

$$1) \quad p_{x_j | w_i} = \frac{1}{\Gamma(p_i)} \lambda_j^{p_i} x_j^{p_i-1} e^{-\lambda_j x_j} \quad p_i, \lambda_j > 0$$

$$p(w_i | x_1 = x_1, \dots, x_k = x_k) = \frac{p(x_1 = x_1, \dots, x_k = x_k | w_i)}{p(x_1, x_2, w_j)}$$

$$\propto p(x_1 | w_i) p(x_2 | w_i) \dots p(x_k | w_i) p(w_i)$$

$$= \prod_{j=1}^k \frac{1}{\Gamma(p_i)} \lambda_j^{p_i} x_j^{p_i-1} e^{-\lambda_j x_j} p(w_i)$$

$$= \frac{p(w_i)}{\Gamma(p_i)} \prod_{j=1}^k (\lambda_j)^{p_i} (x_j)^{p_i-1} e^{-\lambda_j x_j}$$

$$\frac{p(w_1)}{\Gamma(p_1)} \prod_{j=1}^k (\lambda_j)^{p_1} (x_j)^{p_1-1} e^{-\lambda_j x_j} >$$

$$\frac{p(w_2)}{\Gamma(p_2)} \prod_{j=1}^k (\lambda_j)^{p_2} (x_j)^{p_2-1} e^{-\lambda_j x_j}$$

$$\boxed{\frac{p(w_1)}{\Gamma(p_1)} \prod_{j=1}^k (\lambda_j)^{p_1} (x_j)^{p_1-1} > \frac{p(w_2)}{\Gamma(p_2)} \prod_{j=1}^k (\lambda_j)^{p_2} (x_j)^{p_2-1}}$$

Bayes Optimal Classifier

b. Linear when $|p_2 - p_1| = 1$

$$c, p_1=4 \quad c=2 \quad \lambda_1=\lambda_3=1 \quad x=(0.1, 0.2, 0.3, 4) \\ p_2=2 \quad k=4 \quad \lambda_2=\lambda_4=2$$

$$\frac{P(\cancel{w_1})}{\Gamma(p_1)} \prod_{j=1}^k (\lambda_j)^{p_1} (x_j)^{p_1-1} \stackrel{w_1}{>} \frac{P(\cancel{w_2})}{\Gamma(p_2)} \prod_{j=1}^k (\lambda_j)^{p_2} (x_j)^{p_2-1}$$

$$\frac{1}{6} (4)^4 [(0.1)(0.2)(0.3)(4)]^2 \stackrel{w_1}{>} \frac{1}{1} (\cancel{4})^2 [(\cancel{0.1})(\cancel{0.2})(\cancel{0.3})(4)]^1$$

$$\frac{16}{6} (0.024)^2 \stackrel{w_1}{>} 1 \quad \stackrel{w_2}{<} \quad \rightarrow 0.0015 < 1 \rightarrow w_2$$

$$C(\underline{x}) = 2$$

$$d. p_1=3.2 \quad c=2 \quad \lambda_1=1 \\ p_2=8 \quad k=1$$

$$\frac{P(\cancel{w_1})}{\Gamma(p_1)} \prod_{j=1}^k (\lambda_j)^{p_1} (x_j)^{p_1-1} \stackrel{w_1}{>} \frac{P(\cancel{w_2})}{\Gamma(p_2)} \prod_{j=1}^k (\lambda_j)^{p_2} (x_j)^{p_2-1}$$

$$\frac{1}{\Gamma(3.2)} (\cancel{1})^{3.2} x^{2.2} \stackrel{w_1}{>} \frac{1}{\Gamma(8)} (\cancel{1})^8 (x)^7$$

$$\frac{\Gamma(8)}{\Gamma(3.2)} = x^{4.8} \quad x = \sqrt[4.8]{\frac{5040}{2.423}} = 4.912$$

$$x^* = 4.912$$

d. Type-1 Error (False Positive)

