

2.3

By Bayes' rule,

$$p(c|x, \theta, \pi) = \frac{p(c, x | \theta, \pi)}{p(x | \theta, \pi)}$$

Now take log to  $p(c|x, \theta, \pi)$ , we have

$$\begin{aligned} \log p(c|x, \theta, \pi) &= \log \frac{p(c, x | \theta, \pi)}{p(x | \theta, \pi)} \\ &= \log p(c, x | \theta, \pi) - \log p(x | \theta, \pi) \\ &= \log p(c | \pi) p(x | c, \theta) - \log \left[ \sum_{c=0}^9 p(c | \pi) p(x | c, \theta) \right] \\ &= \log p(c | \pi) + \log p(x | c, \theta) - \log \left[ \sum_{c=0}^9 p(c | \pi) p(x | c, \theta) \right] \\ &= \log \left[ p(c | \pi) \prod_{j=1}^{784} p(x_j | c, \theta) \right] - \log \left[ \sum_{c=0}^9 p(c | \pi) p(x | c, \theta) \right] \\ &= \log p(c | \pi) + \sum_{j=1}^{784} \log p(x_j | c, \theta) - \log \left[ \sum_{c=0}^9 p(c | \pi) \prod_{j=1}^{784} p(x_j | c, \theta, \pi) \right] \\ &= \log \pi_c + \sum_{j=1}^{784} [x_j \log \theta_{jc} + (1-x_j) \log (1-\theta_{jc})] - \log \left[ \sum_{c=0}^9 p(c | \pi) \prod_{j=1}^{784} p(x_j | c, \theta, \pi) \right] \end{aligned}$$

For the following function **log-likelihood** :

$$\text{log-likelihood} = \log \pi_c + \sum_{j=1}^{784} [x_j \log \theta_{jc} + (1-x_j) \log (1-\theta_{jc})] - \log \left[ \sum_{c=0}^9 p(c | \pi) \prod_{j=1}^{784} p(x_j | c, \theta, \pi) \right]$$

$\pi_c$  is given by MLE  $\hat{\pi}_c$ ,

$\theta_{jc}$  is given by MAP estimator  $\hat{\theta}_{\text{MAP}_{jc}}$

$x_j$  is given by images