

$$p(\bar{x} | \bar{\mu}, \Sigma, \bar{\pi}) = \sum_k \pi_k N(\bar{x} | \bar{\mu}_k, \Sigma_k)$$

$$L = \prod_n p(\bar{x}_n | \bar{\pi}, \bar{\mu}, \Sigma) = \prod_n \left( \sum_k \pi_k N(\bar{x}_n | \bar{\mu}_k, \Sigma_k) \right)$$

E-war: 
$$E z_{nk} = \frac{\pi_k N(\bar{x}_n | \bar{\mu}_k, \Sigma_k)}{\sum_l \pi_l N(\bar{x}_n | \bar{\mu}_l, \Sigma_l)}$$

$$p(\bar{x}_n \in C_k | \bar{x}_n \in C, \bar{\pi}, \bar{\mu}, \Sigma)$$

$$z_{nk} = 1 \text{ iff } \bar{x}_n \in C_k$$

$$\bar{\pi}, \bar{\mu}, \Sigma \xrightarrow{\max}$$

M-war: 
$$\bar{\pi}, \bar{\mu}, \Sigma = \operatorname{argmax}_{\bar{\pi}, \bar{\mu}, \Sigma} p(\bar{x}, z | \bar{\pi}, \bar{\mu}, \Sigma) = \operatorname{argmax}_{\bar{\pi}, \bar{\mu}, \Sigma} \prod_n \left( \prod_k \pi_k N(\bar{x}_n | \bar{\mu}_k, \Sigma_k) \right)$$

$$\sum_{n,k} \left[ E z_{nk} \ln \pi_k + E z_{nk} \ln N(\bar{x}_n | \bar{\mu}_k, \Sigma_k) \right]$$

$$f(x, z) \rightarrow \max$$

$$p(x|\theta) \xrightarrow{\theta} \max$$

$$p(x, z|\theta) \xrightarrow{\theta} \max$$

$$\ln p(x|\theta) - \ln p(x|\theta^{(m)}) = \ln \int p(x, z|\theta) dz - \ln \int p(x, z|\theta^{(m)}) dz = E_{p(z|x, \theta^{(m)})} [\ln p(x, z|\theta) - \ln p(x, z|\theta^{(m)})]$$

$$\ln E \geq E \ln$$

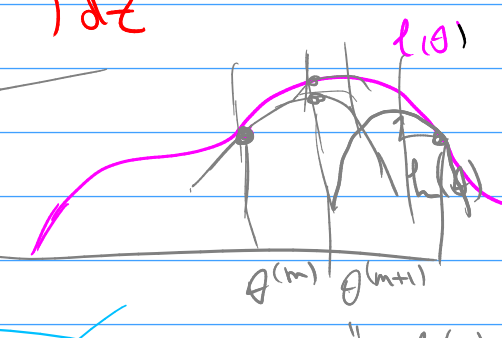
$$= \ln \int \frac{p(x, z|\theta)}{p(z|x, \theta^{(m)})} p(z|x, \theta^{(m)}) dz - \ln \int p(x, z|\theta^{(m)}) dz$$

$$\geq \int \ln \frac{p(x, z|\theta)}{p(z|x, \theta^{(m)})} p(z|x, \theta^{(m)}) dz - \ln p(x|\theta^{(m)})$$

$$l(\theta) = \ln p(x|\theta) - \ln p(x|\theta^{(m)}) \Rightarrow \int \ln \frac{p(x, z|\theta)}{p(z|x, \theta^{(m)})} p(z|x, \theta^{(m)}) dz$$

$$l(\theta, \theta^{(m)})$$

$$l(\theta, \theta^{(m)}) = \int \ln \frac{p(x, z|\theta)}{p(x, z|\theta^{(m)})} p(z|x, \theta^{(m)}) dz$$



