z = 50.3932$$

$$\Rightarrow P(x|w_1) = \frac{1}{4.3082} \cdot e^{-\frac{1}{2} \left(\frac{x-87.67}{2.5166} \right)^2}$$

$$P(x|w_2) = \frac{1}{126.317} \cdot e^{-\frac{1}{2} \left(\frac{x-102.143}{52.3932} \right)^2}$$

$x \rightarrow \text{income.}$

(C) $x_1 = \{ \text{refund, No refund} \}$

$$P(x_1 = \text{refund} | w_1) = 0$$

$$P(x_1 = \text{refund} | w_2) = \frac{3}{5}$$

$$P(x_1 = \text{No refund} | w_1) = 1$$

$$P(x_1 = \text{No, refund} | w_2) = \frac{4}{5}$$

$x_2 = \{ \text{single, married, divorced} \}$

$$P(x_2 = \text{single} | w_1) = \frac{2}{3}$$

$$P(x_2 = \text{single} | w_2) = \frac{2}{5}$$

$$P(x_2 = \text{married} | w_1) = 0$$

$$P(x_2 = \text{married} | w_2) = \frac{4}{5}$$

$$P(x_2 = \text{Divorced} | w_1) = \frac{1}{3}$$

$$P(x_2 = \text{Divorced} | w_2) = \frac{1}{5}$$

We know that a certain amount of samples is required to get an estimate distribution with confidence $1-\alpha$.

But, it is the most accurate estimate we can make without upscaling the 'evade taxes' or downscaling the.

'not evade taxes'

(d), No

As we know Laplace correction corrects for this by not allowing each conditional probability = 0

But

$$P(x|w_i) = P(x_1|w_i) P(x_2|w_i) \cdots P(x_k|w_i) P(w_i)$$

if one of this conditional probability = 0, then ^{the} entire expression is 0.

it's distorted.

(e) Minimum error rate

$$P(w_1|x) > P(w_2|x)$$

$$\Rightarrow \frac{\sum_{i=1}^K P(x_i|w_1) P(w_1)}{P(x)} > \frac{\sum_{i=1}^K P(x_i|w_2) P(w_2)}{P(x)}$$

$$\Rightarrow P(w_1) \cdot \sum_{i=1}^K P(x_i|w_1) > P(w_2) \cdot \sum_{i=1}^K P(x_i|w_2)$$

$$\Rightarrow P(w_1) P(x_1|w_1) \cdot P(x_2|w_1) \cdot P(x_3|w_1) > P(w_2) \cdot P(x_1|w_2) \cdot P(x_2|w_2) P(x_3|w_2)$$

$$P(x_1 = \text{refund} | w_1) = \frac{1}{5} \quad P(x_1 = \text{refund} | w_2) = \frac{4}{9}$$

$$P(x_1 = \text{No refund} | w_1) = \frac{4}{5} \quad P(x_1 = \text{No refund} | w_2) = \frac{5}{9}$$

$$P(x_2 = \text{single} | w_1) = \frac{1}{2} \quad P(x_2 = \text{single} | w_2) = \frac{3}{10}$$

$$P(x_2 = \text{Married} | w_1) = \frac{1}{6} \quad P(x_2 = \text{Married} | w_2) = \frac{2}{2}$$

$$P(x_2 = \text{Divorced} | w_1) = \frac{1}{3} \quad P(x_2 = \text{Divorced} | w_2) = \frac{1}{5}$$

$$P(x_3|w_1) = \frac{1}{4.3082} \cdot e^{-\frac{1}{2} \left(\frac{x-87.67}{2.5166} \right)^2}$$

$$P(x_2|w_2) = \frac{1}{126.317} \cdot e^{-\frac{1}{2} \left(\frac{x-109.143}{57.3932} \right)^2}$$