

# Assignment 8

$$Q1 \quad (a) P(X, C | \theta, \pi) = \prod_{i=1}^N P(X^{(i)}, C^{(i)})$$

$$P(X, C | \theta, \pi) = \prod_{i=1}^N \left[ P(C^{(i)} | \pi) \prod_{j=1}^{784} P(X_j^{(i)} | C^{(i)}, \theta_j c^{(i)}) \right]$$

$$\ell(X, C | \theta, \pi) = \log P(X, C | \theta, \pi) = \sum_{i=1}^N \log \left[ P(C^{(i)} | \pi) \prod_{j=1}^{784} P(X_j^{(i)} | C^{(i)}, \theta_j c^{(i)}) \right]$$

$$= \underbrace{\sum_{i=1}^N \log P(C^{(i)} | \pi)}_{A} + \underbrace{\sum_{j=1}^{784} \sum_{i=1}^N \log P(X_j^{(i)} | C^{(i)}, \theta_j c^{(i)})}_{B}$$

To maximize  $\ell(X, C | \theta, \pi)$ , we maximize A and B individually

$$\begin{aligned} \frac{\partial \ell(X, C | \theta, \pi)}{\partial \theta} &= \sum_{j=1}^{784} \sum_{i=1}^N \frac{\partial \log P(X_j^{(i)} | C^{(i)}, \theta_j c^{(i)})}{\partial \theta} && \theta_{jc}: \text{probability of obtaining } X_j=1 \\ &= \sum_{j=1}^{784} \sum_{i=1}^N \frac{\partial \log \left[ \theta_j c^{(i)} (1 - \theta_j c^{(i)})^{(1-X_j^{(i)})} \right]}{\partial \theta} && \text{from class } C \\ &= \sum_{j=1}^{784} \sum_{i=1}^N \frac{\partial \left[ X_j^{(i)} \log \theta_j c^{(i)} + (1 - X_j^{(i)}) \log (1 - \theta_j c^{(i)}) \right]}{\partial \theta} \\ &= \sum_{j=1}^{784} \sum_{i=1}^N \left[ X_j^{(i)} \frac{\partial \log \theta_j c^{(i)}}{\partial \theta} + (1 - X_j^{(i)}) \frac{\partial (1 - \theta_j c^{(i)})}{\partial \theta} \right] \\ &= \sum_{j=1}^{784} \sum_{i=1}^N X_j^{(i)} \frac{\partial \log \theta_j c^{(i)}}{\partial \theta} + \sum_{j=1}^{784} \sum_{i=1}^N (1 - X_j^{(i)}) \frac{\partial (1 - \theta_j c^{(i)})}{\partial \theta} \\ &= M \in \mathbb{R}^{784 \times 10} \quad \text{where } M_{jc} = \frac{\sum_{i=1}^N \mathbb{I}[X_j^{(i)}=1 \& C^{(i)}=C]}{\theta_{jc}} \\ &\quad - \frac{\sum_{i=1}^N \mathbb{I}[X_j^{(i)}=0 \& C^{(i)}=C]}{1-\theta_{jc}} \end{aligned}$$

$$\frac{\partial \ell(X, C | \theta, \pi)}{\partial \theta} = 0 \iff M = 0$$

$$\iff M_{jc} = 0 \quad \forall j, c$$

$$\iff \frac{\sum_{i=1}^N \mathbb{I}[X_j^{(i)}=1 \& C^{(i)}=C]}{\theta_{jc}} - \frac{\sum_{i=1}^N \mathbb{I}[X_j^{(i)}=0 \& C^{(i)}=C]}{1-\theta_{jc}} = 0$$

$$(1 - \theta_{jC}) \sum_{i=1}^N \mathbb{I}[x_j^{(i)} = 1 \& c^{(i)} = C] = \theta_{jC} \sum_{i=1}^N \mathbb{I}[x_j^{(i)} = 0 \& c^{(i)} = C]$$

$$\frac{1}{\theta_{jC}} - 1 = \frac{\sum_{i=1}^N \mathbb{I}[x_j^{(i)} = 0 \& c^{(i)} = C]}{\sum_{i=1}^N \mathbb{I}[x_j^{(i)} = 1 \& c^{(i)} = C]}$$

