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Problem 1

1.1 Lagrangian is

$$L(\{a_k\}, \{q_k\}, \{\lambda_k\}, \eta) = \sum_{k=1}^K a_k \ln q_k - \lambda_k q_k + \eta \left(1 - \sum_{k=1}^K q_k\right)$$

$$\frac{\partial L}{\partial q_k} = \frac{a_k}{q_k} - \lambda_k - \eta = 0$$

$$a_k = (\lambda_k + \eta) q_k$$

$$\sum_{k=1}^K a_k = (\lambda_k + \eta) \sum_{k=1}^K q_k$$

$$\text{so, } q_k^* = \frac{a_k}{\lambda_k + \eta} = \frac{a_k}{\sum_{k=1}^K a_k} \quad \forall k$$

1.2 $L(\{q_k\}, \{b_k\}, \{\lambda_k\}, \eta) =$

$$\sum_{k=1}^K (q_k b_k - q_k \ln q_k) - \lambda_k q_k + \eta \left(1 - \sum_{k=1}^K q_k\right)$$

$$\frac{\partial L}{\partial q_k} = b_k - (\ln q_k + 1) - \lambda_k - \eta = 0$$

$$\sum_{k=1}^K e^{b_k - \lambda_k - \eta - 1} = \sum_{k=1}^K q_k = 1$$

$$b_k - \lambda_k - \eta - 1 = 0$$

$$\lambda_k q_k = 0 \Rightarrow \lambda_k = 0$$

$$\sum_{k=1}^K e^{b_k} = e^{\eta + 1}$$

$$\eta = \ln \sum_{k=1}^K e^{b_k} - 1$$

$$\begin{aligned} q_k^* &= \frac{e^{b_k} - \ln \sum_{k=1}^K e^{b_k}}{e^{b_k}} \\ &= \frac{\sum_{k=1}^K e^{b_k}}{\sum_{k=1}^K e^{b_k}} \end{aligned}$$

Problem 2 2.1 from 1.1 let $\alpha_k = \sum_n r_{nk}$, $q_k = w_k$

$$w_k^* = \frac{\sum_n r_{nk}}{\sum_{k=1}^K \sum_n r_{nk}} = \frac{\sum_n r_{nk}}{N}$$

$$\text{For } \mu_k \in \Sigma_K, L = \arg \max \sum_n \sum_k r_{nk} \left(-\frac{D}{2} \ln 2\pi - D \ln \sigma_k^2 - \frac{1}{2\sigma_k^2} (x_n - \mu_k)^2 \right)$$

$$\frac{\partial L}{\partial \mu_k} = \sum_n r_{nk} \frac{1}{\sigma_k^2} (x_n - \mu_k) = 0$$

$$\mu_k = \frac{\sum_n r_{nk} x_n}{\sum_n r_{nk}}$$

$$\frac{\partial L}{\partial \sigma_k^2} = - \sum_n \frac{D \cdot r_{nk}}{\sigma_k^2} + \sum_n \frac{r_{nk}}{\sigma_k^3} (x_n - \mu_k)^2 = 0$$

$$\sigma_k^2 = \frac{\sum_n r_{nk} \|x_n - \mu_k\|_2^2}{D \sum_n r_{nk}}$$

$$2.2 \quad \arg \max_{q_n \in \Delta} E_{z_n \sim q_n} [\ln p(x_n, z_n; \theta^t)] - E_{z_n \sim q_n} [\ln q_n]$$

$$= \arg \max_{q_n \in \Delta} q_n \ln p(x_n, z_n; \theta^t) - q_n \ln q_n$$

from 1.2

$$q_n^{(t)}(k) = \frac{p(x_n, z_n; \theta^t)}{\sum_{k=1}^K p(x_n, z_n; \theta^t)} = \frac{p(x_n, z_n=k; \theta^t)}{p(x_n; \theta^t)}$$

$$= p(z_n=k | x_n; \theta^t)$$

$$2.3 \quad p(z_n=k|x_n) = \frac{w_k N(x_n|\mu_k, \Sigma_k)}{\sum w_k N(x_n|\mu_k, \Sigma_k)}$$

$$\text{let } w = \frac{1}{K}, \quad e^{-\frac{1}{2\sigma^2} \|x_n - \mu_k\|_2^2}$$

$$p(z_n=k|x_n) = \frac{e^{-\frac{1}{2\sigma^2} \|x_n - \mu_k\|_2^2}}{\sum_k e^{-\frac{1}{2\sigma^2} \|x_n - \mu_k\|_2^2}}$$

let $\sigma \rightarrow 0$

$$\lim_{\sigma \rightarrow 0} p(z_n=k|x_n) = 1 \text{ if } \|x_n - \mu_k\|_2^2 = \min_j \|x_n - \mu_j\|_2^2 \quad j \in [1, K]$$

$$\lim_{\sigma \rightarrow 0} p(z_n=k|x_n) = 0 \text{ if } \|x_n - \mu_k\|_2^2 \neq \min_j \|x_n - \mu_j\|_2^2 \quad j \in [1, K]$$

$$\text{so, } p(z_n=k|x_n) = I[k = \arg \min_j \|x_n - \mu_j\|_2^2] \quad j \in [1, K]$$

