

4)

(a)

Prior class probabilities of "Yes" class

$$= \frac{\text{no. of Yes classes}}{\text{total no. of classes (Yes & No)}} = \frac{3}{10}$$

Prior class probability of "No"

$$= 7/10$$

(b) To find: $P(x|w_i)$ using MLE for continuous features.
 here, Taxable Income is a continuous feature.

To find: MLE estimate
 to compute sample mean & sample variance for both classes.

↳ Sample mean for class Yes = $\frac{1}{n} \sum \text{Taxable income for Yes}$

$$\mu_{\text{Yes}} = \frac{88 + 90 + 85}{3} = 87.67$$

↳ Sample variance for class "Yes" $\frac{1}{n} \sum (x - \mu_{\text{Yes}})^2$

$$\begin{aligned} \sigma_{\text{Yes}}^2 &= \frac{(88 - 87.67)^2 + (90 - 87.67)^2 + (85 - 87.67)^2}{3} \\ &= \frac{12.66}{3} = 4.222 \end{aligned}$$

∴ For class Yes, pdf is given by

$$P(x|\text{Yes}) = N(\mu_{\text{Yes}}, \sigma_{\text{Yes}}^2)$$

$$x = \text{taxable income} = N(87.67, 4.222)$$

Sample mean for class "No"

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

$$\mu_{No} = 109.14$$

Sample variance for class "No"

$$\sigma_{No}^2 = (122 - 109.14)^2 + (77 - 109.14)^2 + (106 - 109.14)^2 + (210 - 109.14)^2 + (72 - 109.14)^2 + (117 - 109.14)^2 + (60 - 109.14)^2$$

$$\sigma_{No}^2 = \frac{23465.41}{7} = \frac{15236.85}{7}$$

$$= 2176.69$$

\therefore pdf for $P(x|No)$

$$= P(x|w_i = No) = N(109.14, 2176.69)$$

(c) discrete features: Refund and Marital Status.

to calculate pmf as:

$$P_{X|w_i}(x_j|w_i) = P(X=x_j|w_i) = \frac{n_{ji}}{n_i}$$

n_{ji} = no. of instances in class w_i for j feature.

n_i = no. of instances in class w_i

for Refund feature: let $j = \text{Refund}$, let Yes ~~class~~ class = $w_1 = i = 1$

\therefore Ans

for Refund = Yes & evade = Yes

$$n_{ji} = 0$$

$$n_i = 3$$

$$\therefore P_{X|w_i}(x = \text{Refund} = \text{yes} | w_i = \text{yes}) = 0$$

For w_i ,
refund = yes evade = no

$$P_{x|w_i}(\text{yes}, \text{no}) = 3/7$$

refund = no evade = yes

$$P_{x|w_i}(\text{no}, \text{no}) = 3/3 = 1$$

refund = no evade = no

$$P_{x|w_i}(\text{no}, \text{no}) = 4/7$$

For feature: marital status:

$$P_{x|w_i}(\text{Single}, \text{yes}) = 2/3$$

$$P_{x|w_i}(\text{Single}, \text{no}) = 2/7$$

$$P_{x|w_i}(\text{married}, w_i = \text{yes}) = 0/3 = 0$$

$$P_{x|w_i}(\text{married}, w_i = \text{no}) = 4/7$$

$$P_{x|w_i}(\text{divorced}, \text{yes}) = 1/3$$

$$P_{x|w_i}(\text{divorced}, \text{no}) = 1/7$$

With the above estimation, we see that few have $P_{mf} = 0$, this can be problematic, since all the features that belong to feature refund = yes will get classified as evade = yes & all the features with married will get classified as evade = no. in the test set

(d)

Laplace correction helps avoid zero probabilities by adding 1 to each count and normalising by the total count + no. of possible levels l .

Ex: for feature: Refund

$$P_{x|w_i}(\text{refund}=\text{yes} | \text{crade}=\text{yes}) = \frac{(n_{ij}+1)}{n_i+l} = \frac{0+1}{3+2} = 1/5$$

$$P_{x|w_i}(\text{refund}=\text{yes} | \text{crade}=\text{no}) = 4/7+2 = 4/9$$

$$P_{x|w_i}(\text{refund}=\text{no} | \text{crade}=\text{yes}) = \frac{3+1}{3+2} = 4/5$$

$$P_{x|w_i}(\text{refund}=\text{no} | \text{crade}=\text{no}) = \frac{4+1}{7+2} = 5/9$$

for feature: Marital Status

$$P_{x|w_i}(\text{married} | \text{yes}) = \frac{1}{3+3} = 1/6$$

$$P_{x|w_i}(\text{married} | \text{no}) = \frac{5}{9}$$

$$P_{x|w_i}(\text{single} | \text{yes}) = 3/6 = 1/2$$

$$P_{x|w_i}(\text{single} | \text{no}) = 2/6 = 1/3$$

$$P_{x|w_i}(\text{divorced} | \text{yes}) = 2/6 = 1/3$$

$$P_{x|w_i}(\text{divorced} | \text{no}) = 2/9$$

This can be a valid estimate, since we have not encountered any '0' probabilities.

(c) Minimum error rate Decision Rule using Laplace correction.

$$P(w_i|x) = \frac{P(x|w_i) P(w_i)}{P(x)} \quad \text{---(i)}$$

$$\text{Laplace correction: } P(x_j|w_i) = \frac{n_{ji}+1}{n_i+l}$$

$$P(x|w_i) = \prod_{j=1}^K \frac{n_{ji}+1}{n_i+l} \quad \begin{array}{l} j = \text{features} \\ i = \text{class} \end{array} \quad \text{. here, 2 features}$$

this can be simplified to, taking log on (i)

$$\log P(w_i|x) = \log P(w_i) + \sum_{j=1}^K \frac{n_{ji}+1}{n_i+l}$$

\therefore decision rule, becomes

$$w = \arg \max_i \left\{ \log P(w_i) + \sum_{j=1}^K \frac{n_{ji}+1}{n_i+l} \right\}$$