$$\frac{1}{\theta_{jC}} = 1 + \frac{\sum_{i=1}^N \mathbb{I}[x_j^{(i)} = 0 \& c^{(i)} = C]}{\sum_{i=1}^N \mathbb{I}[x_j^{(i)} = 1 \& c^{(i)} = C]}$$

$$\frac{1}{\theta_{jC}} = \frac{\sum_{i=1}^N \mathbb{I}[x_j^{(i)} = 1 \& c^{(i)} = C] + \sum_{i=1}^N \mathbb{I}[x_j^{(i)} = 0 \& c^{(i)} = C]}{\sum_{i=1}^N \mathbb{I}[x_j^{(i)} = 1 \& c^{(i)} = C]}$$

$$\theta_{jC} = \frac{\sum_{i=1}^N \mathbb{I}[x_j^{(i)} = 1 \& c^{(i)} = C]}{\sum_{i=1}^N \mathbb{I}[x_j^{(i)} = 1 \& c^{(i)} = C] + \sum_{i=1}^N \mathbb{I}[x_j^{(i)} = 0 \& c^{(i)} = C]}$$

$$\boxed{\theta_{jC} = \frac{\sum_{i=1}^N \mathbb{I}[x_j^{(i)} = 1 \& c^{(i)} = C]}{\sum_{i=1}^N \mathbb{I}[c^{(i)} = C]} = \frac{\# \text{ of sample where } j^{\text{th}} \text{ bit is 1 and class is } C}{\# \text{ of sample of class } C}}$$

$$\begin{aligned} \log P(t^{(i)} | \pi) &= \log \prod_{j=0}^q \pi_j^{t_j^{(i)}} \\ &= \sum_{j=0}^q \log \pi_j^{t_j^{(i)}} \\ &= \sum_{j=0}^8 \log \pi_j^{t_j^{(i)}} + \log \pi_9^{t_9^{(i)}} \\ &= \sum_{j=0}^8 t_j^{(i)} \log \pi_j + t_9^{(i)} \log (1 - \sum_{j=0}^8 \pi_j) \end{aligned}$$

$$\begin{aligned} \frac{\partial \log P(t^{(i)} | \pi)}{\partial \pi_k} &= t_k^{(i)} \frac{1}{\pi_k} + t_9^{(i)} \frac{1}{1 - \sum_{j=0}^8 \pi_j} \cdot (-1) \quad \forall k \in \{0, 1, \dots, 8\} \\ &= \frac{t_k^{(i)}}{\pi_k} - \frac{t_9^{(i)}}{\pi_9} \end{aligned}$$

$$\text{Therefore } \frac{\partial l(x, c | \theta, \pi)}{\partial \pi} = 0 \Leftrightarrow \sum_{i=1}^N \frac{\partial \log P(t^{(i)} | \pi)}{\partial \pi} = 0$$

$$\Leftrightarrow \sum_{i=1}^N \frac{\partial \log P(t^{(i)} | \pi)}{\partial \pi_k} = 0 \quad \forall k \in \{0, 1, \dots, 9\}$$

$$\Leftrightarrow \sum_{i=1}^N \left[ \frac{t_k^{(i)}}{\pi_k} - \frac{t_9^{(i)}}{\pi_9} \right] = 0$$

$$\frac{1}{\pi_k} \sum_{i=1}^N t_k^{(i)} = \frac{1}{\pi_9} \sum_{i=1}^N t_9^{(i)}$$

$$\pi_k = \pi_q \cdot \frac{\sum_{i=1}^N t_k^{(i)}}{\sum_{i=1}^N t_q^{(i)}} \quad \forall k \in \{0, \dots, 8\}$$

