

$$\ln p(x) = \ln p(z, x) - \ln p(z|x)$$

$$\ln p(x) = \underbrace{\int \ln \frac{p(z, x)}{q(z)} q(z) dz}_{\substack{\text{L}(q) \\ \rightarrow \text{max}}} - \underbrace{\int \ln \frac{p(z|x)}{q(z)} q(z) dz}_{\substack{\text{KL}(q||p(z|x)) \\ \rightarrow \text{min}}}$$

$$q(z) = \prod_i q_i(z_i), \quad z_i \cap z_j = \emptyset$$

$$\ln q_j^*(z_j) = \mathbb{E}_{q^*}^* \left[ \ln p(x, z) \right] + \text{const}$$

$$p(\bar{x}) = \mathcal{N}(\bar{x} | \bar{\mu}, \Lambda^{-1}) \approx q(\bar{x}) = q_1(x_1) q_2(x_2)$$

KL(q||p)  
KL(p||q)

$$\begin{pmatrix} \lambda_{11} & \lambda_{12} \\ \lambda_{21} & \lambda_{22} \end{pmatrix}$$

$$\begin{aligned} \mu_1 &= \mu, & \mu_2 &= \mu \\ \tau_1 &= \lambda_{11}, & \tau_2 &= \lambda_{22} \end{aligned}$$

$$\frac{1}{\sigma^2} \tau = \frac{1}{\sigma^2}$$

$$D = \{x_1, \dots, x_N\} = \ln p(D, \mu, \tau)$$

$$p(\tau, \mu) \times p(D|\mu, \tau) \propto p(\tau, \mu|D)$$

$$\begin{aligned} p(\tau) p(\mu|\tau) &= \prod_{n=1}^N p(x_n|\mu, \tau) \\ &= \text{Gam}(\tau|a_0, b_0) \cdot \mathcal{N}(\mu|\mu_0, (\lambda_0 \tau)^{-1}) \end{aligned}$$

$$= \frac{1}{\Gamma(\cdot)} \tau^{a_0-1} e^{-b_0 \tau} \cdot \sqrt{\frac{\lambda_0 \tau}{2\pi}} e^{-\frac{\lambda_0 \tau}{2} (\mu - \mu_0)^2} = \left( \frac{\tau}{2\pi} \right)^{N/2} e^{-\frac{\tau}{2} \sum_n (x_n - \mu)^2}$$

$$q(\mu, \tau) = q_\mu(\mu) q_\tau(\tau) \approx p(\tau, \mu|D)$$

$$\ln q_{\mu}^*(\mu) = \mathbb{E}_{\tau} [\ln p(\tau, \mu, D)] + \text{const}$$

$$\ln q_{\tau}^*(\tau) = \mathbb{E}_{\mu} [\ln p(\tau, \mu, D)] + \text{const}$$

$$\ln p(\tau, \mu, D) = \underbrace{(a_0 - 1) \ln \tau - b_0 \tau + \frac{1}{2} \tau - \frac{\lambda_0 \tau}{2} (\mu - \mu_0)^2}_{\text{prior}} + \underbrace{\frac{N}{2} \ln \tau - \frac{\tau}{2} \sum_n (x_n - \mu)^2}_{\text{likelihood}} + \text{const}$$

$$\begin{aligned} \ln q_{\mu}^*(\mu) &= \mathbb{E}_{\tau} \left[ -\frac{\lambda_0 \tau}{2} (\mu - \mu_0)^2 - \frac{\tau}{2} \sum_n (x_n - \mu)^2 \right] + \text{const} = \\ &= -\frac{\mathbb{E}[\tau]}{2} \left( \lambda_0 (\mu - \mu_0)^2 + \sum_n (x_n - \mu)^2 \right) + \text{const} \end{aligned}$$

$$\lambda_0 \mu^2 - 2\lambda_0 \mu_0 \mu + N \cdot \mu^2 - 2 \left( \sum_n x_n \right) \mu$$

$$(\lambda_0 + N) \left( \mu - \frac{\lambda_0 \mu_0 + \sum_n x_n}{\lambda_0 + N} \right)^2$$

$$q_{\mu}^*(\mu) = \mathcal{N} \left( \mu \mid \frac{\lambda_0 \mu_0 + \sum_n x_n}{\lambda_0 + N}, (\lambda_0 + N) \cdot \mathbb{E}[\tau] \right)$$

$$\begin{aligned} \ln q_{\tau}^*(\tau) &= (a_0 - 1) \ln \tau - b_0 \tau + \frac{1}{2} \ln \tau + \frac{N}{2} \ln \tau - \frac{\tau}{2} \cdot \mathbb{E}_{\mu} \left[ \lambda_0 (\mu - \mu_0)^2 + \sum_n (x_n - \mu)^2 \right] + \text{const} \\ &= \left( a_0 - 1 + \frac{N+1}{2} \right) \ln \tau - \left( b_0 + \mathbb{E}[\tau] \right) \tau + \text{const} \end{aligned}$$

$$q_{\tau}^*(\tau) = \text{Gam} \left( \tau \mid a_0 + \frac{N+1}{2}, b_0 + \mathbb{E}_{\mu}[\tau] \right)$$

$$\mathbb{E}[\tau] = \frac{a}{b} = \frac{a_0 + \frac{N+1}{2}}{b_0 + \mathbb{E}[\tau]}$$