GUESS

	w_1	w_2
TRUE w_1	correct	T 1
w_2	T 2	correct

$$T_1 = 1 - P(w_1 | x) \quad \leftarrow \text{using zero-one loss function}$$

$$= 1 - \frac{P(x, w_1) - P(x, w_2) P(w_1)}{P(x)} = \frac{\Gamma(8)}{\Gamma(3.2)}$$

$$= 1 - \frac{\frac{P(w_1)}{P(3.2)} x^{2.2}}{\frac{P(w_1)}{P(3.2)} x^{2.2} + \frac{P(w_2)}{\Gamma(8)} x^7} = 1 - \frac{1}{1 + \frac{\Gamma(3.2)}{\Gamma(8)} x^{4.8}} = 1 - \frac{1}{1+1} = \frac{1}{2}$$

$T-1 \text{ Error} = \frac{1}{2}$

$$T_2 = 1 - P(w_2 | x)$$

$$= 1 - \frac{\frac{P(w_2)}{\Gamma(8)} x^7}{\frac{P(w_1)}{P(3.2)} x^{2.2} + \frac{P(w_2)}{\Gamma(8)} x^7} = 1 - \frac{1}{\frac{\Gamma(8)}{\Gamma(3.2)} x^{4.8} + 1} = 1 - \frac{1}{2}$$

$T-2 \text{ Error} = \frac{1}{2}$

$$e. \quad p_1 = p_2 = 4 \quad k=2$$

$$c=2 \quad \lambda_1 = 8 \quad \lambda_2 = 0.3$$

$$P(w_1) = 1/4 \quad P(w_2) = 3/4$$

$$\frac{P(w_1)}{\cancel{\Gamma(p_1)}} \prod_{j=1}^k (\lambda_j)^{p_1} (x_j)^{p_1-1} = \frac{P(w_2)}{\cancel{\Gamma(p_2)}} \prod_{j=1}^k (\lambda_j)^{p_2} (x_j)^{p_2-1}$$

$$\left(\frac{1}{4}\right) [(\cancel{8})(\cancel{0.3})]^4 (x_1, x_2)^3 = \frac{3}{4} [(\cancel{8})(\cancel{0.3})]^4 (x_1, x_2)^3$$

$$2(x_1, x_2)^3 = 0$$

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

$$2) \quad X_i | w_j \sim \text{Lap}(m_{ji}, \lambda_i)$$

$$p(x_i | w_j) = \frac{\lambda_i}{2} e^{-\lambda_i |x_i - m_{ji}|}, \quad \lambda_i > 0 \quad \begin{matrix} i \in \{1, \dots, k\} \\ j \in \{1, \dots, c\} \end{matrix}$$

$$p(w_1) = p(w_2) = \dots = p(w_c)$$

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

$$\frac{p(x | w_i) p(w_i)}{\cancel{p(x)}} > \frac{p(x | w_j) p(w_j)}{\cancel{p(x)}} \quad \forall j \neq i$$

$$p(w_i) \prod_{j=1}^k p(x_j | w_i) > p(w_\eta) \prod_{j=1}^k p(x_j | w_\eta) \quad \forall \eta \neq i$$

$$\cancel{p(w_i)} \prod_{j=1}^k \left[\cancel{\frac{\lambda_j}{2}} e^{-\lambda_j |x_j - m_{ji}|} \right] > \cancel{p(w_\eta)} \prod_{j=1}^k \left[\cancel{\frac{\lambda_j}{2}} e^{-\lambda_j |x_j - m_{j\eta}|} \right]$$

$$\sum_{j=1}^k -\lambda_j |x_j - m_{ji}| \cancel{\ln(e)} > \sum_{j=1}^k -\lambda_j |x_j - m_{j\eta}|$$

$$\sum_{j=1}^k \left[\lambda_j (|x_j - m_{ji}| - |x_j - m_{j\eta}|) \right] > 0 \quad \forall \eta \neq i$$

↑ weighted Manhattan Distance classifier

↑
= min Manhattan Distance classifier when $m_{ji} = m_{j\eta}$

3) a.

