

SENG 474 - Assignment 1

Dennop Banerjee
V00R37868

1. a) Entropy = $\sum_{i=0}^n P(x_i) \cdot \log_2 P(x_i)$

① Age -

• $P(\text{young}) = \frac{1}{3}$ • $P(\text{pre-presbyopic}) = \frac{1}{3}$ • $P(\text{presbyopic}) = \frac{1}{3}$

② Spectacle prescription -

• $P(\text{myope}) = \frac{1}{2}$ • $P(\text{hypermetrope}) = \frac{1}{2}$

③ Astigmatism -

• $P(\text{no}) = \frac{1}{2}$ • $P(\text{yes}) = \frac{1}{2}$

④ Tear-prod-rate -

• $P(\text{reduced}) = \frac{1}{2}$ • $P(\text{normal}) = \frac{1}{2}$

⑤ Contact Lenses -

• $P(\text{soft}) = \frac{5}{24}$ • $P(\text{hard}) = \frac{1}{6}$ • $P(\text{none}) = \frac{5}{8}$

① young • none

	Soft	Hard	None	
① Young	$\frac{1}{4}$	$\frac{1}{4}$	$\frac{1}{2}$	$-\frac{1}{4} \log_2(\frac{1}{4}) - \frac{1}{4} \log_2(\frac{1}{4}) - \frac{1}{2} \log_2(\frac{1}{2}) = 1.5$
② Pre-presby.	$\frac{1}{4}$	$\frac{1}{8}$	$\frac{5}{8}$	$-\frac{1}{4} \log_2(\frac{1}{4}) - \frac{1}{8} \log_2(\frac{1}{8}) - \frac{5}{8} \log_2(\frac{5}{8}) = 1.3$
Presbyopic	$\frac{1}{8}$	$\frac{1}{8}$	$\frac{3}{4}$	$= 1.06$
③ Myope	$\frac{1}{6}$	$\frac{1}{4}$	$\frac{1}{2}$	$= 1.38$
Hypermetrope	$\frac{1}{4}$	$\frac{1}{2}$	$\frac{2}{3}$	$= 1.19$
④ No	$\frac{5}{12}$	$\frac{0}{12}$	$\frac{7}{12}$	$= 0.98$
Yes	$\frac{0}{12}$	$\frac{1}{3}$	$\frac{2}{3}$	$= 0.92$
⑤ Reduced	$\frac{0}{12}$	$\frac{0}{12}$	$\frac{12}{12}$	$= 0$
Normal.	$\frac{5}{12}$	$\frac{1}{3}$	$\frac{1}{4}$	$= 1.55$

$$\text{Expected Age} = (1.5)(\frac{1}{3}) + (1.3)(\frac{1}{3}) + (1.06)(\frac{1}{3}) = 1.29$$

$$\text{Expected Specs} = (1.38)(\frac{1}{2}) + (1.18)(\frac{1}{2}) = 1.28$$

$$\text{Expected Astigmatism} = (0.98)(\frac{1}{2}) + (0.92)(\frac{1}{2}) = 0.95$$

$$\text{Expected Tears} = (0)(\frac{1}{2}) + (1.55)(\frac{1}{2}) = 0.78$$

Lvl 1	Soft	Hard	None
Young	$\frac{1}{2}$	$\frac{1}{2}$	$\frac{1}{4}$
Pre-presbyopic	$\frac{1}{2}$	$\frac{1}{4}$	$\frac{1}{4}$
Presbyopic	$\frac{1}{4}$	$\frac{1}{4}$	$\frac{1}{2}$
Myope	$\frac{1}{3}$	$\frac{1}{2}$	$\frac{1}{6}$
Hypermetrope	$\frac{1}{2}$	$\frac{1}{6}$	$\frac{1}{3}$
No	$\frac{5}{6}$	$\frac{0}{6}$	$\frac{1}{6}$
Yes	$\frac{0}{6}$	$\frac{2}{3}$	$\frac{1}{3}$
Reduced	-	-	-
Normal	-	-	-

