

# Homework 3

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## 1 Some Lemmas in Linear Algebra

**Lemma 1.**

$$x^T A x = \text{tr}(A x x^T).$$

**Lemma 2.**

$$\frac{\partial \log |A|}{\partial A} = 2A^{-1} - \text{diag}(A^{-1}) \quad \text{and} \quad \frac{\partial \text{tr}(AB)}{\partial A} = (B + B^T) - \text{diag}(B).$$

## 2 Update $\Sigma_k$ in EM

Now we want to find a good enough  $\Sigma_k$ . Our goal is to maximize the likelihood

$$\mathcal{L} = \sum_{n=1}^N \ln \left( \sum_{k=1}^K \pi_k \mathcal{N}(x_n | \mu_k, \Sigma_k) \right)$$

where

$$\mathcal{N}(x_n | \mu_k, \Sigma_k) = \frac{1}{(2\pi)^{D/2} |\Sigma_k|^{1/2}} \exp \left( -\frac{1}{2} (x_n - \mu_k)^T \Sigma_k^{-1} (x_n - \mu_k) \right).$$

Thus

$$\begin{aligned} \frac{\partial L}{\partial \Sigma_k} &= \sum_{n=1}^N \left\{ \frac{1}{\sum_{j=1}^K \pi_j \mathcal{N}(x_n | \mu_j, \Sigma_j)} \times \frac{\partial}{\partial A} \pi_k \mathcal{N}(x_n | \mu_k, \Sigma_k) \right\} \\ &= \sum_{n=1}^N \left\{ \frac{\pi_k \mathcal{N}(x_n | \mu_k, \Sigma_k)}{\sum_{j=1}^K \pi_j \mathcal{N}(x_n | \mu_j, \Sigma_j)} \times \frac{\partial}{\partial A} \left[ -\frac{1}{2} \log |\Sigma_k| - \frac{1}{2} (x_n - \mu_k)^T \Sigma_k^{-1} (x_n - \mu_k) \right] \right\} \\ &= \sum_{n=1}^N \left\{ \frac{\pi_k \mathcal{N}(x_n | \mu_k, \Sigma_k)}{\sum_{j=1}^K \pi_j \mathcal{N}(x_n | \mu_j, \Sigma_j)} \times \frac{\partial}{\partial A} \left[ \frac{1}{2} \log |\Sigma_k^{-1}| - \frac{1}{2} \text{tr}(\Sigma_k^{-1} N_{n,k}) \right] \right\} \\ &= \sum_{n=1}^N \left\{ \frac{\pi_k \mathcal{N}(x_n | \mu_k, \Sigma_k)}{\sum_{j=1}^K \pi_j \mathcal{N}(x_n | \mu_j, \Sigma_j)} \times \frac{1}{2} \left[ 2(\Sigma_k - N_{n,k}) - \text{diag}(\Sigma_k - N_{n,k}) \right] \right\} \end{aligned}$$

where  $N_{n,k} = (x_n - \mu_k)(x_n - \mu_k)^T$ .  $\partial L / \partial \Sigma_k = 0$  gives

$$\sum_{n=1}^N \left[ \frac{\pi_k \mathcal{N}(x_n | \mu_k, \Sigma_k)}{\sum_{j=1}^K \pi_j \mathcal{N}(x_n | \mu_j, \Sigma_j)} \times (\Sigma_k - N_{n,k}) \right] = 0$$

which entails

$$\Sigma_k = \left( \sum_{n=1}^N \frac{\pi_k \mathcal{N}(x_n | \mu_k, \Sigma_k)}{\sum_{j=1}^K \pi_j \mathcal{N}(x_n | \mu_j, \Sigma_j)} \right)^{-1} \sum_{n=1}^N \left[ \frac{\pi_k \mathcal{N}(x_n | \mu_k, \Sigma_k)}{\sum_{j=1}^K \pi_j \mathcal{N}(x_n | \mu_j, \Sigma_j)} \times (x_n - \mu_k)(x_n - \mu_k)^T \right].$$

For simplicity,

$$\gamma(z_n k) \triangleq \frac{\sum_{n=1}^N \pi_k \mathcal{N}(x_n | \mu_k, \Sigma_k)}{\sum_{j=1}^K \pi_j \mathcal{N}(x_n | \mu_j, \Sigma_j)}$$

Finally

$$\Sigma_k = \frac{1}{\sum_{n=1}^N \gamma(z_n k)} \sum_{n=1}^N \left[ \gamma(z_n k) (x_n - \mu_k)(x_n - \mu_k)^T \right].$$

Nota Bena: if we update  $\mu$  first, then we could let  $\mu_k$  above be  $\mu_k^{\text{new}}$ .