$$P(\alpha_i | x) = \sum_{j=1}^4 \lambda(\alpha_i | w_j) p(w_j | x)$$

$$\begin{aligned} p(x_1) &= p(x_1 | w_1) p(w_1) + \dots + p(x_1 | w_4) p(w_4) \\ &= \frac{1}{3} \left(\frac{1}{10} \right) + \frac{1}{2} \left(\frac{1}{5} \right) + \frac{1}{6} \left(\frac{1}{2} \right) + \frac{2}{5} \left(\frac{1}{5} \right) \end{aligned}$$

$$p(x_1) = 0.2967$$

$$p(x_2) = \frac{1}{3} \left(\frac{1}{10} \right) + \frac{1}{4} \left(\frac{1}{5} \right) + \frac{1}{3} \left(\frac{1}{2} \right) + \frac{2}{5} \left(\frac{1}{5} \right)$$

$$p(x_2) = 0.33$$

$$p(x_3) = \frac{1}{3} \left(\frac{1}{10} \right) + \frac{1}{4} \left(\frac{1}{5} \right) + \frac{1}{2} \left(\frac{1}{2} \right) + \frac{1}{5} \left(\frac{1}{5} \right)$$

$$p(x_3) = 0.3733$$

$$\begin{aligned} R(\alpha_i | x_1) &= \frac{1}{p(x_1)} \sum_{j=1}^4 [\lambda(\alpha_i | w_j) p(x_1 | w_j) p(w_j)] \\ &= \frac{1}{0.2967} \left(0 \left(\frac{1}{3} \right) \left(\frac{1}{10} \right) + \left(\frac{1}{2} \right) \left(\frac{1}{5} \right) + \left(\frac{1}{6} \right) \left(\frac{1}{2} \right) + \left(\frac{2}{5} \right) \left(\frac{1}{5} \right) \right) \end{aligned}$$

$$= \frac{0.77}{0.2967} = 2.5952 = R(\alpha_i | x_1)$$

$$R(\alpha_2 | x_1) = 2.5506$$

$$R(\alpha_3 | x_1) = 1.5506 \leftarrow \text{min for } x_1$$

$$R(\alpha_4 | x_1) = 1.8539$$

$$E(\alpha_1 | x_2) = 2.7079$$

$$E(\alpha_2 | x_2) = 2.5455$$

$$E(\alpha_3 | x_2) = \boxed{1.0909}$$

$$E(\alpha_4 | x_2) = 1.4646$$

$$E(\alpha_1 | x_1) = 2.5955$$

$$E(\alpha_2 | x_1) = 2.5506$$

$$E(\alpha_3 | x_1) = \boxed{1.5506}$$

$$E(\alpha_4 | x_1) = 1.8539$$

$$E(\alpha_1 | x_3) = 2.7054$$

$$E(\alpha_2 | x_3) = 1.6161$$

$$E(\alpha_3 | x_3) = \boxed{0.7500}$$

$$E(\alpha_4 | x_3) = 1.5179$$

$$b. \quad E = E(\alpha(x_1) | x_1) p(x_1) + \dots + E(\alpha(x_3) | x_3) p(x_3)$$

$$= 1.5506(0.2967) + 1.0909(0.33) + 0.75(0.3733)$$

$$\boxed{E = 1.1}$$

4) w_1 = evades taxes

w_2 = doesn't evade taxes

$$a. \boxed{p(w_1) = 3/10 \quad p(w_2) = 7/10}$$

b. Conditional Gaussianity

$$p(x|w_i) = N(\mu_i, \sigma_i^2)$$

$$= \frac{1}{\sqrt{2\pi} \sigma_i} \exp\left[-\frac{1}{2} \left(\frac{x - \mu_i}{\sigma_i}\right)^2\right] \quad x \in \mathbb{R}$$

$$\mu_1 = \frac{88 + 90 + 85}{3} = 87.67$$

$$\sigma_1^2 = \frac{\sum (x - \mu_i)^2}{n - 1} = 6.33 \quad \sigma_1 = 2.5166$$

$$\mu_2 = \frac{122 + 77 + 106 + 210 + 72 + 117 + 60}{7} = 109.143$$

$$\sigma_2^2 = 2539.5 \quad \sigma_2 = 50.3932$$

$$p(x|w_1) = \frac{1}{6.3082} \exp\left[-\frac{1}{2} \left(\frac{x - 87.67}{2.5166}\right)^2\right]$$

$$p(x|w_2) = \frac{1}{126.317} \exp\left[-\frac{1}{2} \left(\frac{x - 109.143}{50.3932}\right)^2\right]$$

