

E - step:

$$Q(\theta) = E_{z|x} [L(\theta)]$$

M - step:

$$\hat{\theta} = \arg \max_{\theta} Q(\theta)$$

$$P(x|\theta) = (1-\alpha) \mathcal{N}(x|\mu_0, \Sigma_0) + \alpha \mathcal{N}(x|\mu_1, \Sigma_1)$$

$$\theta = (\alpha, \mu_0, \Sigma_0, \mu_1, \Sigma_1) \quad 0 \leq \alpha \leq 1$$

$$D = \{x_1, \dots, x_n\}$$

E-step:

$$\gamma_i^t = \Pr(z_i = 1 | x_i, \theta^t) = E_{z_i|x_i} [z_i] \quad 0 \leq \gamma_i^t \leq 1$$

M-step:

$$\hat{\mu}_0 = \frac{\sum_{i=1}^n x_i (1 - \gamma_i^t)}{\sum_{i=1}^n (1 - \gamma_i^t)}$$

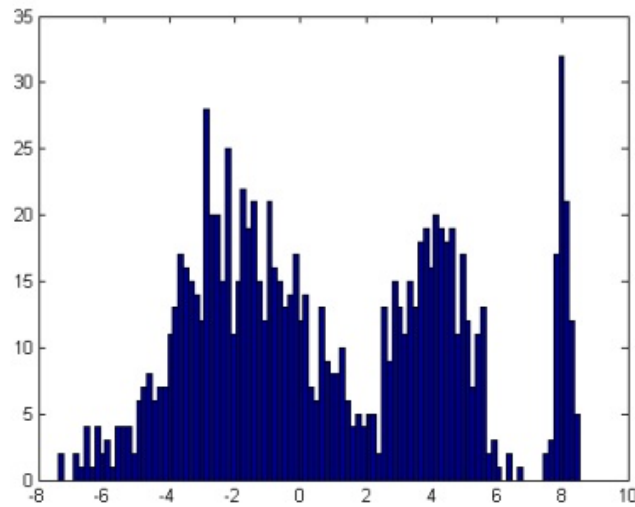
$$\hat{\alpha} = \frac{\sum_{i=1}^n \gamma_i^t}{n}$$

$$\hat{\Sigma}_1 = \frac{\sum_{i=1}^n \gamma_i^t (x_i - \mu_1)(x_i - \mu_1)^T}{\sum_{i=1}^n \gamma_i^t}$$

