

2.2

Since the prior given is $\text{Beta}(2, 2)$, so the PDF of prior will be

$$f(\theta) = \frac{\theta(1-\theta)}{6}$$

In MAP, we're supposed to optimize $P(x, c | \theta, \pi) f(\theta)$, and we have the likelihood function from previous question.

$$\begin{aligned} \log(P(x | \theta) f(\theta)) &= \left(\sum_{i=1}^N \sum_{c=0}^9 \mathbb{1}\{C^{(i)} = c\} \log \pi_c + \sum_{i=1}^N \sum_{j=1}^{784} [x_j^{(i)} \log \theta_{jc} + (1 - x_j^{(i)}) \log (1 - \theta_{jc})] \right) + \log \left(\frac{\theta(1-\theta)}{6} \right) \\ &= \left(\sum_{i=1}^N \sum_{c=0}^9 \mathbb{1}\{C^{(i)} = c\} \log \pi_c + \sum_{i=1}^N \sum_{j=1}^{784} [x_j^{(i)} \log \theta_{jc} + (1 - x_j^{(i)}) \log (1 - \theta_{jc})] \right) + \log(\theta) + \log(1-\theta) - \log(6) \end{aligned}$$

Then take the derivative respect to θ , and use part of result from previous question.

$$\sum_{i=1}^N \left(\frac{x_j^{(i)}}{\theta_{jc}} - \frac{1 - x_j^{(i)}}{1 - \theta_{jc}} \right) \mathbb{1}\{C^{(i)} = c\} + \frac{1}{\theta} - \frac{1}{1 - \theta} = 0$$

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

Solving for θ , we have
$$\hat{\theta}_{\text{map}} = \frac{\sum_{i=1}^N [x_j^{(i)} \mathbb{1}\{C^{(i)} = c\}] + 1}{\sum_{i=1}^N [\mathbb{1}\{C^{(i)} = c\}] + 2}$$