

$$\sigma(s_i + q_j) = p(i \text{ has } j)$$

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

$$t: s_1 s_2 \dots s_k$$

$$z_{ij} = [i \text{ has } j] \quad s_i \in t_t$$

$$s_i + q_j \sim \sigma^{-1}(p(\cdot))$$

$$t_{ij} = 0 \Rightarrow s_{ij} = 0 \quad \forall i \in t_t$$

$$t_{ij} = 1 \Rightarrow \exists i: s_{ij} = 1$$

$$\mathbb{E} z_{ij} = p(i \text{ has } j | s_i, q_j) =$$

E-war

$$= p(i \text{ has } j | \text{know } t \text{ has } j) =$$

$$= \frac{\sigma(s_i + q_j)}{1 - \prod_{i' \in t_t} (1 - \sigma(s_{i'} + q_j))}$$

$$H[q] = \int q \ln q \, dz$$

$$H[q + \delta q] = \dots$$

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

$$p(x|\theta) = \int p(x, z|\theta) \, dz$$

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

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

$$\ln p(x|\theta) = \mathbb{E}_{q(z)} [\ln p(x, z|\theta) - \ln p(z|x, \theta)] =$$

$$= \int q(z) \ln p(x, z|\theta) \, dz - \int \ln p(z|x, \theta) q(z) \, dz$$

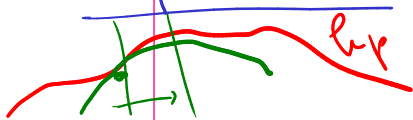
$$= \int \ln q(z) q(z) \, dz + \int \ln q(z) q(z) \, dz$$

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

$$\ln p(x|\theta) = \mathcal{L}(q, \theta) +$$

$$KL(q(z) \parallel p(z|x, \theta))$$

$$\Rightarrow \ln q, \theta^{(m)} = \ln p(x|\theta^{(m)})$$

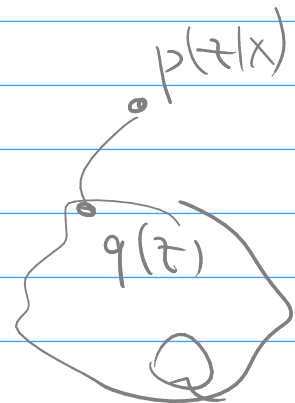


$$p(x, z) \quad p(z|x) = ? \quad p(x)$$

$$\ln p(x) = \mathbb{E}_{q(z)} [\ln p(x, z) - \ln p(z|x)]$$

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

$$\ln p(x) = \underbrace{\ln(q)}_{\substack{\uparrow q(z) \\ \text{max}}} + \underbrace{KL(q || p(z|x))}_{\substack{\downarrow q(z) \text{ min}}} \quad \Leftrightarrow$$



$$q(z) = \prod_{i=1}^M q_i(z_i), \quad z_i \cap z_j = \emptyset$$

$$Z = \{z_1, z_2, \dots, z_M\}, \quad z_i \subset Z, \quad q(z_1, \dots, z_M) = q(z_1, z_2, \dots, z_M)$$

$$\ln(q) = \int \ln \frac{p(x, z)}{q(z)} q(z) dz =$$

$$= \int \ln p(x, z) q(z) dz - \int \ln q(z) q(z) dz =$$

$$= \int \ln p(x, z) \cdot \prod_i q_i(z_i) dz - \sum_j \int \ln q_j(z_j) \cdot \prod_i q_i(z_i) dz$$

$$\ln(q_j) = \int \ln p(x, z) \prod_i q_i(z_i) dz - \int \ln q_j(z_j) q_j(z_j) dz_j + \text{const}$$

$$\int \ln p(x, z) \prod_i q_i(z_i) dz = \int \left( \int \ln p(x, z) \prod_{i \neq j} q_i(z_i) dz_{-j} \right) q_j(z_j) dz_j$$