$$x_i \quad (z_i)$$

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GMM: 1-D Example



$$\mu_1 = -2$$

$$\sigma_1 = 2$$

$$\pi_1 = 0.6$$

$$\mu_2 = 4$$

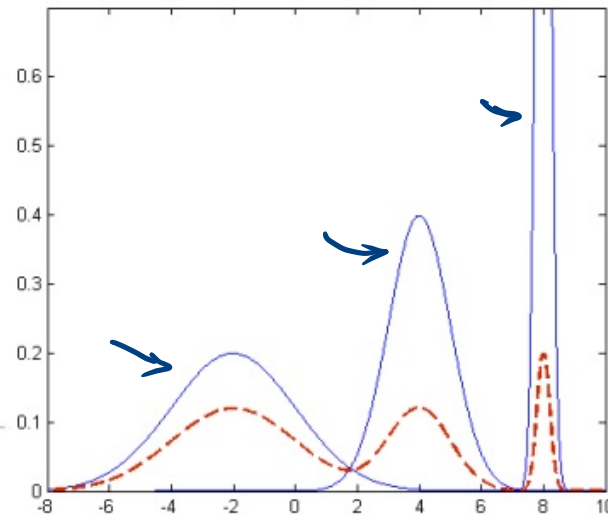
$$\sigma_2 = 1$$

$$\pi_2 = 0.3$$

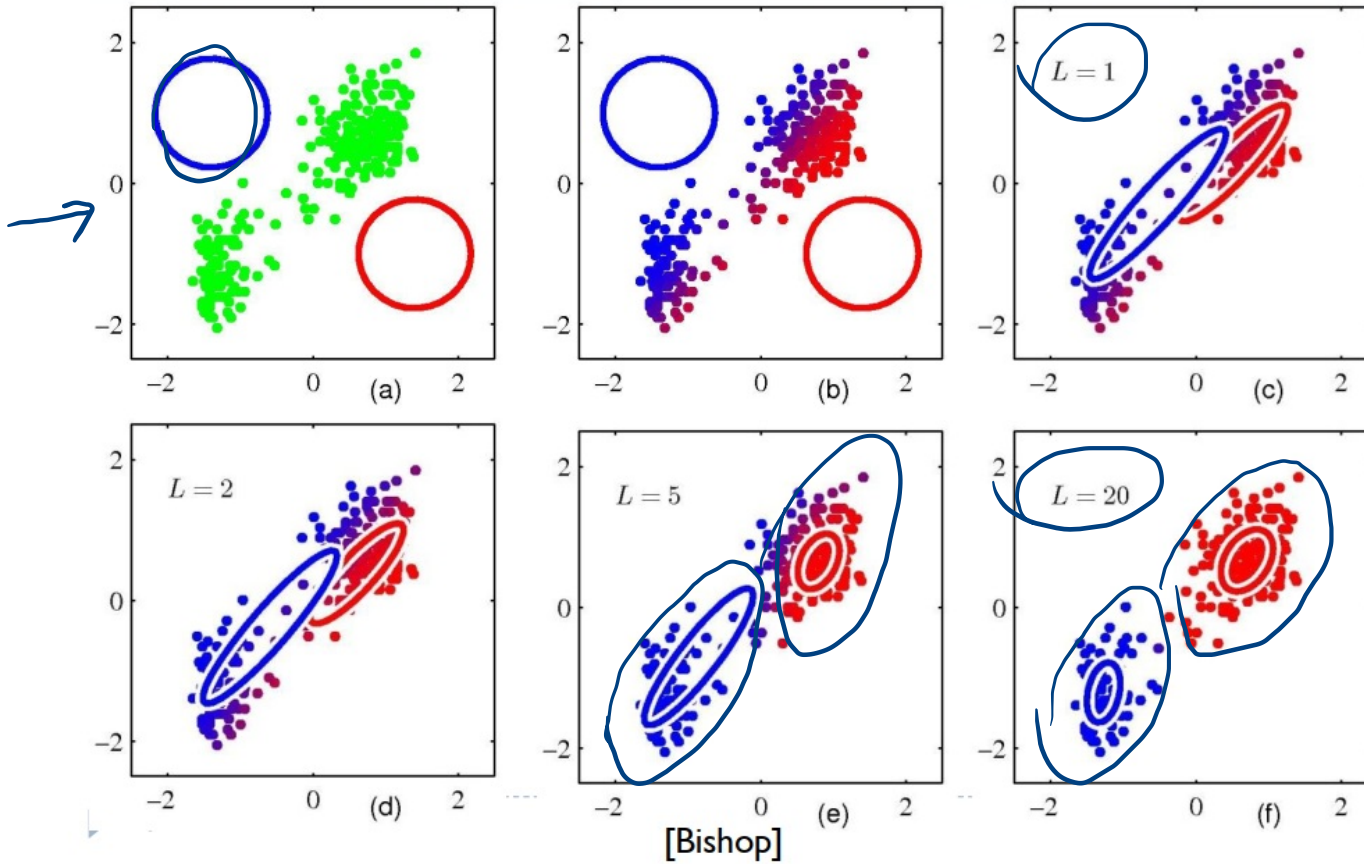
$$\mu_3 = 8$$

