

HMM :-

$$\textcircled{1} \quad A_{jk} = p(z_{nk}=1 \mid z_{n-1}=j)$$

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

$$\theta = \{\pi, A, \phi\}$$

$$p(z_1|\pi) = \prod_{k=1}^K \pi_k^{z_{1k}} ; \quad \sum_k \pi_k = 1$$

for an HMM

$$p(x, z|\theta) = p(z_1|\pi) \prod_{n=2}^N p(z_n|z_{n-1}, A) \prod_{m=1}^M p(x_m|z_m, \phi)$$

The only info we have at hand are $X = \{\bar{x}_1, \bar{x}_2, \dots, \bar{x}_N\}$

$$\Rightarrow p(x|\theta) = \sum_z p(x, z|\theta)$$

This leads to standard ML solution. \Rightarrow Applying EM

algorithm we end up with

$$Q(\theta, \theta^i) = \mathbb{E}_{Z|X, \theta} \cdot \log p(z, x|\theta)$$

$$= \sum_z p(z|x, \theta) \log p(x, z|\theta)$$

$$r(\bar{z}_n) = p(\bar{z}_n | z, \theta^{(i)})$$

$$\ell_p(z_{n-1}, z_n) = p(z_{n-1}, z_n | z, \theta)$$

$$r(z_n) = p(z_{nk=1} | x, \theta^{(i)})$$

$$\ell_p(z_{n-1}, z_{nk}) = p(z_{n+j}=1, z_{nk}=1 | x, \theta^{(i)})$$

$$p(z, x | \theta) = p(z | \pi) \prod_{n=2}^N p(z_n | z_{n-1}, A) \prod_{m=1}^N p(x_m | z_m, \phi)$$

$$\log p(z, x | \theta) = \log(p(z | \pi)) + \sum_{n=2}^N \log p(z_n | z_{n-1}, A) + \sum_{m=1}^N \log p(x_m | z_m, \phi)$$

$$\Rightarrow Q(\theta, \theta^{(i)}) = \sum_z p(z | x, \theta) \log p(z | \pi) + \sum_{n=2}^N \log p(z | x, \theta) \log p(z_n | z_{n-1}, A) + \sum_{m=1}^N \log p(z | x, \theta) \log p(x_m | z_m, \phi)$$

$$\left\{ \begin{array}{l} p(z | \pi) = \prod_{k=1}^K \pi_k^{z_{ik}} \\ \log p(z | \pi) = \sum_{k=1}^K \log \pi_k^{z_{ik}} \end{array} \right\}$$

$$\Rightarrow Q(\theta, \theta^{(i)}) \text{ First term} = \sum_z p(z | x, \theta) \sum_{k=1}^K \log \pi_k = \sum_{k=1}^K r(z_{ik}) \log \pi_k$$

Second term:

$$\begin{aligned} \ell_p(z_{n-1}, z_n) &= p(z_{n-1}, z_n | z, \theta) \\ \Rightarrow p(z_n | z, \theta) &= \sum_{z_{n-1}} \ell_p(z_{n-1}, z_n) \end{aligned}$$

↳ Second term

$$= \sum_{z} \sum_{n=2}^N p(z|x,\theta) \log p(\bar{z}_n | \bar{z}_{n-1}, A)$$
$$= \sum_{n=2}^N \sum_{j=1}^K \sum_{k=1}^K \epsilon_q(z_{n,j}; z_{nk}) \log A_{jk}$$

↳ The third term

$$\sum_z \sum_{m=1}^N p(z|x,\theta) \log p(\bar{x}_m | \bar{z}_m, \phi)$$

$$= \sum_{m=1}^N \sum_{k=1}^K \gamma(z_{mk}) \log p(x_m | \phi_k)$$

$$\therefore Q(\theta, \theta^{old}) = \sum_{k=1}^K \gamma(z_{mk}) \log \pi_k + \sum_{n=2}^N \sum_{j=1}^K \sum_{k=1}^K \epsilon_q(z_{nj}, z_{nk}) \log A_{jk}$$