$$\ln \tilde{p}(x, z_j)$$

$$\ln \tilde{p}(x, z_j) = \mathbb{E}_{i \neq j} \ln q_i(z_i) [\ln p(x, z)] + \text{const}$$

$$L(q_j) = \int \ln \tilde{p}(x, z_j) q_j(z_j) dz_j - \int \ln q_j q_j dz_j =$$

$$= \int \ln \frac{\tilde{p}(x, z_j)}{q_j} q_j dz_j = -\mathbb{E} L(q_j \| \tilde{p}) + \text{const}$$

$$L(q_j) \xrightarrow{q_j} \max \Leftrightarrow \underline{q_j^*(z_j) = \tilde{p}(x, z_j)}$$

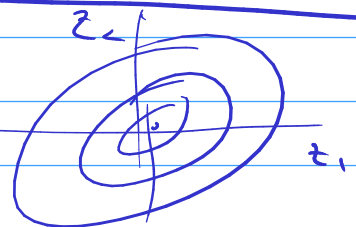
$$\text{Ecm } q(z) = \prod_{i=1}^M q_i(z_i) \text{ due } z_i \wedge z_j = \emptyset, \text{ to}$$

$$\forall j \quad \boxed{\ln q_j^*(z_j) = \mathbb{E}_{i \neq j} [\ln p(x, z)] + \text{const}}$$

$$\text{we } q^* = \prod_i q_i^* \text{ - so onl. approx. } p(z|x)$$

$$(1) \quad p(\bar{z}) = \mathcal{N}(\bar{z} | \bar{\mu}, \Lambda^{-1}) \quad \bar{z} = \begin{pmatrix} z_1 \\ z_2 \end{pmatrix}$$

$$\begin{pmatrix} \mu_1 \\ \mu_2 \end{pmatrix} \quad \begin{pmatrix} \Lambda_{11} & \Lambda_{12} \\ \Lambda_{12} & \Lambda_{22} \end{pmatrix}$$



$$p(\bar{z}) \approx q(\bar{z}) = q_1(z_1) q_2(z_2)$$

$$\begin{aligned} \ln q_1^*(z_1) &= \mathbb{E}_{q_2^*} [\ln p(z_1, z_2)] + \text{const} \\ \ln q_2^*(z_2) &= \mathbb{E}_{q_1^*} [\ln p(z_1, z_2)] + \text{const} \end{aligned}$$

$$\begin{aligned} \ln p(z_1, z_2) &= -\ln 2\pi - \frac{1}{2} \ln \det \Lambda - \frac{1}{2} (\bar{z} - \bar{\mu})^T \Lambda (\bar{z} - \bar{\mu}) = \\ &= \text{const} - \frac{1}{2} \left( (z_1 - \mu_1)^2 \Lambda_{11} + 2(z_1 - \mu_1)(z_2 - \mu_2) \Lambda_{12} + (z_2 - \mu_2)^2 \Lambda_{22} \right) \end{aligned}$$

$$\ln q_1^*(z_1) = \mathbb{E}_{q_2^*(z_2)} \left[ -\frac{1}{2} (z_1 - \mu_1)^2 \Lambda_{11} - z_1 \Lambda_{12} (z_2 - \mu_2) \right] + \text{const}$$

$$\ln q_1^*(z_1) = -\frac{1}{2} \Lambda_{11} z_1^2 + z_1 \mu_1 \Lambda_{11} - z_1 \Lambda_{12} (\mathbb{E}_{q_2^*} [z_2] - \mu_2) + \text{const}$$

$$\ln q_1^*(z_1) = -\frac{1}{2} \underline{\Lambda_{11}} z_1^2 + (\mu_1 \Lambda_{11} - \Lambda_{12} (\mathbb{E}[z_2] - \mu_2)) \underline{z_1} + \text{const}$$