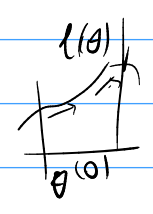
$$= \int \ln p(x, z|\theta) p(z|x, \theta^{(m)}) dz - \int \ln p(x, z|\theta^{(m)}) p(z|x, \theta^{(m)}) dz$$

$$Q(\theta, \theta^{(m)}) = \int \ln p(x, z|\theta) p(z|x, \theta^{(m)}) dz =$$

for  $m=0, \dots$  
$$= E_{p(z|x, \theta^{(m)})} [\ln p(x, z|\theta)]$$

E-war: maximum  $Q(\theta, \theta^{(m)})$

M-war:  $\theta^{(m+1)} := \operatorname{argmax}_{\theta} Q(\theta, \theta^{(m)})$



$$\theta^{(0)} = \left( \underbrace{\pi_k^{(0)}}_{\text{Data}}, \underbrace{\bar{x}_k^{(0)}}_{\text{Data}}, \underbrace{\xi_k^{(0)}}_{\text{d.I.}} \right), k=1-K$$

$$\bar{z}_n = (\dots z_{nk} \dots)$$

$$z_{nk} = 1 \text{ iff } \bar{x}_n \in C_k$$

$$p(x, z | \theta) = \prod_n \prod_k \left( \pi_k N(\bar{x}_n | \bar{\mu}_k, \xi_k) \right)^{z_{nk}}$$

$$\ln p(x, z | \theta) = \sum_n \sum_k z_{nk} \cdot \ln(\pi_k N(\bar{x}_n | \bar{\mu}_k, \xi_k))$$

$$Q(\theta, \theta^{(m)}) = \mathbb{E}_{z|x, \theta^{(m)}} [\ln p(x, z | \theta)] =$$

$$= \mathbb{E}_{z|x, \theta^{(m)}} \left[ \sum_n \sum_k z_{nk} \ln(\pi_k N(\bar{x}_n | \bar{\mu}_k, \xi_k)) \right] =$$

$$= \sum_n \sum_k \left( \ln \pi_k N(\bar{x}_n | \bar{\mu}_k, \xi_k) \right) \cdot \mathbb{E}_{z|x, \theta^{(m)}} [z_{nk}]$$

$$\theta^{(m+1)} = \underset{\theta}{\operatorname{argmax}} Q(\theta, \theta^{(m)})$$

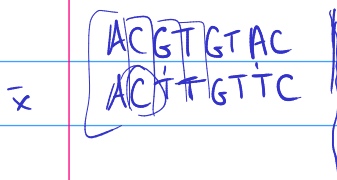
$$\Rightarrow \frac{p(\bar{x}_n, z_{nk} | \theta^{(m)})}{\sum_l p(\bar{x}_n, z_{nl} | \theta^{(m)})} =$$

$$= \frac{\pi_k N(\bar{x}_n | \bar{\mu}_k, \xi_k)}{\sum_l \pi_l N(\bar{x}_n | \bar{\mu}_l, \xi_l)}$$

$$y = (x, \underline{z})$$

$$p(x, z | \theta) = \prod_n \theta_i \varphi_i$$

$$p(x | \theta) = \prod_n \left( \sum_k \theta_k \varphi_k \right) \rightarrow \max$$



$$p_{i|s,k} = p(x_i = s | x \in C_k) \quad \sum_s p_{i|s,k} = 1 \quad \forall i,k$$

$$\pi_k = p(\bar{x} \in C_k)$$

$$E\text{-var: } \mathbb{E} z_{nk} = p(\bar{x}_n \in C_k) = \frac{\pi_k p(\bar{x}_n | C_k)}{\sum_l \pi_l p(\bar{x}_n | C_l)} = \frac{\pi_k \cdot \prod_i p_{i, x_{in}, k}}{\sum_l \pi_l \cdot \prod_i p_{i, x_{in}, l}}$$

$$M\text{-var: } \theta := \underset{\pi, P}{\operatorname{argmax}} \prod_n \prod_k \left( \pi_k \prod_i p_{i, x_{in}, k} \right)^{\mathbb{E} z_{nk}} =$$

$$= \underset{\pi, P}{\operatorname{argmax}} \sum_n \mathbb{E} z_{nk} \left( \ln \pi_k + \sum_i \ln p_{i, x_{in}, k} \right)$$

$$\left( \begin{array}{c} \vdots \\ l_i \\ \vdots \end{array} \right) \quad l_i = \frac{\log(\sum e^{l_i})}{n p_i \log(n p_i \log(n p_i \log(n p_i \log(n p_i \log(n p_i \dots))))}$$

