

Q2:

E-step:

$$\begin{aligned} Q_t^{(i)}(c_t) &= P(c_t = k | x_t, \theta^{(i-1)}) \\ &\propto P(x_t | c_t = k, \theta^{(i-1)}) \cdot P(c_t = k | \theta^{(i-1)}) \\ &\propto \Phi(x_t, \lambda) \cdot \pi_k^{(i-1)} \\ Q_t^{(i)}(c_t) &= \frac{\lambda_k^{x_t} e^{-\lambda_k}}{\sum_{k=1}^K e^{-\lambda_k}} \end{aligned}$$

M-step for  $\pi_k$ :

$$\begin{aligned} \pi_k &= \underset{\pi_k}{\operatorname{argmax}} \left( \sum_{t=1}^n \sum_{k=1}^K Q_t^{(i)}(k) \log(\pi_k) \right) \\ \pi_k &= \frac{\sum_{t=1}^n Q_t^{(i)}(k)}{n} \end{aligned}$$

M-step for  $\lambda_k$ :

$$\begin{aligned} \lambda_k &= \underset{\lambda_k}{\operatorname{argmax}} \left( \sum_{t=1}^n \sum_{k=1}^K Q_t^{(i)}(k) \log(P(x_t | c_t = k, \theta)) \right) \\ &= \underset{\lambda_k}{\operatorname{argmax}} \left( \sum_{t=1}^n \sum_{k=1}^K Q_t^{(i)}(k) \log \left( \frac{\lambda_k^{x_t} e^{-\lambda_k}}{x_t!} \right) \right) \\ &= \underset{\lambda_k}{\operatorname{argmax}} \left( \sum_{t=1}^n \sum_{k=1}^K Q_t^{(i)}(k) (x_t \log(\lambda_k) - \lambda_k - \log(x_t!)) \right) \end{aligned}$$

Take the derivative of the RHS and equate it to 0

Then we get:

$$\sum_{t=1}^n Q_t^{(i)}(k) \left( x_t \frac{1}{\lambda_k} - 1 \right) = 0$$

So:

$$\lambda_k = \frac{\sum_{t=1}^n Q_t^{(i)}(k) x_t}{\sum_{t=1}^n Q_t^{(i)}(k)}$$