$$\sigma_3 = 0.2$$

$$\pi_3 = 0.1$$



EM & GMM: Example



$$\gamma_i^t \geq \frac{1}{2}$$

$$D = \{x_1, \dots, x_n\}$$

$$H = \{z_1, \dots, z_n\}$$

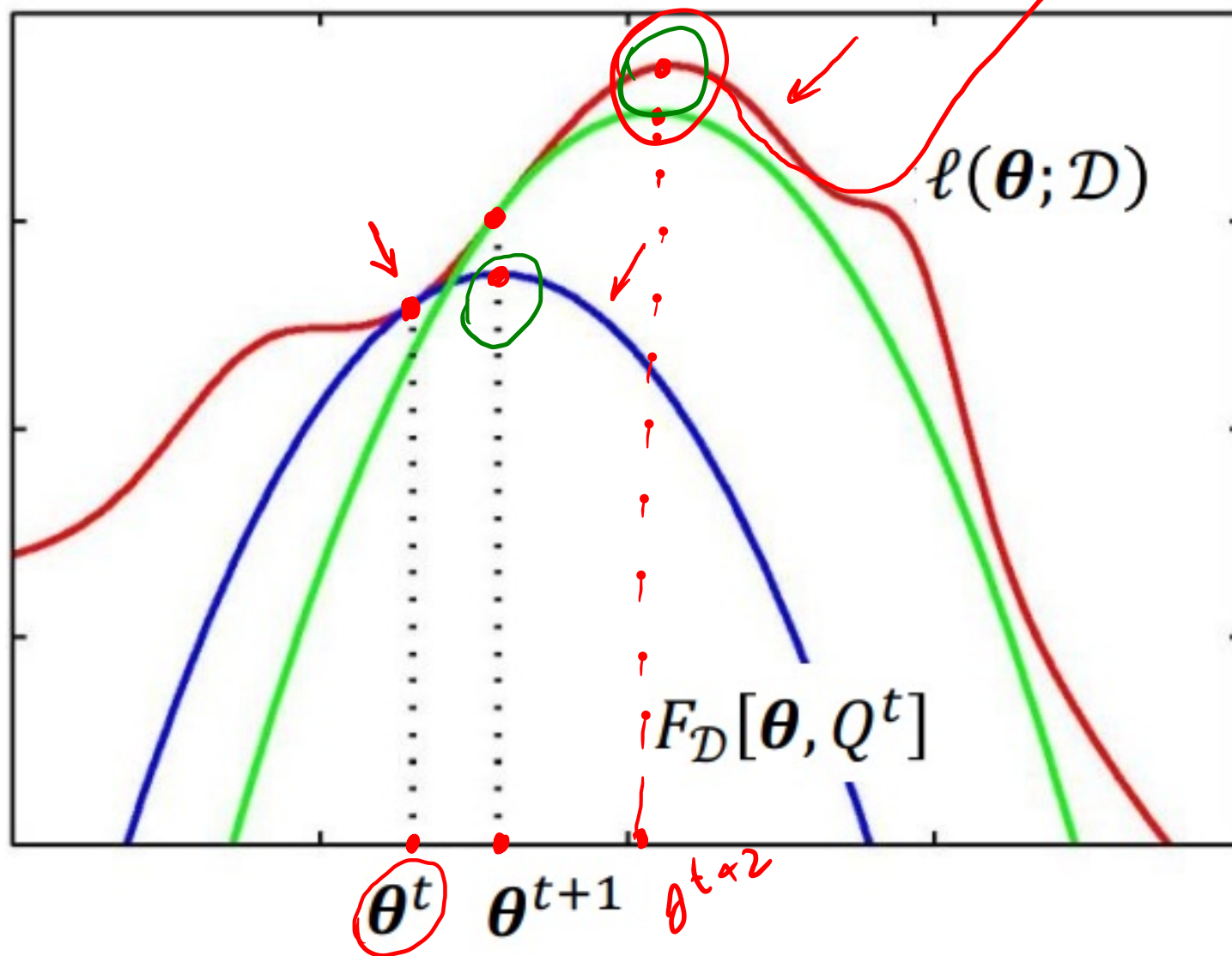
$$\hat{\theta}_{ML} = \arg \max_{\theta} \log P(D|\theta) = \arg \max_{\theta} \log P(x_1, \dots, x_n | \theta)$$