$$-\frac{1}{2} \log_2(\frac{1}{2}) - \frac{1}{2} \log_2(\frac{1}{2}) = 1$$

$$- \dots = 1.5$$

$$= 1.5$$

$$= 1.46$$

$$= 1.46$$

$$= 0.65$$

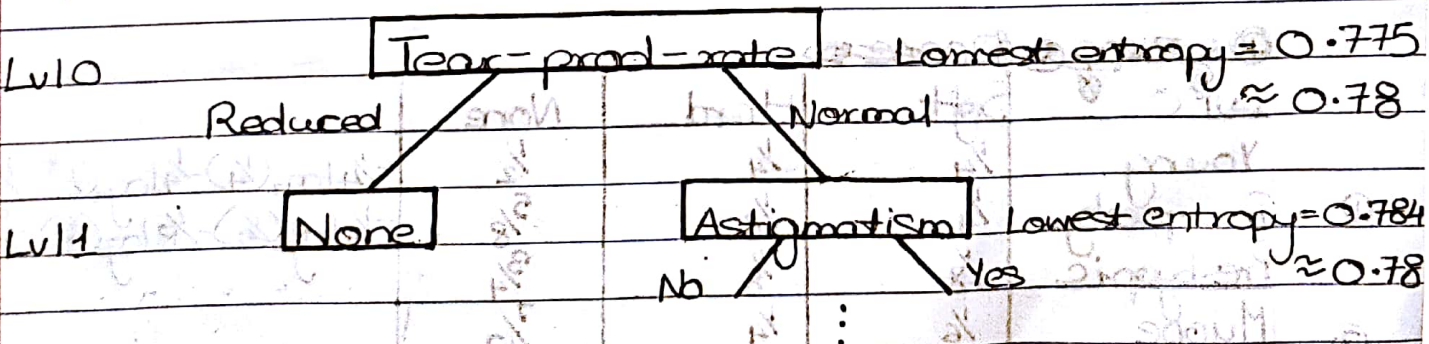
$$= 0.92$$

$$\text{Expected Age 1} = (1)(\frac{1}{3}) + (1.5)(\frac{1}{3}) + (1.5)(\frac{1}{3}) = 1.33$$

$$\text{Expected Specs 1} = (1.46)(\frac{1}{2}) + (1.46)(\frac{1}{2}) = 1.46$$

$$\text{Expected Astigmatism 1} = (0.65)(\frac{1}{2}) + (0.92)(\frac{1}{2}) = 0.78$$

~~Expected Tears 1~~



b) The difference in my calculated entropy values and the values obtained from Scikit can be explained using multiple reasons such as rounding errors, calculation errors, human

errors. All errors aside, computers and humans have different calculation algorithms (not necessarily, but computers are known to use more efficient algorithms which results in negligible errors in calculation).

2. a) $Pr(R) = 33.33\%$ $Pr(R)$ | $Pred(R)$ | $Pr(Y)$ | $Pred(Y)$ | $Pr(B)$ | $Pred(B)$
 $Pr(Y) = "$ $\frac{1}{3}$ | 0 | $\frac{1}{3}$ | 1 | $\frac{1}{3}$ | 1
 $Pr(B) = "$ $\Rightarrow (\frac{1}{3} \times 0) + (\frac{1}{3} \times 1) + (\frac{1}{3} \times 1) =$
 $\Rightarrow \frac{2}{3}$

\therefore Expected error rate = $\frac{2}{3}$

b) $Pr(R)$	$Pred(R)$	$Pr(Y)$	$Pred(Y)$	$Pr(B)$	$Pred(B)$
$\frac{1}{3}$	0.3	$\frac{1}{3}$	1	$\frac{1}{3}$	0.7

$Pred(R) = Pr(Y) \vee Pr(B) = 0 + 0.3 = 0.3$

$Pred(Y) = Pr(R) \vee Pr(B) = 0.7 + 0.3 = 1$

$Pred(B) = Pr(Y) \vee Pr(R) = 0 + 0.7 = 0.7$

$\Rightarrow (\frac{1}{3} \times 0.3) + (\frac{1}{3} \times 1) + (\frac{1}{3} \times 0.7)$
 $\Rightarrow \frac{1}{10} + \frac{1}{3} + \frac{7}{30}$
 $\Rightarrow \frac{2}{3}$

\therefore Expected error rate = $\frac{2}{3}$

c) $Pr(R)$	$Pred(R)$	$Pr(Y)$	$Pred(Y)$	$Pr(B)$	$Pred(B)$
$\frac{1}{2}$	0	$\frac{1}{4}$	1	$\frac{1}{4}$	1

$\Rightarrow (\frac{1}{2} \times 0) + (\frac{1}{4} \times 1) + (\frac{1}{4} \times 1)$

$\Rightarrow \frac{1}{2}$

\therefore Expected error rate = $\frac{1}{2}$

d) $P_X(R)$	$P_{\text{pred}}(1R)$	$P_X(Y)$	$P_{\text{pred}}(1Y)$	$P_X(B)$	$P_{\text{pred}}(1B)$
$\frac{1}{2}$	0.3	$\frac{1}{4}$	1	$\frac{1}{4}$	0.7

$$\Rightarrow \left(\frac{1}{2} \times 0.3\right) + \left(\frac{1}{4} \times 1\right) + \left(\frac{1}{4} \times 0.7\right)$$