$$+ \sum_{k=1}^K \sum_{n=1}^N \gamma(z_{nk}) \log (p(x_n | \phi_k))$$

D) Emission probabilities AKA ~~hidden~~ ~~variables~~

$$p(\bar{z}_n | \phi_k) \sim \mathcal{N}(\mu_k, \Sigma_k)$$

parameters we have at hand are $\mu_k, \Sigma_k, A_{jk}, \pi_k$

for π_k $\frac{\partial Q(O, \theta^{old})}{\partial \pi_k} = 0$

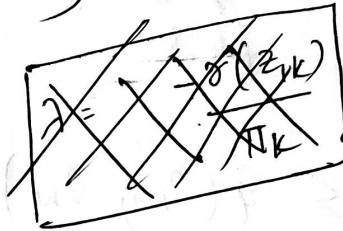
also $\sum_k \pi_k = 1$

ignore other terms

\Rightarrow Introduce Lagrange multipliers and

$$\Rightarrow \frac{\partial}{\partial \pi_k} \left[\sum_{k=1}^K \gamma(z_{ik}) \log \pi_k + \lambda \left[\sum_k \pi_k - 1 \right] \right] = 0.$$

$$\Rightarrow \frac{1}{\pi_k} \left[\gamma(z_{ik}) \right] + \lambda = 0 \quad \text{--- (1)}$$



π_k on both sides

multiply

$$\gamma(z_{ik}) + \lambda \pi_k = 0$$

$$\sum_{k=1}^K \gamma(z_{ik}) + \lambda \sum_{k=1}^K \pi_k = 0$$

$$\lambda = -\sum_{k=1}^K \gamma(z_{ik}) \quad \text{--- (2)}$$

Subst. (2) in (1) we have

$$\frac{1}{\pi_k} [\gamma(z_{ik})] = -\sum_{k=1}^K \gamma(z_{ik})$$

$$\Rightarrow \boxed{\pi_k = \frac{\gamma(z_{ik})}{\sum_k \gamma(z_{ik})}}$$

$$N^k \frac{\partial Q(\theta, \theta^{old})}{\partial A_{jk}}$$

$$= \frac{\partial}{\partial A_{jk}} \left[\sum_{n=2}^N \sum_{j=1}^K \sum_{k=1}^K \epsilon_q(z_{nj}, z_{nk}) (\log A_{jk}) \right] = 0.$$

and $\quad A_{jk}^o = p(z_{nk}=1 | z_{nj}=1)$

$\Rightarrow \boxed{\sum_{k=1}^K A_{jk}^o = 1}$

by probability definition

$$\Rightarrow \frac{\partial}{\partial A_{jk}^o} \left[\sum_n \sum_j \sum_k \epsilon_q(z_{nj}, z_{nk}) \log A_{jk} + \lambda \left[\sum_k A_{jk}^{-1} \right] \right] = 0$$

$$\sum_{n=2}^N \left(\frac{1}{A_{jk}^o} \right) \left(\epsilon_q(z_{nj}, z_{nk}) \right) + \lambda = 0$$

$$\Rightarrow \sum_{n=2}^N \sum_{k=1}^K \epsilon_q(z_{nj}, z_{nk}) + \lambda = 0$$

$$\boxed{\lambda = - \sum_{n=2}^N \sum_{k=1}^K \epsilon_q(z_{nj}, z_{nk})}$$

$$\rightarrow A_{jk} = \sum_{n=2}^N \epsilon_p(z_{n-j}, z_{nk})$$

$$A_{jk} = \frac{\sum_{n=2}^N \epsilon_p(z_{n-j}, z_{nk})}{\sum_{n=2}^N \sum_{k=1}^K \epsilon_p(z_{n-j}, z_{nk})}$$

for $\mu_k \in \Sigma_k$

$$Q(\theta, \theta^{old}) \Big|_{\mu_k \in \Sigma_k} = \sum_{k=1}^K \sum_{n=1}^N \gamma(z_{nk}) \left[\log \left(\frac{1}{\sqrt{(2\pi)^k |\Sigma|}} \right) - \frac{1}{2} \frac{(\mathbf{x}_n - \mu_k)^T \Sigma^{-1}}{(\mathbf{x}_n - \mu_k)} \right]$$

$$= \sum_{k=1}^K \sum_{n=1}^N \left[\gamma(z_{nk}) \log \left(\frac{1}{\sqrt{(2\pi)^k |\Sigma|}} \right) - \cancel{\gamma(z_{nk})} \cancel{\left(\frac{(\mathbf{x}_n - \mu_k)^T \Sigma^{-1}}{2} \right)} \right.$$

$$- \cancel{\frac{\gamma(z_{nk})}{2}} \left[(\mathbf{x}_n^T \Sigma^{-1} - \mu_k^T \Sigma^{-1})(\mathbf{x}_n - \mu_k) \right]$$