We know  $\sum_{k=0}^8 \pi_k + \pi_q = 1$

$$\sum_{k=0}^8 \left( \pi_q \cdot \frac{\sum_{i=1}^N t_k^{(i)}}{\sum_{i=1}^N t_q^{(i)}} \right) + \pi_q = 1$$

$$\left( 1 + \frac{\sum_{k=0}^8 \sum_{i=1}^N t_k^{(i)}}{\sum_{i=1}^N t_q^{(i)}} \right) \cdot \pi_q = 1$$

$$\pi_q = \frac{\sum_{i=1}^N t_q^{(i)}}{\sum_{k=0}^8 \sum_{i=1}^N t_k^{(i)} + \sum_{i=1}^N t_q^{(i)}}$$

$$\pi_q = \frac{\sum_{i=1}^N t_q^{(i)}}{\sum_{k=0}^8 \sum_{i=1}^N t_k^{(i)}} = \frac{\sum_{i=1}^N t_q^{(i)}}{N} = \frac{\sum_{i=1}^N \mathbb{I}[t_q^{(i)} = 1]}{N}$$

AND  $\pi_k = \pi_q \cdot \frac{\sum_{i=1}^N t_k^{(i)}}{\sum_{i=1}^N t_q^{(i)}}$

$$= \frac{\sum_{i=1}^N t_q^{(i)}}{N} \cdot \frac{\sum_{i=1}^N t_k^{(i)}}{\sum_{i=1}^N t_q^{(i)}}$$

$$= \frac{\sum_{i=1}^N t_k^{(i)}}{N} = \frac{\sum_{i=1}^N \mathbb{I}[t_k^{(i)} = 1]}{N}$$

Therefore  $\bar{\pi}_k = \frac{\sum_{i=1}^N \mathbb{I}[t_k^{(i)} = 1]}{N} = \frac{\# \text{ of sample of class } k}{\# \text{ of sample}} \quad \text{for } k \in \{0, \dots, 9\}$

$$(b) p(c | x^{(i)}, \theta, \pi) = \frac{p(x^{(i)} | c, \theta, \pi) \cdot p(c | \pi)}{p(x^{(i)})}$$

$$\begin{aligned} &= \frac{p(x^{(i)} | c, \theta, \pi) \cdot p(c | \pi)}{\sum_c (p(x^{(i)} | c) \cdot p(c | \pi))} \\ &= \frac{\prod_j [x_j^{(i)} (\theta_{jc})^{x_j^{(i)}} (1 - \theta_{jc})^{1 - x_j^{(i)}}] \cdot \pi_c}{\sum_c [\prod_j x_j^{(i)} (\theta_{jc})^{x_j^{(i)}} (1 - \theta_{jc})^{1 - x_j^{(i)}}] \cdot \pi_c} \end{aligned}$$

$$\begin{aligned} \log p(c | x^{(i)}, \theta, \pi) &= \sum_j [x_j^{(i)} \log \theta_{jc} + (1 - x_j^{(i)}) \log (1 - \theta_{jc})] + \log \pi_c \\ &\quad - \log \sum_c [\prod_j x_j^{(i)} (\theta_{jc})^{x_j^{(i)}} (1 - \theta_{jc})^{1 - x_j^{(i)}}] \cdot \pi_c \end{aligned}$$

(c) The average log-likelihood per data point cannot be computed, since the probability can be so small such that it is rounded to 0.  $\log 0$  is undefined.

(d) Graph in jupyter notebook

$$(e) P(\theta) = \text{Beta}(3, 3)$$

$$P(\theta | x) = \frac{P(x | \theta) \cdot P(\theta)}{P(x)} = \frac{P(x | c, \theta) \cdot P(\theta)}{P(x | c)}$$

$$\text{Beta}(\theta | 3, 3) = \left( \frac{\Gamma(3+3)}{\Gamma(3)\Gamma(3)} \right) \theta^2 (1-\theta)^2 = D$$

$$\frac{\partial \log P(\theta)}{\partial \theta_{ab}} = \frac{\partial \sum_c \log \theta_{ic}}{\partial \theta_{ab}} + \frac{\partial \sum_c \log (1-\theta_{ic})}{\partial \theta_{ab}} + \frac{\partial \sum_c D}{\partial \theta_{ab}} \geq 0$$

$$= \frac{2}{\theta_{ab}} + \frac{-2}{1-\theta_{ab}}$$

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

Therefore  $\frac{\partial \log P(\theta)}{\partial \theta} = M_2 \in \mathbb{R}^{784 \times 10}$  and  $M_{2,jc} = -1 \quad \forall j, c$