Likelihood

$$= \arg \max_{\theta} \sum_{i=1}^n \log \underline{P(x_i | \theta)} = \arg \max_{\theta} \sum_{i=1}^n \log \sum_{z_i} P(x_i, z_i | \theta)$$

log likelihood

$$\theta^{t+1} = \theta^t + \eta \nabla_{\theta} \mathcal{L}(\theta^t)$$

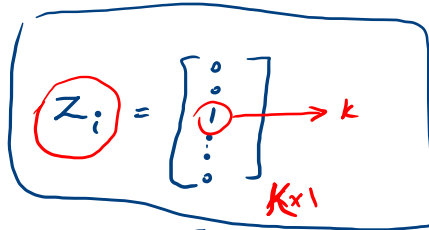


$$\underline{P(x|\theta)} = \sum_{k=1}^K \alpha_k \mathcal{N}(x|\mu_k, \Sigma_k) \quad \sum_{k=1}^K \alpha_k = 1$$

$$\alpha_k \geq 0$$

$$D = \{x_1, x_2, x_3, \dots, x_n\}$$

$$\rightarrow H = \{z_1, z_2, z_3, \dots, z_n\}$$



$$P(x_i, z_i|\theta) = \prod_{k=1}^K \left(\alpha_k \mathcal{N}(x_i|\mu_k, \Sigma_k) \right)^{z_{ik}} \quad \leftarrow$$

~~$$= \sum_{k=1}^K z_k \alpha_k \mathcal{N}(x|\mu_k, \Sigma_k)$$~~

$$L(\theta) = \log P(D, H|\theta) = \log P(x_1, z_1, x_2, z_2, \dots, x_n, z_n|\theta) = \log \prod_{i=1}^n P(x_i, z_i|\theta) = \sum_{i=1}^n \log P(x_i, z_i|\theta)$$

$$= \sum_{i=1}^n \sum_{k=1}^K z_{ik} (\log \alpha_k + \log \mathcal{N}(x_i|\mu_k, \Sigma_k))$$

$$Q(\theta) = E_{z_i} [L(\theta)] = \sum_{i=1}^n \sum_{k=1}^K E[z_{ik}] (\log \alpha_k + \log \mathcal{N}(x_i|\mu_k, \Sigma_k))$$

γ_{ik}^+

$$E [z_{ik}] = \Pr(z_{ik} = 1 | x_i, \theta^t)$$

$z_i | x_i$

$$= \frac{p(x_i | z_{ik} = 1, \theta^t) P(z_{ik} = 1 | \theta^t)}{p(x_i | \theta^t)}$$

$$= \frac{\mathcal{N}(x_i | \mu_k^t, \Sigma_k^t) \alpha_k^t}{\sum_{k=1}^K \alpha_k \mathcal{N}(x_i | \mu_k^t, \Sigma_k^t)} = \gamma_{ik}^t$$

$$\frac{\partial Q(\theta)}{\partial \alpha_k} = 0$$

$$\hat{\alpha}_k = \frac{\sum_{i=1}^n \gamma_{ik}^t}{n}$$

$$\hat{\mu}_k = \frac{\sum_{i=1}^n \gamma_{ik}^t x_i}{\sum_{i=1}^n \gamma_{ik}^t}$$

$$Q(\theta) = \sum_{i=1}^n \left(\sum_{k=1}^K \gamma_{ik}^t \right) \left(\log \alpha_k + \log N(x_i | \mu_k, \Sigma_k) \right) - \lambda \left(\sum_{k=1}^K \alpha_k - 1 \right)$$

$$\frac{\partial Q(\theta)}{\partial \alpha_j} = \sum_{i=1}^n \gamma_{ij}^t \frac{1}{\alpha_j} - \lambda = 0 \Rightarrow \alpha_j = \frac{\sum_{i=1}^n \gamma_{ij}^t}{\lambda} \quad j=1, \dots, K$$

$$1 = \sum_{j=1}^K \alpha_j = \frac{\sum_{j=1}^K \sum_{i=1}^n \gamma_{ij}^t}{\lambda}$$

$$\alpha_j = \frac{\sum_{i=1}^n \gamma_{ij}^t}{\sum_{j=1}^K \sum_{i=1}^n \gamma_{ij}^t} = \frac{\sum_{i=1}^n \gamma_{ij}^t}{n}$$