$$q_1^*(z_1) = \mathcal{N}(z_1 | m_1, \alpha_1^{-1})$$

$$\underline{\alpha_1 = \Lambda_{11}}$$

$$m_1 = \frac{1}{\Lambda_{11}} (\mu_1 \Lambda_{11} - \Lambda_{12} (\dots))$$

$$m_1 = \mu_1 - \frac{\Lambda_{12}}{\Lambda_{11}} (\mathbb{E}_{q_2^*}[z_2] - \mu_2)$$

$$\ln q_2^*(z_2) = \mathbb{E}_{q_1^*(z_1)} \left[ -\frac{1}{2} (z_2 - \mu_2)^2 \Lambda_{22} - z_2 \Lambda_{12} (z_1 - \mu_1) \right] + \text{const}$$

$$\ln q_2^*(z_2) = -\frac{1}{2} \Lambda_{22} z_2^2 + (\mu_2 \Lambda_{22} - \Lambda_{12} (\mathbb{E}[z_1] - \mu_1))$$

$$q_2^*(z_2) = \mathcal{N}(z_2 | m_2, \alpha_2^{-1}), \quad \alpha_2 = \Lambda_{22}$$

$$m_2 = \mu_2 - \frac{\Lambda_{12}}{\Lambda_{22}} \left( \underbrace{\mathbb{E}_{q_1^*}[z_1]}_{m_1} - \mu_1 \right)$$

$$\left\{ \begin{aligned} m_1 &= \mu_1 - \frac{\Lambda_{12}}{\Lambda_{11}} (m_2 - \mu_2) \\ m_2 &= \mu_2 - \frac{\Lambda_{12}}{\Lambda_{22}} (m_1 - \mu_1) \end{aligned} \right.$$

$$m_1 = \mu_1$$

$$m_2 = \mu_2$$

$$KL(q||p) \rightarrow \min$$

$$p(\bar{z}) = \mathcal{N}(\bar{z} | \bar{\mu}, \begin{pmatrix} \Lambda_{11} & \Lambda_{12} \\ \Lambda_{22} & \Lambda_{22} \end{pmatrix}^{-1}) \approx q_1(z_1) q_2(z_2)$$

$$\left\{ \begin{aligned} q_1(z_1) &= \mathcal{N}(z_1 | \mu_1, \Lambda_{11}^{-1}) \\ q_2(z_2) &= \mathcal{N}(z_2 | \mu_2, \Lambda_{22}^{-1}) \end{aligned} \right.$$

$$KL(p||q) \rightarrow \min$$

$$\int p(z) \ln \frac{p(z)}{q(z)} dz = \int p \ln p dz - \underbrace{\int p(z) \ln q(z) dz}_{\rightarrow \max}$$