$y = 1$  - "yours requires"

$$\eta(\bar{x}) = \bar{w}^T \bar{x} + w_0$$

$$p(y=1 | \bar{x}) = \sigma(\eta(\bar{x})) = \frac{1}{1 + e^{-\eta(\bar{x})}}$$

$$\pi = p(y=1)$$

$\bar{x}$	$z=1$
$z=1$	$z=0$

$$p(d|x) = \frac{n_{d|x}}{n_x} \quad \text{- prospective}$$

	$d=0$	$d=1$
$x=0$	$\pi_{00}$	$\pi_{01}$
$x=1$	$\pi_{10}$	$\pi_{11}$

$$\pi_0 = p(s=1 | d=0) \quad \text{- retrospective}$$

$$\pi_1 = p(s=1 | d=1)$$

$$\sigma(\eta(x)) = \frac{e^{\eta(x)}}{1 + e^{\eta(x)}}$$

$$\pi_{00} = p(d=0, x=0 | s=1) \leftarrow \sigma(\eta(x)) = \pi_1$$

$$\frac{\pi_{xd}}{\pi_{xd} + \pi_{x\bar{d}}} = p(d | s=1, x) = \frac{p(d|x) p(s=1 | d, x)}{p(d|x) p(s=1 | d, x) + p(\bar{d}|x) p(s=1 | \bar{d}, x)}$$

$$= \frac{\pi_1 \sigma(\eta(x))}{\pi_1 \sigma(\eta(x)) + \pi_0 (1 - \sigma(\eta(x)))} = \frac{1}{1 + \left(\frac{\pi_0}{\pi_1}\right) e^{-\eta(x)}}$$

$$\sigma(\eta^*(x)) = \frac{\pi_{xd}}{\pi_{xd} + \pi_{x\bar{d}}} = \frac{1}{1 + e^{-\eta(x) + \log \frac{\pi_0}{\pi_1}}} = \sigma(\eta^*(x))$$

$$\begin{aligned} & \text{|| } \underline{w^T x + w_0} \\ & \text{|| } \underline{w_0 - \log \frac{\pi_0}{\pi_1}} \end{aligned}$$

$$\begin{aligned} & \text{|| } \underline{\eta(x) - \log \frac{\pi_0}{\pi_1}} \\ & \text{|| } \underline{w^T x + w_0 + \log \frac{\pi_0}{\pi_1}} \end{aligned}$$

$$p(z=1 | s=1, \bar{x}) = p(z=1 | y=1, s=1, \bar{x}) p(y=1 | s=1, \bar{x}) + p(z=1 | y=0, s=1, \bar{x}) p(y=0 | s=1, \bar{x})$$

$$\text{|| } \frac{p(z=1, y=1 | s=1)}{p(y=1 | s=1)} = \frac{n_p / (n_p + n_u)}{n_p + \pi n_u} = \frac{n_p}{n_p + \pi n_u}$$

$n_p = \# \{z=1\}$   
 $n_u = \# \{z=0\}$

$$p(z=1 | s=1, \bar{x}) = \frac{n_p}{n_p + \pi n_u} \sigma(\eta^*(\bar{x}))$$

$$\begin{aligned} \frac{\pi_1}{\pi_0} &= \frac{p(s=1, y=1)}{p(s=1, y=0)} \\ &= \frac{n_p + \pi n_u}{(1 - \pi) n_u} \end{aligned}$$

$$\prod_n p(z_n | s=1, \bar{x}_n) \rightarrow \max$$

$$\text{E-wr: } \underline{y^{(m)}} = E y^{(m)} = \frac{e^{\eta(\bar{x})} + 1}{1 + e^{\eta(\bar{x})} + 2}$$

$$C \cdot \frac{1}{1+e^{-\eta^*}} =$$

M-var:  $\pi(y=1)$   
коррел. пер.  $\pi(\eta^*(\bar{x})) \sim y^{(m)}$

E-var:

$$\underline{\eta(\bar{x})} = \eta^*(\bar{x}) - \log \frac{n_p + \pi n_u}{\pi n_u}$$

$$y^{(m)} = \frac{e^{\eta(\bar{x})} + 1}{1 + e^{\eta(\bar{x})} + 2}$$