$\mu_0 = 0 = a_0 = b_0 = \lambda_0$   
non-informative  
priors

$$\mathbb{E}_{\mu} \left[ \lambda_0 (\mu - \mu_0)^2 + \sum_n (x_n - \mu)^2 \right] =$$

$$E[X^2] = \mu^2 + \frac{1}{\lambda}$$

$X \sim N(x|\mu, \sigma^2)$

$$= E_{\mu} \left[ \mu^2 (\lambda_0 + N) - 2\mu (\lambda_0 \mu_0 + \sum_n X_n) + \lambda_0 \mu_0^2 + \sum_n X_n^2 \right]$$

$$= (\lambda_0 + N) E_{\mu} [\mu^2] - 2(\lambda_0 \mu_0 + \sum_n X_n) E[\mu] + \lambda_0 \mu_0^2 + \sum_n X_n^2$$

$$= (\lambda_0 + N) \left( \frac{(\lambda_0 \mu_0 + \sum_n X_n)^2}{(\lambda_0 + N)} + \frac{1}{(\lambda_0 + N) E_{\tau}} \right) - \frac{2(\lambda_0 \mu_0 + \sum_n X_n)}{\lambda_0 + N} + \lambda_0 \mu_0^2 + \sum_n X_n^2$$

$$= \frac{(\lambda_0 \mu_0 + \sum_n X_n)^2}{\lambda_0 + N} + \frac{1}{E_{\tau}} - \frac{2(\lambda_0 \mu_0 + \sum_n X_n)}{\lambda_0 + N} + \lambda_0 \mu_0^2 + \sum_n X_n^2$$

$$E_{\tau} = \frac{(N+1)/\lambda}{\frac{1}{2} \left( \frac{(\sum X_n)^2}{N} + \frac{1}{E_{\tau}} - \frac{2}{N} + \sum_n X_n^2 \right)}$$

$$\frac{1}{E_{\tau}} = \frac{\lambda}{N+1} \cdot \frac{1}{E_{\tau}} + \frac{1}{N+1} \left( \frac{(\sum X_n)^2}{N} - \frac{2}{N} + N \sum_n X_n^2 \right)$$

$$\frac{1}{E_{\tau}} \frac{N}{N+1} = \frac{1}{N(N+1)} \left( \dots \right)$$

$$\frac{1}{E_{\tau}} = \frac{\lambda}{N} \left( \frac{1}{N} (\sum X_n)^2 + \sum_n X_n^2 - \frac{1}{N} (\sum X_n)^2 \right)$$

$\sigma^2$

$$\sigma = \sqrt{\frac{1}{N} \sum_n (X_n - \bar{X})^2}$$

$$\sum_n X_n^2 - 2 \sum X_n \cdot \frac{\sum X_n}{N} + \left( \frac{\sum X_n}{N} \right)^2 \cdot N$$

$$\sum_n X_n^2 - \frac{2}{N} (\sum X_n)^2 + \frac{1}{N} (\sum X_n)^2$$

$$= \sum_n X_n^2 - \frac{1}{N} (\sum X_n)^2$$

$$\bar{x}_1, \dots, \bar{x}_n$$

$$\bar{x}_n \sim \left( \bar{z}_n = \begin{bmatrix} 0 \\ 1 \end{bmatrix} \right)$$

$$p(\bar{x}) = \sum_{k=1}^K \pi_k \mathcal{N}(\bar{x} | \bar{\mu}_k, \Lambda_k^{-1})$$

$$p(\bar{x}_n | \bar{z}_n, \bar{\mu}, \Lambda) = \prod_k \mathcal{N}(\bar{x}_n | \bar{\mu}_k, \Lambda_k^{-1})^{z_{nk}}$$

$$\theta = (\bar{\pi}, \bar{\mu}, \Lambda, \Lambda_k)$$

$$p(X, z | \theta) = p(z | \theta) p(X | z, \theta) = \left( \prod_{n=1}^n \prod_{k=1}^K \pi_k^{z_{nk}} \right) \left( \prod_n \prod_k \mathcal{N}(\bar{x}_n | \bar{\mu}_k, \Lambda_k^{-1})^{z_{nk}} \right)$$

