

## Question 2

$$(a) \quad p(x = \text{apple}) = 0.2 \times \frac{3}{10} + 0.6 \times \frac{3}{10} + \frac{1}{2} \times 0.2 = 0.34$$

$$(b) \quad p(x = \text{orange}) = 0.2 \times \frac{4}{10} + 0.6 \times \frac{3}{10} + 0.2 \times \frac{1}{2} = 0.36$$

$$p = \frac{0.6 \times \frac{3}{10}}{0.36} = \frac{1}{2}$$

## Question 3

$$(a) \quad (1) \quad j \in \{1, \dots, N\}: \quad E[L] = \sum_{k=1}^N L_{kj} p(C_k | x)$$

$$= l_s \cdot \sum_{k \neq j} p(C_k | x) + 0 \cdot p(C_j | x)$$

$$= l_s \cdot (1 - p(C_j | x))$$

To make  $E[L]$  smallest, we wish  $p(C_j | x)$  is biggest

so we want  $p(C_j | x) \geq p(C_k | x)$  for any  $k$ ,  $x \rightarrow C_j$

$$(2) \quad j = N+1: \quad E[L] = \sum_{k=1}^N L_{kj} p(C_k | x)$$

$$= l_r \cdot \sum_{k=1}^N p(C_k | x)$$

$$= l_r$$

we get an fixed value  $l_r$  for  $j = N+1$

To make  $E[L]$  smallest, we wish  $l_s \cdot (1 - p(C_j|x)) \leq l_r$   
so the fixed value  $l_r$  is the upper bound

$$1 - p(C_j|x) \leq \frac{l_r}{l_s}$$

$$p(C_j|x) \geq 1 - \frac{l_r}{l_s}$$

This is the claim

(b) if  $l_r = 0$

we have  $\text{classify}(x) \rightarrow \begin{cases} C_j, \text{ iff } \forall k: p(C_j|x) \geq p(C_k|x) \\ \quad \wedge p(C_j|x) = 1 \\ C_{\text{rej}}, \text{ otherwise} \end{cases}$

so unless  $p(C_j|x) = 1$ , we always choose  $C_{\text{rej}}$

(c) if  $l_r > l_s$ , then  $p(C_j|x)$  always  $\geq 1 - \frac{l_r}{l_s}$

then we always get  $\text{classify}(x) \rightarrow C_j$

never  $C_{\text{rej}}$

### Question 4

$$p(x|\theta) = \theta^2 x \exp(-\theta x) g(x)$$

$$x_1, \dots, x_N > 0 \quad g(x) = 1$$

$$p(x|\theta) = \theta^2 x \exp(-\theta x)$$

$$L(\theta) = \prod_{n=1}^N p(x_n|\theta)$$

$$E(\theta) = -\sum_{n=1}^N \ln p(x_n|\theta) = -\sum_{n=1}^N \ln (\theta^2 x_n \exp(-\theta x_n))$$

$$\frac{\partial E(\theta)}{\partial \theta} = -\sum_{n=1}^N \frac{2\theta x_n \exp(-\theta x_n) + \theta^2 x_n \exp(-\theta x_n) (-x_n)}{\theta^2 x_n \exp(-\theta x_n)}$$

$$= -\sum_{n=1}^N \frac{2 - x_n \theta}{\theta} \stackrel{!}{=} 0$$

$$\Leftrightarrow 2N = \left(\sum_{n=1}^N x_n\right) \cdot \tilde{\theta}$$

$$\Leftrightarrow \tilde{\theta} = \frac{2N}{\sum_{n=1}^N x_n}$$