$$\Rightarrow \frac{3}{20} + \frac{1}{4} + \frac{7}{40}$$

$$\Rightarrow \frac{23}{40}$$

$$\therefore \text{Expected error rate} = \frac{23}{40}$$

3. a) $\theta = P_X(X=T)$ $D = \{T, T, T, T, T, T, F, F, F\}$
 ~~$\theta = P_X(X=T)$~~ $\alpha_T = \#T$ $\alpha_F = \#F$

$$P(D) = P(TATATATATATAFFAFF) \\ = \theta^7 (1-\theta)^3$$

$$\log_e(P(D)) = \log_e(\theta^7 (1-\theta)^3) \\ = 7 \log_e \theta + 3 \log_e (1-\theta)$$

$$\frac{d \log_e(P(D))}{d\theta} = \frac{7}{\theta} + \left(-\frac{3}{1-\theta}\right)$$

$$\frac{7}{\theta} - \frac{3}{1-\theta} = 0$$

$$\Rightarrow \frac{7(1-\theta) - 3\theta}{\theta(1-\theta)} = 0 \quad \text{--- (x } \theta(1-\theta) \text{)}$$

$$\Rightarrow 7 - 7\theta - 3\theta = 0$$

$$\Rightarrow 7 - 10\theta = 0$$

$$\Rightarrow 10\theta = 7$$

$$\Rightarrow \theta = \frac{7}{10}$$

$$b) \theta^{(\beta_1-1)} \times (1-\theta)^{(\beta_2-1)} \quad \boxed{\theta = 1/2}$$

$$P(\theta) \propto \theta^{(\beta_1-1)} (1-\theta)^{(\beta_2-1)}$$

$$\Rightarrow P(\theta|D) = \theta^7 (1-\theta)^3 \theta^{(\beta_1-1)} (1-\theta)^{(\beta_2-1)}$$

$$\Rightarrow P(\theta|D) = \left(\frac{1}{2}\right)^7 \left(\frac{1}{2}\right)^3 \left(\frac{1}{2}\right)^3 \left(\frac{1}{2}\right)^5$$

$$\Rightarrow P(\theta|D) = \left(\frac{1}{2}\right)^8$$

$$\Rightarrow P(\theta|D) = \frac{1}{2^{18}}$$

$$\boxed{\beta_1 = 4}$$

$$\theta = \frac{\beta_1}{\beta_1 + \beta_2 + 1}$$

$$\Rightarrow \frac{1}{2} = \frac{4}{\beta_2 + 1 + 4 + 1}$$

$$\Rightarrow \frac{1}{2} = \frac{4}{\beta_2 + 2}$$

$$\Rightarrow \beta_2 + 2 = 8$$

$$\Rightarrow \boxed{\beta_2 = 6}$$

$$4. a) \hat{y}_i = w_0 + w_1 x_{i,1} + w_2 x_{i,2}^4$$

$$E(x) = \frac{1}{2N} \sum_{i=1}^N (y_i - \hat{y}_i)^2$$

$$= \frac{1}{2N} \sum_{i=1}^N (y_i - w_0 - w_1 x_{i,1} - w_2 x_{i,2}^4)^2$$

$$b) w_0 \leftarrow w_0 - K \frac{d}{dw_0} E(w_0, w_1, w_2)$$

$$= w_0 - K \frac{d}{dw_0} \left(\frac{1}{2N} \sum_{i=1}^N (y_i - w_0 - w_1 x_{i,1} - w_2 x_{i,2}^4)^2 \right)$$

$$= w_0 - K \frac{1}{2N} \sum_{i=1}^N \frac{d}{dw_0} (y_i - w_0 - w_1 x_{i,1} - w_2 x_{i,2}^4)^2$$

$$= w_0 + \frac{K}{N} \sum_{i=1}^N (y_i - w_0 - w_1 x_{i,1} - w_2 x_{i,2}^4)$$

$$c) w_1 \leftarrow w_1 - K \frac{d}{dw_1} E(w_0, w_1, w_2)$$

$$= w_1 - K \frac{d}{dw_1} \left(\frac{1}{2N} \sum_{i=1}^N (y_i - w_0 - w_1 x_{i,1} - w_2 x_{i,2}^4)^2 \right)$$

$$= w_1 + \frac{K}{N} \sum_{i=1}^N (y_i - w_0 - w_1 x_{i,1} - w_2 x_{i,2}^4) x_{i,1}$$

$$d) w_2 \leftarrow w_2 - K \frac{d}{dw_2} E(w_0, w_1, w_2)$$

$$= w_2 - K \frac{d}{dw_2} \left(\frac{1}{2N} \sum_{i=1}^N (y_i - w_0 - w_1 x_{i,1} - w_2 x_{i,2}^4)^2 \right)$$

$$= w_2 + \frac{K}{N} \sum_{i=1}^N (y_i - w_0 - w_1 x_{i,1} - w_2 x_{i,2}^4) x_{i,2}^4$$