$x \rightarrow$ income

c. $X_1 \rightarrow \{\text{refund, no refund}\}$

$$P(X_1 = \text{refund} | w_1) = 0 \quad P(X_1 = \text{R} | w_2) = 3/7$$

$$P(X_1 = \text{NR} | w_1) = 1 \quad P(X_1 = \text{NR} | w_2) = 4/7$$

$X_2 \rightarrow \{\text{Single, Married, Divorced}\}$

$$P(X_2 = S | w_1) = 2/3$$

$$P(X_2 = S | w_2) = 2/7$$

$$P(X_2 = M | w_1) = 0$$

$$P(X_2 = M | w_2) = 4/7$$

$$P(X_2 = D | w_1) = 1/3$$

$$P(X_2 = D | w_2) = 1/7$$

↑ The SLN says that a certain amount of samples (n) is required to get an estimated distribution w/ confidence $1 - \alpha$.

However, this is the most accurate estimate we can make w/o upscaling the 'evades taxes' group & downscaling the 'doesn't evade taxes' group.

d. Assuming these features are conditionally independent

$$P(\underline{x} | w_i) = P(x_1 | w_i) P(x_2 | w_i) \dots P(x_k | w_i) P(w_i)$$

↑ If one of these conditional probs = 0, then the entire expression is 0. Laplace correction corrects for this by not allowing each conditional prob to = 0.

e. Minimum Error Rate

$$P(w_1 | x) \stackrel{w_1}{>} P(w_2 | x)$$

$$\frac{\left[\prod_{i=1}^k P(x_i | w_1) \right] P(w_1)}{P(x)} \stackrel{w_1}{>} \frac{\left[\prod_{i=1}^k P(x_i | w_2) \right] P(w_2)}{P(x)}$$

$$P(w_1) \prod_{i=1}^k P(x_i | w_1) - P(w_2) \prod_{i=1}^k P(x_i | w_2) \stackrel{w_1}{>} 0$$

$$P(w_1) P(x_1 | w_1) P(x_2 | w_1) P(x_3 | w_1) \stackrel{w_1}{>} P(w_2) P(x_1 | w_2) P(x_2 | w_2) P(x_3 | w_2)$$

where $x_1 = \text{refund}$, $x_2 = \text{Marital Status}$, $x_3 = \text{Income}$

$$P(x_1 = R | w_1) = \frac{0+1}{3+2} = \frac{1}{5} \quad P(x_1 = R | w_2) = \frac{3+1}{7+2} = \frac{4}{9}$$

$$P(x_1 = NR | w_1) = \frac{3+1}{3+2} = \frac{4}{5} \quad P(x_1 = NR | w_2) = \frac{4+1}{7+2} = \frac{5}{9}$$

$$P(x_2 = S | w_1) = \frac{2+1}{3+3} = \frac{1}{2} \quad P(x_2 = S | w_2) = \frac{2+1}{7+3} = \frac{3}{10}$$

$$P(x_2 = M | w_1) = \frac{0+1}{3+3} = \frac{1}{6} \quad P(x_2 = M | w_2) = \frac{4+1}{7+3} = \frac{1}{2}$$

$$P(x_2 = D | w_1) = \frac{1+1}{3+3} = \frac{1}{3} \quad P(x_2 = D | w_2) = \frac{1+1}{7+3} = \frac{1}{5}$$

$$P(x_3 | w_1) = \frac{1}{6.3082} \exp \left[-\frac{1}{2} \left(\frac{x - 87.67}{2.5166} \right)^2 \right]$$

$$P(x_3 | w_2) = \frac{1}{126.317} \exp \left[-\frac{1}{2} \left(\frac{x - 109.143}{50.3932} \right)^2 \right]$$

5) Notes:

Gaussian class condition

$$N(\mu, \sigma) = \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{(x-\mu)^2}{2\sigma^2}}$$