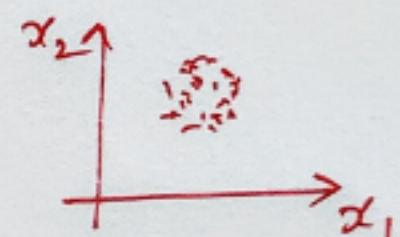


$$P(x | \mu, \sigma^2) = \frac{1}{\sqrt{2\pi} \sigma} e^{-\frac{(x-\mu)^2}{2\sigma^2}} \quad x \in \mathbb{R}$$

$x \in \mathbb{R}^d$

$$P(x | \mu_d, \Sigma_{d \times d}) = \frac{1}{(2\pi)^{d/2} \sqrt{|\Sigma|}} e^{-\frac{1}{2} (x-\mu)^T \Sigma^{-1} (x-\mu)} = \mathcal{N}(x | \mu, \Sigma)$$



Mixture Gaussian Distributions

$$P(x | \mu, \Sigma) = \sum_{k=1}^K \pi_k \mathcal{N}(x | \mu_k, \Sigma_k)$$

$\begin{matrix} z_1, z_2, z_3, z_4 \\ z_1=1, z_2=1 \end{matrix}$

Let us introduce a  $K$ -dimensional binary random vector

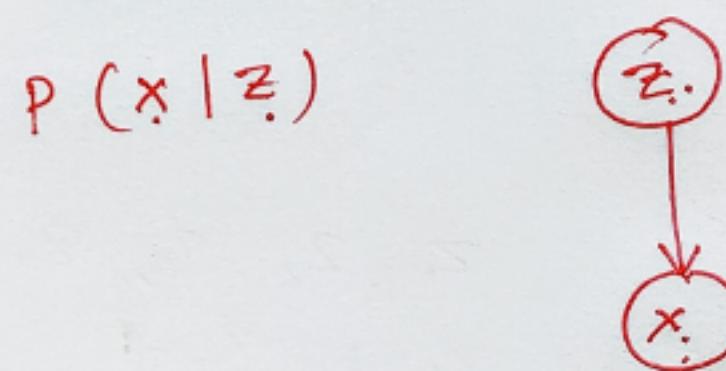
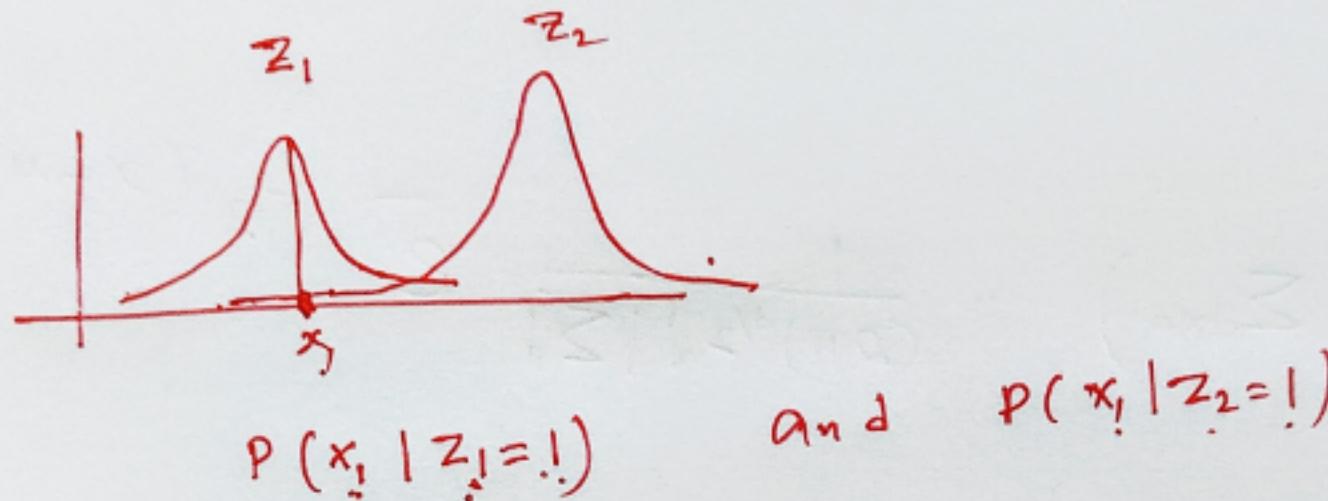
$$\underline{z} \in \{0, 1\}^K$$

$$\underline{z} = [z_1 \ z_2 \ \dots \ z_K]^T \quad z_i \in \{0, 1\} \quad \sum_{i=1}^K z_i = 1$$

$$\begin{aligned} \underline{z} &= [1, 0, 0, 0] \\ \underline{z} &= [0, 0, 1, 0] \end{aligned}$$