$$\frac{\partial Q(\theta, \theta^{old})}{\partial \mu_k} = \frac{\partial}{\partial \mu_k} \left(\sum_{k=1}^K \sum_{n=1}^N \frac{-\gamma(z_{nk})}{2} \left[\mathbf{x}_n^T \Sigma^{-1} \mathbf{x}_n - \mathbf{x}_n^T \Sigma^{-1} \mu_k - \mu_k^T \Sigma^{-1} \mathbf{x}_n + \mu_k^T \Sigma^{-1} \mu_k \right] \right)$$

$$= \sum_{n=1}^N -\frac{\gamma(z_{nk})}{2} \left[-(\mathbf{x}_n^T \Sigma^{-1})^T - (\Sigma^{-1} \mathbf{x}_n) + (\Sigma^{-1} + \Sigma^{-1 T}) \mu_k \right]$$

$$= \sum_{n=1}^N -\frac{\gamma(z_{nk})}{2} \left[-(\mathbf{x}_n^T \Sigma^{-1})^T - (\Sigma^{-1} \mathbf{x}_n) + (\Sigma^{-1} + \Sigma^{-1 T}) \mu_k \right] = 0$$

$$\Rightarrow \sum_{n=1}^N \gamma(z_{nk}) \left[(\mathbf{x}_n^\top \boldsymbol{\Sigma}^\dagger)^\top + (\boldsymbol{\Sigma}^\dagger \mathbf{x}_n) \right] = \sum_{n=1}^N (\boldsymbol{\Sigma}^\dagger + \boldsymbol{\Sigma}) \mathbf{x}_n^\top \gamma(z_{nk})$$

$$\Rightarrow \sum_{n=1}^N \gamma(z_{nk}) \left[\mathbf{x}_n \left(\cancel{\boldsymbol{\Sigma}^\dagger + \boldsymbol{\Sigma}^{-\top}} \right) \right] = \mu_k \sum_{n=1}^N (\boldsymbol{\Sigma}^\dagger + \boldsymbol{\Sigma}^{-\top}) \gamma(z_{nk})$$

$$\Rightarrow \boxed{\mu_k = \frac{\sum_{n=1}^N \mathbf{x}_n \gamma(z_{nk})}{\sum_{n=1}^N \gamma(z_{nk})}}$$

~~$$(\boldsymbol{\Sigma}^\dagger + \boldsymbol{\Sigma}^{-\top})$$~~

WJ for $\boldsymbol{\Sigma}_k$

$$\frac{\partial}{\partial \boldsymbol{\Sigma}_k} \left[\sum_{n=1}^N \sum_{k=1}^K \gamma(z_{nk}) \left[\log \left(\frac{1}{\sqrt{(2\pi)^k |\boldsymbol{\Sigma}|}} \right) - \frac{1}{2} (\mathbf{x}_n - \mu_k)^\top \boldsymbol{\Sigma}^\dagger (\mathbf{x}_n - \mu_k) \right] \right]$$

$$\frac{\partial}{\partial \boldsymbol{\Sigma}} (\mathbf{a}^\top \mathbf{x}^\top \mathbf{b}) = -\mathbf{X}^{-T} \mathbf{a} \mathbf{b}^\top \mathbf{X}^{-T}$$

here $\mathbf{b} = \mathbf{a} = (\mathbf{x}_n - \mu_k)$

$$\frac{\partial \alpha}{\partial \boldsymbol{\Sigma}_k} = -\boldsymbol{\Sigma}^{-T} (\mathbf{x}_n - \mu_k) (\mathbf{x}_n - \mu_k)^\top \boldsymbol{\Sigma}^{-T}$$

~~$$\frac{\partial}{\partial \boldsymbol{\Sigma}} -\frac{1}{2} \log |\boldsymbol{\Sigma}| = -\frac{1}{2} [(\mathbf{X}^{-1})^\top]$$~~

$$\frac{\partial}{\partial \boldsymbol{\Sigma}} -\frac{1}{2} \log |\boldsymbol{\Sigma}| = -\frac{1}{2} [(\mathbf{X}^{-1})^\top]$$

$$\therefore \frac{\partial}{\partial \Sigma_k} Q(\theta, \theta^{old}) = 0$$

$$\Rightarrow \sum_{n=1}^N \gamma(z_{nk}) \left[\cancel{\frac{1}{N} \sum} - \frac{1}{N} \sum \Sigma^T (x_n - \mu_k) (x_n - \mu_k)^T \cancel{\frac{1}{N}} \right] = 0$$

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

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