

25/9/19

EE5601: Representation Learning

- Review
- Examples of GMM - multiple populations
- EM for GMM

Recall: $p(\underline{x}; \theta) = \sum_{k=1}^K \pi_k \mathcal{N}(\underline{x}; \underline{\mu}_k, \Sigma_k) ; \quad \sum_{k=1}^K \pi_k = 1 ;$
 $0 \leq \pi_k \leq 1.$

Prrior: $p(z_k = 1) \equiv \pi_k ; \quad p(\underline{z}) = \prod_{k=1}^K \pi_k^{z_k}$

Posterior: $p(z_k = 1 | \underline{x}) = \frac{p(\underline{x} | z_k = 1) \cdot p(z_k = 1)}{p(\underline{x})}$

$$= \frac{\pi_k \cdot \mathcal{N}(\underline{x}; \underline{\mu}_k, \Sigma_k)}{\sum_{j=1}^K \pi_j \mathcal{N}(\underline{x}; \underline{\mu}_j, \Sigma_j)}$$

$= \gamma(z_k) \quad \left| \begin{array}{l} \gamma(z_k) \text{ is the responsibility} \\ \text{of the } k^{\text{th}} \text{ mixture comp. in} \\ \text{explaining } \underline{x} \end{array} \right.$

Why $\gamma(z_{nk}) \equiv p(z_k = 1 | \underline{x}_n)$.

- Let us find the max likelihood estimates for $\{\pi_k, \underline{\mu}_k, \Sigma_k\}_{k=1}^K$
- Assume Σ_k is a diagonal matrix; (for simplified analysis)
- Let us find $\frac{\partial L(\underline{X}; \theta)}{\partial \mu_{kd}}$ and set it to 0.

$$\text{Recall } L(X; \theta) = \sum_{n=1}^N \log \left[\sum_{j=1}^K \pi_j \mathcal{N}(\underline{x}_n; \underline{\mu}_j, \Sigma_j) \right].$$

$$\text{Also, } \mathcal{N}(\underline{x}; \underline{\mu}_k, \Sigma_k) = \frac{1}{\sqrt{(2\pi)^d |\Sigma_k|}} \exp \left[-\frac{1}{2} (\underline{x} - \underline{\mu}_k)^T \Sigma_k^{-1} (\underline{x} - \underline{\mu}_k) \right]$$

Note that if Σ_k is a diagonal matrix,

$$-\frac{1}{2} (\underline{x} - \underline{\mu}_k)^T \Sigma_k^{-1} (\underline{x} - \underline{\mu}_k) = -\frac{1}{2} \sum_{i=1}^d \frac{(x_i - \mu_{ki})^2}{\sigma_{ki}^2}$$

Simplifying $\frac{\partial L(X; \theta)}{\partial \mu_{kd}}$:

$$\frac{\partial}{\partial \mu_{kd}} \left[\sum_{n=1}^N \log \left[\sum_{j=1}^K \pi_j \mathcal{N}(\underline{x}_n; \underline{\mu}_j, \Sigma_j) \right] \right]$$

$$= \sum_{n=1}^N \left[\frac{\pi_k \mathcal{N}(\underline{x}_n; \underline{\mu}_k, \Sigma_k) \cdot \left[2 \cdot \Sigma_k^{-1} \cdot (\underline{x}_n - \underline{\mu}_k) \right] z_k}{\underbrace{\sum_{j=1}^K \pi_j \mathcal{N}(\underline{x}_n; \underline{\mu}_j, \Sigma_j)}_{\gamma(z_{nk})}} \right] = 0$$