$$p(X, z, \bar{\pi}, \bar{\mu}, \Lambda) =$$

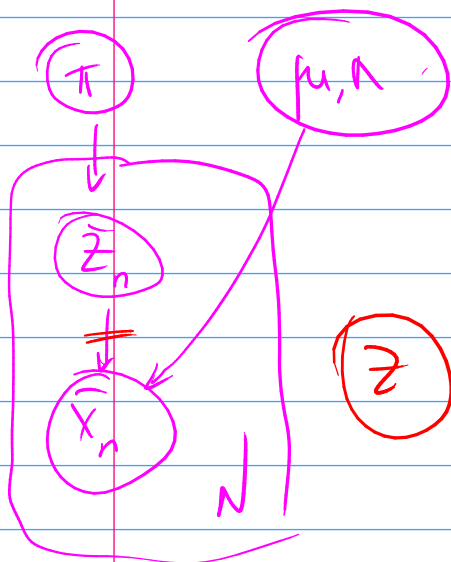
$$= p(z | \bar{\pi}) p(X | z, \bar{\mu}, \Lambda) p(\bar{\pi}, \bar{\mu}, \Lambda)$$

Wishart

$$p(\bar{\pi}) = \text{Dir}(\bar{\pi} | \bar{\alpha}_0) \propto \prod_k \pi_k^{\alpha_0 - 1}$$

$$p(\bar{\mu}, \Lambda) = p(\bar{\mu} | \Lambda) p(\Lambda) = \prod_{k=1}^K \mathcal{N}(\bar{\mu}_k | \bar{m}_0, (\beta_0 \Lambda_k)^{-1}) \cdot$$

$$\cdot W(\Lambda_k | W_0, \nu_0)$$



$$p(z, \pi, \mu, \Lambda | X) \approx q(z, \pi, \mu, \Lambda) = q(z) q(\pi, \mu, \Lambda)$$

$$\ln q^*(z) = E_{\pi, \mu, \Lambda} [\ln p(X, z, \pi, \mu, \Lambda)] + \text{const}$$

$$\ln q^*(\pi, \mu, \Lambda) = E_z [\ln p(X, z, \pi, \mu, \Lambda)] + \text{const}$$

$$\ln p(x, z, \pi, \mu, \Lambda) = \ln p(\pi) + \ln p(\mu, \Lambda) + \ln p(z|\pi) + \ln p(x|z, \mu, \Lambda)$$

$$\ln q^*(z) = E_{\pi, \mu, \Lambda} \left[ \ln p(z|\pi) + \ln p(x|z, \mu, \Lambda) \right] + \text{const}$$

$$= E_{\pi} \left[ \sum_n \sum_k z_{nk} \ln \pi_k \right] + E_{\pi, \mu, \Lambda} \left[ \sum_n \sum_k z_{nk} \left( \frac{1}{2} \ln \det \Lambda_k - \frac{D}{2} \ln 2\pi - \frac{1}{2} (\bar{x}_n - \bar{\mu}_k)^T \Lambda_k (\bar{x}_n - \bar{\mu}_k) \right) \right]$$

$$= \sum_n \sum_k z_{nk} E_{\pi} [\ln \pi_k] + \sum_n \sum_k z_{nk} E_{\pi, \mu, \Lambda} [\dots] + \text{const}$$

$$= \sum_n \sum_k z_{nk} \left( \dots \right) + \text{const}$$

(1)  $q^*(z) \propto \prod_n \prod_k p_{nk}^{z_{nk}}$ , zge  $E z_{nk} = \gamma_{nk}$  zge

$$\Gamma_{nk} = \frac{\gamma_{nk}}{\sum_k' \gamma_{nk}}$$

$$\ln \gamma_{nk} = E [\ln \pi_k] + \frac{1}{2} E [\ln \det \Lambda] - \frac{D}{2} \ln 2\pi - \frac{1}{2} E [(-)^T \Lambda_k (-)]$$

$$\ln q^*(\pi, \mu, \Lambda) = E_z [\dots] + \text{const}$$

$$= \ln p(\pi) + \ln p(\mu, \Lambda) + \sum_n \sum_k (\ln \pi_k) E [z_{nk}] +$$

$$\sum_k (\ln \det \Lambda_k) E [z_{nk}] + \sum_n \sum_k [E z_{nk}] \cdot \ln p(x_n | \mu_k, \Lambda_k) =$$

$$= \left( \ln p(\pi) + \sum_{n,k} (E z_{nk}) \ln \pi_k \right) + \sum_k \left( \ln p(\mu_k, \Lambda_k) + \sum_n [E z_{nk}] \ln p(x_n | \mu_k, \Lambda_k) \right) + \text{const}$$