Problem 3 3.1 $P(X_{T+1}=S | O_{1:T}=O_{1:T})$

$$= P(X_{T+1}=S, X_T=S' | O_{1:T}=O_{1:T})$$

$$= P(X_{T+1}=S | X_T=S', O_{1:T}=O_{1:T}) P(X_T=S' | O_{1:T}=O_{1:T})$$

$$= P(X_{T+1}=S | X_T=S') \alpha_{S'}(T)$$

$$= \alpha_{S,S'} \alpha_{S'}(T)$$

3.2

(A) ~~$\alpha_{S}(2) = \sum_{S'} \alpha_{S'} \alpha_{S'}(t-1)$~~

$$= P(X_2=B | X_1=A)$$

$$\alpha_A(1) = \pi_A b_{A,0} = 0.7 \times 0.4 = 0.28$$

$$\alpha_B(1) = \pi_B b_{B,0} = 0.3 \times 0.7 = 0.21$$

$$\begin{aligned}\alpha_A(2) &= P(X_3=A | X_2=A) \alpha_A(1) P(O_t=0n | X_2=A) \\ &\quad + P(O_t=0n | X_2=B) P(X_{T+1}=A | X_2=B) \alpha_B(1) \\ &= 0.2 \times 0.28 \times 0.6 + 0.6 \times 0.21 \times 0.7 \\ &= 0.1218\end{aligned}$$

$$\begin{aligned}\alpha_B(2) &= P(O_t=0n | X_2=B) P(X_{T+1}=B | X_2=B) \alpha_A(1) \\ &\quad + P(O_t=0n | X_2=B) P(X_{T+1}=B | X_2=B) \alpha_B(1) \\ &= 0.3 \times (0.7 \times 0.28 + 0.2 \times 0.21) \\ &= 0.0714\end{aligned}$$

~~$\alpha_A(3) = P(X_3=A) = P(X_1=A) P(X_2=A | X_1=A) P(X_3=A | X_2=A)$~~

$$= + P$$

$$\begin{aligned}P(X_3=A) &= \sum_{S_1, S_2} P(X_1=S_1) P(X_2=S_2 | X_1=S_1) P(X_3=A | X_2=S_2) \\ &= 0.7 \times 0.2 \times 0.2 + 0.7 \times 0.7 \times 0.7 + 0.3 \times 0.2 \times 0.7 + 0.3 \times 0.7 \times 0.2 \\ &= 0.455\end{aligned}$$

$$\begin{aligned}P(X_3=B) &= 0.7 \times 0.2 \times 0.7 + 0.7 \times 0.7 \times 0.2 + 0.3 \times 0.7 \times 0.7 + 0.3 \times 0.2 \times 0.2 \\ &= 0.355\end{aligned}$$

$$P(X_3=A|D_{1:2}) = \frac{\alpha_A(2) P(X_3=B|x_2=A) + \alpha_B(2) P(X_3=A|x_2=B)}{\alpha_A(2) + \alpha_B(2)}$$

$$= 0.3848$$

$$P(X_3=B|D_{1:2}) = \frac{\alpha_A(2) P(X_3=B|x_2=A) + \alpha_B(2) P(X_3=B|x_2=B)}{\alpha_A(2) + \alpha_B(2)}$$

$$= 0.5152$$

$$P(X_3=\text{End}|D_{1:2}) = \cancel{0.5}, -0.3848 - 0.5152 = 0.1$$

most likely state is B

most likely state at $t=3$ is A

$$(b) \delta_{A(1)} = 0.28$$

$$\delta_{B(1)} = 0.21$$

$$\begin{aligned}\delta_{A(2)} &= P(D_2 = \text{on} | X_2 = A) \max\{P(X_2 = A | X_1 = A) \delta_{A(1)}, P(X_2 = B | X_1 = A) \delta_{B(1)}\} \\ &= 0.6 \max\{0.2 \times 0.28, 0.7 \times 0.21\} \\ &= 0.0882\end{aligned}$$

$$\Delta_A(2) = B$$

$$\begin{aligned}\delta_{B(2)} &= P(D_2 = \text{on} | X_2 = B) \max\{P(X_2 = B | X_1 = A) \delta_{A(1)}, P(X_2 = B | X_1 = B) \delta_{B(1)}\} \\ &= 0.3 \max\{0.7 \times 0.28, 0.2 \times 0.21\} \\ &= 0.0588\end{aligned}$$

$$\Delta_B(2) = A$$

$$\begin{aligned}\delta_{A(3)} &= P(D_3 = \text{on} | X_3 = A) \max\{P(X_3 = A | X_2 = A) \delta_{A(2)}, P(X_3 = B | X_2 = A) \delta_{B(2)}\} \\ &= 0.6 \max\{0.2 \times 0.0882, 0.7 \times 0.0588\} \\ &= 0.0247\end{aligned}$$

$$\Delta_A(3) = B$$

$$\begin{aligned}\delta_{B(3)} &= P(D_3 = \text{on} | X_3 = B) \max\{P(X_3 = B | X_2 = A) \delta_{A(2)}, P(X_3 = B | X_2 = B) \delta_{B(2)}\} \\ &= 0.3 \max\{0.7 \times 0.0882, 0.2 \times 0.0588\} \\ &= 0.0185 < 0.0247\end{aligned}$$

$$\Delta_B(3) = A$$

so, most likely sequence is ABA

$$3.3 \quad \text{let } P(X_t = s) = \underbrace{w_s}_{t \in \mathbb{N}}$$

$$P(X_{t+1} = s' | X_t = s) = w_{s'}$$

$$P(x_t | X_t = s) = \mathcal{N}(\mu_t, \Sigma_t) \quad N(M_t, \Sigma_t)$$

$$\begin{aligned}P(X_2 = s' | O_1, O_2) &= P(X_2 = s' | O_2) = \frac{P(X_2 = s, O_2)}{P(O = O_2)} = \frac{P(X_2 = s) P(O_2 | X_2 = s)}{P(O = O_2)} \\ &= \frac{w_s \times N(M_t, \Sigma_t)}{\sum_k w_k \times N(M_t, \Sigma_t)} = P(z = s | O = O_2)\end{aligned}$$