Homework #4

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Problem 1

We need to show that the two training objectives are equivalent. Let's call,

$$A = \sum_{i=1}^{N} (y_i - (w \cdot \phi(x_i) + w_0))^2 + \lambda ||w||^2$$

$$B = \sum_{i=1}^{N} (y_i - (w \cdot \phi(x_i) + w_0))^2 + \alpha(||w||^2 - \tau)$$

Where we have used a Lagrange multiplier, α , for the constraint.

For the optimal solutions, we know:

$$\nabla_w A(w_\lambda^*) = 0$$

and using the KKT conditions,

$$\nabla_w B(w_{\alpha}^*) = \nabla_w A(w_{\alpha}^*) = 0 \text{ and } \alpha(||w_{\alpha}^*||^2 - \tau) = 0$$

where w_{λ}^* are the optimal weights for a given λ and w_{α}^* are the optimal weights for a given α .

Say we set $\tau=||w_{\lambda}^*||^2$, then we have $\lambda=\alpha$ and $w_{\lambda}^*=w_{\alpha}^*$, satisfying the KKT condition.

Therefore, the optimal weights will be the same for a $\tau = ||w_{\lambda}^*||^2$.

Problem 2

See attached notebook.

Problem 3

The posterior probability $\gamma_{i,c} = p(z_i = c|x_i; \theta, \pi)$ may be written as,

$$\gamma_{i,c} = \frac{\pi_c p(x_i | \theta_c)}{\sum_{l=1}^k \pi_l p(x_i | \theta_l)}$$

For a Bernoulli mixture,

$$p(x_i|\theta_c) = \prod_{j=1}^d \theta_{c,j}^{x_{i,j}} (1 - \theta_{c,j})^{(1-x_{i,j})}$$

Therefore,

$$\gamma_{i,c} = \frac{\pi_c \prod_{j=1}^d \theta_{c,j}^{x_{i,j}} (1 - \theta_{c,j})^{(1 - x_{i,j})}}{\sum_{l=1}^k \pi_l \prod_{j=1}^d \theta_{l,j}^{x_{i,j}} (1 - \theta_{l,j})^{(1 - x_{i,j})}}$$

Problem 4

The M-step updates are,

$$\pi^{new}, \theta^{new} = argmax_{\pi,\theta} \left[\sum_{i=1}^{N} \sum_{c=1}^{k} \gamma_{i,c} (\log \pi_c + \log p(x_i; \theta_c)) \right]$$

Let us set,

$$A = \sum_{i=1}^{N} \sum_{c=1}^{k} \gamma_{i,c} (\log \pi_c + \log p(x_i; \theta_c)) = \sum_{i=1}^{N} \sum_{c=1}^{k} \gamma_{i,c} (\log \pi_c + \sum_{i=1}^{d} [x_{i,j} \log \theta_{c,j} + (1 - x_{i,j}) \log (1 - \theta_{c,j})])$$

Solving for π^{new} using Lagrange multipliers and differentiating,

$$A = \sum_{i=1}^{N} \sum_{c=1}^{k} \gamma_{i,c} (\log \pi_c) + \alpha (\sum_{c=1}^{k} \pi_c - 1) + \dots$$

$$\frac{\partial A}{\partial \pi_c} = 0 \to \sum_{i=1}^{N} \gamma_{i,c} = -\alpha \pi_c$$

$$\sum_{c=1}^{k} \sum_{i=1}^{N} \gamma_{i,c} = -\sum_{c=1}^{k} \alpha \pi_c \to \sum_{i=1}^{N} 1 = -\alpha \to \alpha = -N$$

$$\pi_k^{new} = \frac{1}{N} \sum_{i=1}^{N} \gamma_{i,k}$$

for a given subclass k. Solving for θ^{new} ,

$$\frac{\partial A}{\partial \theta_{k,r}} = 0 = \sum_{i=1}^{N} \gamma_{i,k} \left(\frac{x_{i,r}}{\theta_{k,r}} - \frac{1 - x_{i,r}}{1 - \theta_{k,r}} \right) = \sum_{i=1}^{N} \gamma_{i,k} (x_{i,r} - \theta_{k,r}) = 0$$

$$\theta_{k,r}^{new} = \frac{\sum_{i=1}^{N} \gamma_{i,k} x_{i,r}}{\sum_{i=1}^{N} \gamma_{i,k}}$$

for a given subclass k and parameter r.

Problem 5

See attached notebook.