BE

$$S = \{x_1, x_2, \dots, x_N\} \quad x_i \in \mathbb{R}^d, \quad \text{P}$$



$$\underline{P(x, z)} = \underline{P(z)} \cdot \underline{P(x|z)}$$

Bayes Theorem.

$$\begin{aligned} P(z_k=1) &= \pi_k \\ P(z_k=0) &= 1 - \pi_k \end{aligned}$$

P

$$\begin{aligned} \underline{P(z)} &= P(z_1, z_2, \dots, z_n) \\ &= P(z_1) P(z_2) \cdots P(z_n) \\ &= \prod_{k=1}^K (\pi_k)^{\alpha_k} \\ \downarrow \quad \sum_{k=1}^K \pi_k &= 1. \end{aligned}$$

$$0 \leq \pi_k \leq 1.$$

$$p(x | z_k=1) = \mathcal{N}(x | \mu_k, \Sigma_k)$$

$$p(x | z_1=1) = \mathcal{N}(x | \mu_1, \Sigma_1)$$

$$p(x | z_2=1) = \mathcal{N}(x | \mu_2, \Sigma_2)$$

$$p(x | z) = \prod_{k=1}^K \mathcal{N}(x | \mu_k, \Sigma_k)$$

$z = [z_1 \quad \dots \quad z_K]$

$$p(x, z) = \sum p(z) \cdot p(x | z)$$

$$p(x) = \sum_z p(z) p(x | z)$$

$$= \sum_{k=1}^K p(z_k=1) \cdot p(x | z_k=1)$$

$$p(x) = \sum_{k=1}^K \pi_k \mathcal{N}(x | \mu_k, \Sigma_k)$$

$$\gamma(z_{n_k}) = p(z_k=1 | x=x_n) = \frac{p(z_k=1) p(x_n | z_k=1)}{p(x_n)}$$

$$= \frac{p(z_k=1) p(x_n | z_k=1)}{\sum_{j=1}^K p(z_j=1) \cdot p(x_n | z_j=1)}$$

$$\gamma(z_{nk}) = \frac{\pi_k N(x_n | \mu_k, \Sigma_k)}{\sum_{j=1}^K \pi_j N(x_n | \mu_j, \Sigma_j)}.$$

$$X = \{x_1, x_2, \dots, x_N\}$$

$$P(X) = P(x_1) \cdot P(x_2) \cdots P(x_N)$$

$$\log(P(X)) = \sum_{n=1}^N \log(P(x_i))$$

$$\Leftrightarrow L(X | \mu, \Sigma, \pi) = \sum_{n=1}^N \log \left( \sum_{j=1}^K \pi_j \cdot N(x_n | \mu_j, \Sigma_j) \right)$$

$$\max_{\mu, \Sigma, \pi} L(X | \mu, \Sigma, \pi).$$

$$\nabla_{\mu_k} L = \vec{0}$$

$$\nabla_{M_K} L = \nabla_{M_K} \sum_{n=1}^N \log \left[ \sum_{j=1}^K \pi_j N(x_n | \mu_j, \Sigma_j) \right]$$

$$= \sum_{n=1}^N \frac{\pi_K \nabla_{M_K} [N(x_n | \mu_K, \Sigma_K)]}{\left[ \sum_{j=1}^K \pi_j N(x_n | \mu_j, \Sigma_j) \right]}$$

$$\frac{d}{dx} \log(x) = \frac{1}{x}$$

$$\frac{d}{dx} \log(f(x)) = \frac{1}{f(x)} \cdot \frac{df(x)}{dx}$$

$$\nabla_{M_K} \frac{1}{(2\pi)^{d/2} \sqrt{|\Sigma_K|}} e^{-\frac{1}{2} (x_n - \mu_K)^T \Sigma_K^{-1} (x_n - \mu_K)} \parallel \frac{d}{dx} e^{f(x)} = e^{f(x)} \cdot \frac{df(x)}{dx}$$