$q \rightarrow \min$

$$q = \prod q_i$$

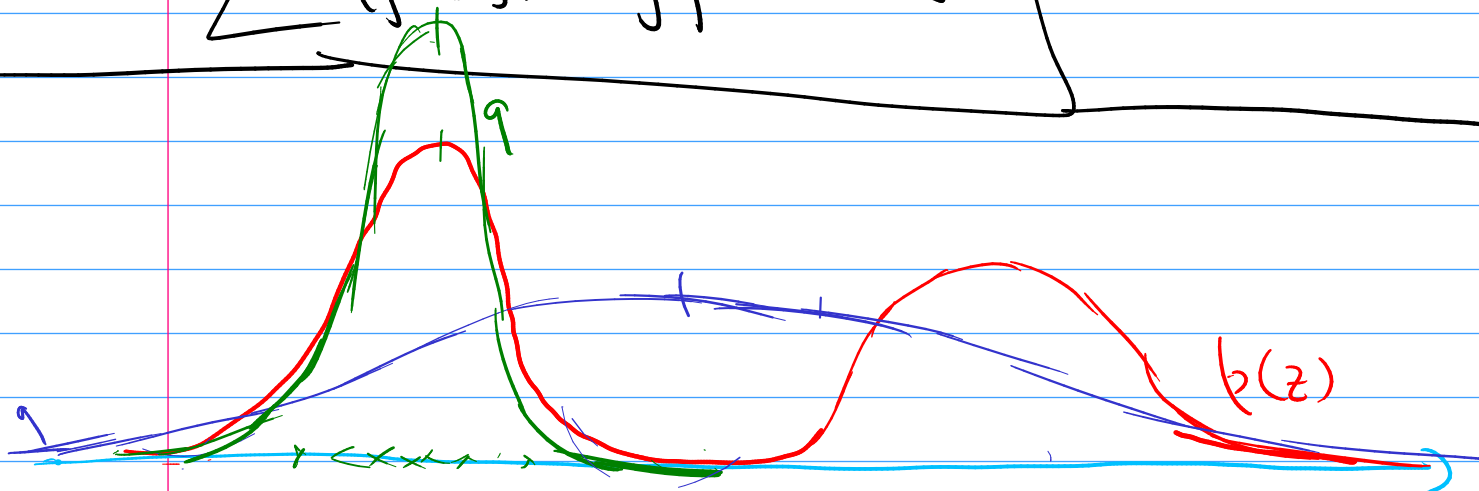
$$\sum_i \int p(z) \ln q_i(z_i) dz$$

$$j: \int \underline{p(z)} \ln q_j(z_j) dz \rightarrow \max$$

$$\int \underline{\tilde{p}(z_j)} \ln q_j(z_j) dz_j$$

"  $\int p(z) dz_{-j}$

$$q_j^*(z_j) = \int p(z) dz_{-j} \quad KL(p||q)$$



$$KL(p||q) = \int \ln \frac{p}{q} p dz \rightarrow \min$$

$$KL(q||p) = \int \ln \frac{q}{p} q dz \rightarrow \min$$

$$(2) D = \{x_1, x_2, \dots, x_N\}$$

$$p(D|\mu, \tau) = \prod_{n=1}^N \sqrt{\frac{\tau}{2\pi}} e^{-\frac{\tau}{2} (x_n - \mu)^2}$$

$$p(\tau) = \text{Gam}(\tau | a_0, b_0) \propto \tau^{a_0-1} e^{-b_0 \tau}$$

$$p(\mu|\tau) = \mathcal{N}(\mu | \mu_0, (\lambda_0 \tau)^{-1})$$

$$p(\underline{\mu, \tau} | D) \propto \overbrace{p(\tau) p(\mu|\tau)}^{\text{prior}} \underbrace{p(D|\mu, \tau)}_{\text{likelihood}}$$

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

$$p(Z, X) = p(\mu, \tau, D) = \underline{p(\tau)} \underline{p(\mu | \tau)} \underline{p(D | \mu, \tau)}$$

$$\ln p(\mu, \tau, D) = \text{const} + (a_0 - 1) \underline{\ln \tau} - \underline{b_0 \tau} + \frac{1}{2} \underline{\ln \tau} - \frac{\lambda_0 \tau}{2} (\mu - \mu_0)^2 + \frac{N}{2} \underline{\ln \tau} - \frac{\tau}{2} \sum_{n=1}^N (x_n - \mu)^2$$

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

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

$$q_\tau^*(\tau) = \text{Gam}(\tau | a, b)$$

$$a = a_0 + \frac{N+1}{2}; \quad b = b_0 + \frac{1}{2} \mathbb{E}_\mu \left[ \lambda_0 (\mu - \mu_0)^2 + \sum_n (x_n - \mu)^2 \right]$$