(2)  $q^*(\pi, \mu, \Lambda) = q^*(\pi) \cdot \prod_{k=1}^K q^*(\mu_k, \Lambda_k)$

$$\ln q^*(\pi) = \sum_k (\alpha_{0k} + \sum_n E z_{nk} - 1) \ln \pi_k + \text{const}$$

$$q^*(\pi) \propto \prod_k \pi_k^{\alpha_{0k} + \sum_n E z_{nk} - 1}$$

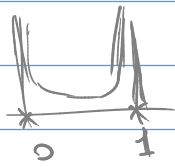
$$q^*(\pi) = \text{Dir}(\underline{\pi} | \underline{\alpha}), \quad \alpha_k = \alpha_{0k} + \sum_n E z_{nk}$$

1.  $q^{(0)} = p(\pi, \mu, \lambda)$

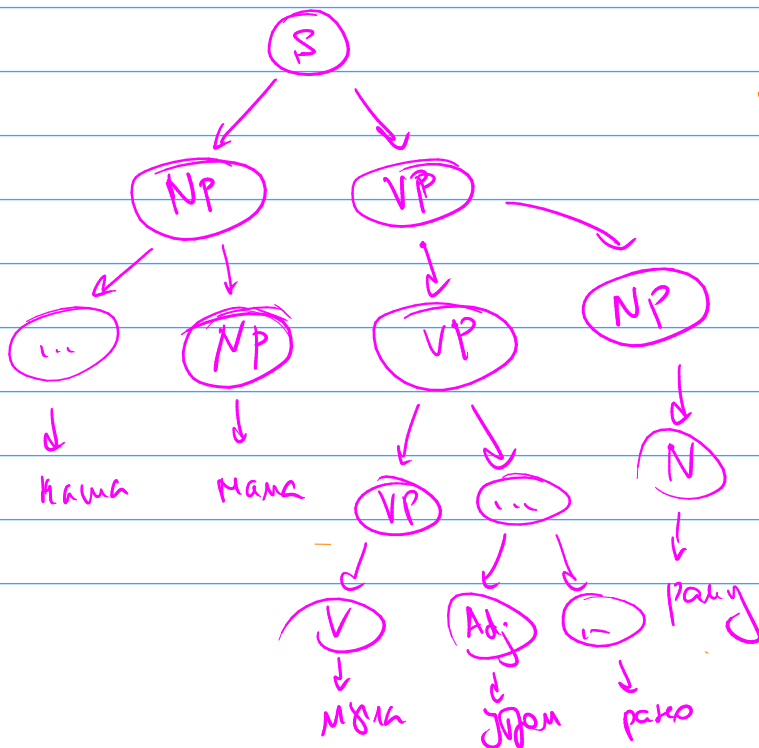
2. Итерации:

(E-шаг)  $z_{nk}^{(i)} \leftarrow q^{(i-1)}(\pi, \mu, \lambda)$  по  $y_k^{(i)}$

(M-шаг)  $q^{(i)}(\pi, \mu, \lambda) \leftarrow z_{nk}^{(i)}$  по  $y_k^{(i)}$

$p(\pi) = \text{Dir}(\underline{\pi} | \underline{\alpha}_0)$    $\underline{\alpha}_0 = \left( \frac{1}{100}, \dots, \frac{1}{100} \right)$

Чомский кама мана мана яма япон пако



- S → NP VP
- NP → Det NP
- NP → NP Det
- VP → VP NP
- NP → N
- VP → V
- VP → VP Adj

Pron	N	V	N	Adj	Adj
↑	↑	↑	↑	↑	↑
Мама	мама	мама	мама	япон	япон
		↓	↓		
мама	мама	мама	мама	япон	япон

Венский порт А. С. Пушкин (р. 1799, с. Мухоморова)

↓ ↓ ↓ ↓ ↓ ↓ ↓ ↓ ↓ ↓

0 0 B-Pers Pers Pers 0 B-Date B-Loc Loc

$x_1, x_2, x_3, \dots, x_{t-1}, x_t, x_{t-1}, \dots$

$$p(x_t | x_1 \dots x_{t-1}) \approx p(x_t | x_{t-1})$$

$(x_1, x_2, \dots, \underline{x_t}, x_{t+1})$

n-grams

$$p(x_t | x_{t-1}, x_{t-2}, \dots, x_{t-n})$$

(Obj, Pred, Subj)

- knowledge bases  
Freebase BB

(X, is, poet)

(X, born-in, 1799)