$$= - \frac{1}{(2\pi)^{d/2} \sqrt{|\Sigma_K|}} e^{-\frac{1}{2} (x_n - \mu_K)^T \Sigma_K^{-1} (x_n - \mu_K)} \cdot \nabla \left[ -\frac{1}{2} (x_n - \mu_K)^T \Sigma_K^{-1} (x_n - \mu_K) \right]$$

$$\nabla_{M_K} L = \sum_{n=1}^N \left( \frac{\pi_K N(x_n | \mu_K, \Sigma_K)}{\left[ \sum_{j=1}^K \pi_j N(x_n | \mu_j, \Sigma_j) \right]} \right) \nabla \left[ -\frac{1}{2} (x_n - \mu_K)^T \Sigma_K^{-1} (x_n - \mu_K) \right] = 0$$

$$A^T = A, A \in \mathbb{R}^{n \times n}$$

$$\nabla_x x^T A x = 2Ax$$

$$= \sum_{n=1}^N \underbrace{\gamma(n_K)}_{=1} \nabla \left[ -\frac{1}{2} (x_n - \mu_K)^T \Sigma_K^{-1} (x_n - \mu_K) \right] = 0$$

$$= \left( \sum_{n=1}^N \sum_{k=1}^K \gamma(n_K) \right) \sum_{n=1}^N (x_n - \mu_K) = 0$$

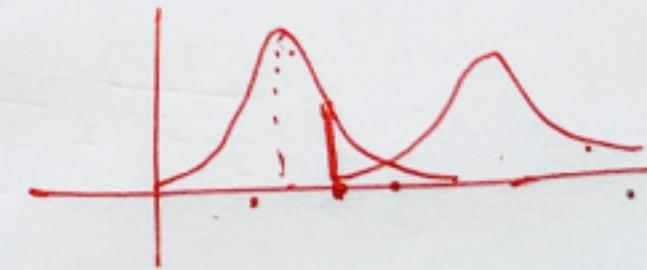
$$\sum_{n=1}^N \gamma(z_{nk}) (x_n - \mu_k) = 0$$

$$\sum_{n=1}^N \gamma(z_{nk}) x_n = \mu_k \sum_{n=1}^N \gamma(z_{nk})$$

$$\boxed{\mu_k = \frac{1}{N_k} \sum_{n=1}^N \gamma(z_{nk}) \cdot x_n}$$

$$\gamma(z_{nk}) = \underline{f(z_k=1 | x_n)}$$

$$N_k = \sum_{n=1}^N \gamma(z_{nk})$$



$$\nabla_{\Sigma_k} L = 0$$

$$\nabla_{\Sigma_k} \sum_{n=1}^N \log \left[ \sum_{j=1}^K \pi_j N(x_n | \mu_j, \Sigma_j) \right]$$

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

$$\cancel{\sum_k} = \frac{1}{(2\pi)^{d/2}} e^{-\frac{1}{2}(x_n - \mu_k)^T \Sigma_k^{-1} (x_n - \mu_k)} \left[ -\frac{1}{2} \frac{1}{\sqrt{\Sigma_k}} + \frac{1}{2} \frac{1}{\sqrt{|\Sigma_k|}} \sum_k^T (x_n - \mu_k) (\mu_k - \mu_k)^T \Sigma_k^{-1} \right]$$

$$= \frac{1}{2} \mathcal{N}(x_n | \mu_k, \Sigma_k) \left[ -\Sigma_k^{-1} + \Sigma_k^{-1} (x_n - \mu_k) (x_n - \mu_k)^T \Sigma_k^{-1} \right]$$

$$\nabla_{\Sigma_k} L = \sum_{k=1}^N \frac{\Pi_k \mathcal{N}(x_n | \mu_k, \Sigma_k)}{\sum_{j=1}^K \Pi_j \mathcal{N}(x_n | \mu_j, \Sigma_j)} \left[ -\Sigma_k^{-1} + \Sigma_k^{-1} (x_n - \mu_k) (x_n - \mu_k)^T \Sigma_k^{-1} \right] = 0$$

$$\Leftrightarrow \sum_{n=1}^N r(z_{nk}) \left[ -\Sigma_k^{-1} + \Sigma_k^{-1} (x_n - \mu_k) (x_n - \mu_k)^T \Sigma_k^{-1} \right] = 0$$

$$\sum_k \cancel{\sum_n} \sum_{n=1}^N r(z_{nk}) \Sigma_k^{-1} \Sigma_k = \sum_k \sum_{n=1}^N r(z_{nk}) (x_n - \mu_k) (x_n - \mu_k)^T \Sigma_k^{-1} \Sigma_k$$