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

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

$$= -\frac{\lambda_0 \mathbb{E} \tau}{2} (\mu - \mu_0)^2 - \frac{\mathbb{E} \tau}{2} \sum_n (x_n - \mu)^2 + \text{const}$$

$$= -\frac{\mathbb{E} \tau}{2} \left( b \mu^2 - 2 \lambda_0 \mu_0 \cdot \mu + N \cdot \mu^2 - 2 \mu \cdot \sum_n x_n \right) + \text{const}$$

$$= -\frac{\mathbb{E} \tau (\lambda_0 + N)}{2} \left( \mu - \frac{\lambda_0 \mu_0 + \sum_n x_n}{\lambda_0 + N} \right)^2 + \text{const}$$

$$q_\mu^*(\mu) = N(\mu | m, \alpha^{-1})$$

$$\alpha = (\lambda_0 + N) \cdot \mathbb{E} \tau$$

$$m = \frac{\lambda_0 \mu_0 + \sum_n x_n}{\lambda_0 + N}$$

$$= \frac{1}{N} \sum_n x_n$$

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

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

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

$$E_{\mu}[\mu] = \frac{\lambda_0 \mu_0 + \sum_n x_n}{\lambda_0 + N} ; \quad E_{\mu}[\mu^2] = [E_{\mu}]^2 + \text{Var}[\mu] =$$

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

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

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

$$= \frac{1}{E_{\tau}} - \frac{(\sum x_n)^2}{N} + \sum x_n^2$$

$$E_{\tau} = \frac{a}{b} = \frac{a_0 + \frac{N+1}{2}}{b_0 + \frac{1}{2} \left( \frac{1}{E_{\tau}} - \frac{(\sum x_n)^2}{N} + \sum x_n^2 \right)} = \frac{N+1}{\frac{1}{E_{\tau}} - \frac{(\sum x_n)^2}{N} + \sum x_n^2}$$

$$\left( \frac{1}{E_{\tau}} \right) = \frac{1}{a_0 + \frac{N+1}{2}} \cdot \left( b_0 + \frac{1}{2} \cdot \left( \frac{1}{E_{\tau}} \right) - \frac{1}{2} \frac{(\sum x_n)^2}{N} + \frac{1}{2} \left( \sum x_n^2 \right) \right)$$

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

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

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

$$\frac{1}{E\tau} = \frac{1}{N} \cdot \left( \sum x_n^2 - \frac{1}{N} (\sum x_n)^2 \right)$$

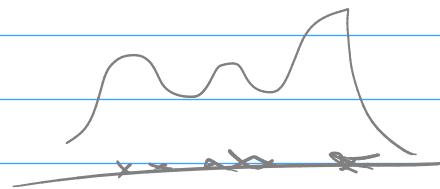
$$\bar{x} = \frac{1}{N} \cdot \sum x_n$$

$$\begin{aligned} \sigma^2 &= \frac{1}{N} \cdot \sum_n (x_n - \bar{x})^2 = \frac{1}{N} \sum_n (x_n^2 - 2x_n \bar{x} + \bar{x}^2) = \\ &= \frac{1}{N} \left( \sum_n x_n^2 + N \bar{x}^2 - 2 \bar{x} \cdot \sum x_n \right) = \\ &= \frac{1}{N} \left( \sum_n x_n^2 + \frac{1}{N} (\sum x_n)^2 - 2 \cdot \frac{1}{N} \cdot (\sum x_n)^2 \right) \end{aligned}$$

③ Смесь гауссианов  $X = \{x_1 \dots x_N\}$   $\pi_k, \sum \pi_k = 1$

$k$ -го компонента  $N(\bar{x} | \mu_k, \lambda_k^{-1})$ ,  $k = 1 \dots K$

$$p(x_n | \dots) = \sum_{k=1}^K \pi_k N(x_n | \mu_k, \lambda_k^{-1})$$



$$p(\bar{\mu}, \bar{\lambda}, \bar{\pi}) = p(\bar{\pi}) \prod_{k=1}^K p(\mu_k, \lambda_k)$$

$$p(\bar{\pi}) = \text{Dir}(\bar{\pi} | \alpha_0) \propto \pi_1^{\alpha_0-1} \pi_2^{\alpha_0-1} \dots \pi_K^{\alpha_0-1}$$

$$p(\mu_k, \lambda_k) = \text{Gam}(\lambda_k | a_0, b_0) N(\mu_k | \mu_0, (b_0 \lambda_k)^{-1})$$