$$\begin{aligned} \frac{\partial l(\theta | x)}{\partial \theta} &= \frac{\partial \log P(\theta | x)}{\partial \theta} = \frac{\partial \log \prod_i [P(x^{(i)} | \theta) \cdot P(\theta) / P(x | c)]}{\partial \theta} \\ &= \frac{\partial \sum_i l(x^{(i)} | \theta)}{\partial \theta} + \frac{\partial \log P(\theta)}{\partial \theta} \\ &= M + \frac{\partial \log P(\theta)}{\partial \theta} = M + M_2 \end{aligned}$$

$$\text{According to (a)} \quad M \in \mathbb{R}^{784 \times 10} \quad \text{where } M_{jc} = \frac{\sum_{i=1}^N \mathbb{I}[x_j^{(i)} = 1 \& c^{(i)} = c]}{\theta_{jc}}$$

$$\frac{\sum_{i=1}^N \mathbb{I}[x_j^{(i)} = 0 \& c^{(i)} = c]}{1 - \theta_{jc}}$$

$$\text{Therefore } \frac{\partial l(\theta | x)}{\partial \theta} = M + M_2 = M_3 \in \mathbb{R}^{784 \times 10}$$

$$M_{3jc} = 0 \Leftrightarrow M_{jc} + M_{2jc} = 0$$

$$\frac{\sum_{i=1}^N \mathbb{I}[x_j^{(i)} = 1 \& C^{(i)} = c]}{\theta_{jc}} - \frac{\sum_{i=1}^N \mathbb{I}[x_j^{(i)} = 0 \& C^{(i)} = c]}{1 - \theta_{jc}} + \frac{2}{\theta_{jc}} - \frac{2}{1 - \theta_{jc}} = 0$$

$$\frac{2 + \sum_{i=1}^N \mathbb{I}[x_j^{(i)} = 1 \& C^{(i)} = c]}{\theta_{jc}} = \frac{2 + \sum_{i=1}^N \mathbb{I}[x_j^{(i)} = 0 \& C^{(i)} = c]}{1 - \theta_{jc}}$$

$$\frac{1}{\theta_{jc}} - 1 = \frac{2 + \sum_{i=1}^N \mathbb{I}[x_j^{(i)} = 0 \& C^{(i)} = c]}{2 + \sum_{i=1}^N \mathbb{I}[x_j^{(i)} = 1 \& C^{(i)} = c]}$$

$$\frac{1}{\theta_{jc}} = \frac{2 + \sum_{i=1}^N \mathbb{I}[x_j^{(i)} = 0 \& C^{(i)} = c]}{2 + \sum_{i=1}^N \mathbb{I}[x_j^{(i)} = 1 \& C^{(i)} = c]} + 1$$

$$\theta_{jc} = \frac{2 + \sum_{i=1}^N \mathbb{I}[x_j^{(i)} = 0 \& C^{(i)} = c]}{4 + \sum_{i=1}^N \mathbb{I}[x_j^{(i)} = 0 \& C^{(i)} = c] + \sum_{i=1}^N \mathbb{I}[x_j^{(i)} = 1 \& C^{(i)} = c]}$$

$$\theta_{jc} = \frac{2 + \sum_{i=1}^N \mathbb{I}[x_j^{(i)} = 0 \& C^{(i)} = c]}{4 + \sum_{i=1}^N \mathbb{I}[C^{(i)} = c]}$$

$$\theta_{jc} = \frac{2 + \# \text{ of sample where } j\text{-th bit is 1 and class is } c}{4 + \# \text{ of sample where class is } c}$$

## Question 2

- (a) True      (b) False

$$P(x_1, x_2) = P(x_1) \cdot P(x_2)$$

$$P(x_1, x_2 | c) = P(x_1 | c) \cdot P(x_2 | c)$$

$$P(x_1) = \sum_c P(x_1 | c), P(x_2) = \sum_c P(x_2 | c)$$

$$P(x_1, x_2) = \sum_c P(x_1, x_2 | c)$$

$$P(x_1) \cdot P(x_2) = \sum_c P(x_1 | c) \cdot \sum_c P(x_2 | c) \geq \sum_c P(x_1, x_2 | c) \text{ Thus, (b) is false.}$$