$$\sum_k \left( \sum_{n=1}^N r(z_{nk}) \right) = \sum_{n=1}^N r(z_{nk}) (x_n - \mu_k) (x_n - \mu_k)^T$$

$$\Sigma_k = \frac{1}{N_k} \sum_{n=1}^N r(z_{nk}) (x_n - \mu_k) (x_n - \mu_k)^T$$

$$\nabla_{\Sigma_k} (\mathcal{N}(x_n | \mu_k, \Sigma_k))$$

$$\nabla_{\Sigma_k} \left[ \frac{1}{(2\pi)^{d/2} \sqrt{|\Sigma_k|}} e^{-\frac{1}{2} [x_n - \mu_k]^T \Sigma_k^{-1} [x_n - \mu_k]} \right]$$

$$\begin{aligned} \frac{d}{dx}(f \cdot g) \\ = f \frac{dg}{dx} + g \frac{df}{dx} \end{aligned}$$

$$\begin{aligned} \frac{1}{(2\pi)^{d/2}} \nabla_{\Sigma_k} \left[ \nabla_{\Sigma_k} \left[ \frac{1}{\sqrt{|\Sigma_k|}} \right] e^{-\frac{1}{2} [x_n - \mu_k]^T \Sigma_k^{-1} [x_n - \mu_k]} \right. \\ \left. + \frac{1}{\sqrt{|\Sigma_k|}} \nabla_{\Sigma_k} e^{-\frac{1}{2} [x_n - \mu_k]^T \Sigma_k^{-1} [x_n - \mu_k]} \right] \end{aligned}$$

$$\nabla_{\Sigma_k} \left[ \frac{d}{d|\Sigma_k|} |\Sigma_k|^{-k_2} \right] \quad || \quad \frac{d}{dx} |x| = |x| x^{-1}$$

$$= -\frac{1}{2} |\Sigma_k|^{-\frac{3}{2}} \cdot |\Sigma_k| \cdot \Sigma_k^{-1}$$

$$= -\frac{1}{2} |\Sigma_k|^{-\frac{1}{2}} \Sigma_k^{-1}$$

$$\begin{aligned} \nabla_{\Sigma_k} \left[ e^{-\frac{1}{2} p^T \Sigma_k^{-1} p} \right] &= -\frac{1}{2} e^{-\frac{1}{2} p^T \Sigma_k^{-1} p} \cdot \nabla_{\Sigma_k} (p^T \Sigma_k^{-1} p) \\ &= +\frac{1}{2} e^{-\frac{1}{2} p^T \Sigma_k^{-1} p} \Sigma_k^{-1} p p^T \Sigma_k^{-1} \end{aligned}$$

$$\begin{aligned} \frac{d}{dx} a^T x^{-1} b \\ = -x^{-1} a b^T x^{-1} \end{aligned}$$

$$\frac{\partial}{\partial \pi_k} L(x|\mu, \Sigma, \pi) = 0$$

$$\sum_{j=1}^K \pi_j = 1$$

$$\begin{aligned} \frac{\partial}{\partial \pi_k} & \left[ \sum_{n=1}^N \log \left( \sum_{j=1}^K \pi_j N(x_n | \mu_j, \Sigma_j) \right) \right] \\ &= \sum_{n=1}^N \frac{N(x_n | \mu_j, \Sigma_j)}{\sum_{j=1}^K \pi_j N(x_n | \mu_j, \Sigma_j)} \end{aligned}$$

$$\sum_{j=1}^K \pi_j = 1$$

$$f = L + \lambda \left( \sum_{j=1}^K \pi_j - 1 \right)$$

$$\frac{\partial f}{\partial \pi_k} = \frac{\partial L}{\partial \pi_k} + \lambda = 0$$

$$\sum_{k=1}^K \sum_{n=1}^N \frac{N(x_n | \mu_j, \Sigma_j) \pi_k}{\sum_{j=1}^N \pi_j N(x_n | \mu_j, \Sigma_j)} = -\lambda \left( \sum_{k=1}^K \pi_k \right)^{-1}$$

$$\sum_{k=1}^K \sum_{n=1}^N \gamma(\pi_{nk}) = -\lambda = N \Rightarrow \lambda = -N$$

$$\sum_{n=1}^N \gamma(x_n) = -\lambda \sum_{n=1}^N \gamma(\pi_{nk})$$