$$z_{nk} = [x_n \text{ порожд. } N(\mu_k, \lambda_k^{-1})]$$

$$\ln p(X, Z, \bar{\pi}, \bar{\mu}, \bar{\lambda}) = \ln \left[ p(\bar{\pi}) \prod_{k=1}^K p(\mu_k, \lambda_k) \prod_{n,k} (\pi_{nk} N(x_n | \mu_k, \lambda_k^{-1}))^{z_{nk}} \right]$$

$$= \ln p(\bar{\pi}) + \sum_k \left[ \ln p(\lambda_k) + \ln p(\mu_k | \lambda_k) \right] +$$

$$+ \sum_n \sum_k z_{nk} (\ln \pi_{nk} + \ln N(x_n | \mu_k, \lambda_k^{-1}))$$



$p(x|X)$   $p(z, \bar{\pi}, \bar{\mu}, \bar{\lambda} | X) \approx q(z, \bar{\pi}, \bar{\mu}, \bar{\lambda})$  " эмпирич. предполож.

$q(z) q(\bar{\pi}, \bar{\mu}, \bar{\lambda})$

$$\begin{aligned} \ln q^*(z) &= \mathbb{E}_{\pi, \mu, \lambda} [\ln p(x, z, \bar{\pi}, \bar{\mu}, \bar{\lambda})] + \text{const} = \\ &= \mathbb{E}_{\pi, \mu, \lambda} [\ln p(z | \pi) + \ln p(x | z, \mu, \lambda)] + \text{const} = \\ &= \mathbb{E}_{\pi, \mu, \lambda} \left[ \sum_{n,k} z_{nk} \left( \ln \pi_k + \ln \mathcal{N}(x_n | \mu_k, \lambda_k^{-1}) \right) \right] + \text{const} = \\ &\quad - \frac{1}{2} \ln 2\pi + \frac{1}{2} \ln \lambda_k - \frac{\lambda_k}{2} (x_n - \mu_k)^2 \end{aligned}$$

$$= \sum_{n,k} z_{nk} \left( \mathbb{E}[\ln \pi_k] - \frac{1}{2} \ln 2\pi + \frac{1}{2} \mathbb{E}[\ln \lambda_k] - \frac{1}{2} \mathbb{E}[\lambda_k (x_n - \mu_k)^2] \right) + \text{const}$$

$$q^*(z) \propto \prod_n \prod_k p_{nk}^{z_{nk}}, \text{ где}$$

$$\ln p_{nk} = \mathbb{E}[\ln \pi_k] -$$

$$\mathbb{E}[z_{nk}] = p(z_{nk}=1) = \frac{p_{nk}}{\sum_l p_{nl}} = z_{nk},$$

$$\text{где } \ln z_{nk} = \text{---} + \text{const}$$

$$\ln q^*(\bar{\pi}, \bar{\mu}, \bar{\lambda}) = \mathbb{E}_z [\ln p(x, z, \bar{\pi}, \bar{\mu}, \bar{\lambda})] + \text{const} =$$

$$\begin{aligned}
&= \ln p(\pi) + \sum_k \left( \ln p(\lambda_k) + \ln p(\mu_k | \lambda_k) \right) + \\
&\quad + \mathbb{E}_z \left[ \ln p(z | \pi) \right] + \mathbb{E}_z \left[ \ln p(x | z, \pi, \mu, \lambda) \right] = \\
&= (\alpha_0 - 1) \sum_k \ln \pi_k + \sum_k \left( (\alpha_0 - 1) \ln \lambda_k - b_0 \lambda_k + \frac{1}{2} \ln \lambda_k - \frac{\lambda_k \beta_0}{2} (\mu_k - \mu_0)^2 \right) \\
&\quad + \sum_{n,k} \mathbb{E}_z[z_{nk}] \ln \pi_k + \sum_{n,k} \mathbb{E}[z_{nk}] (\dots)
\end{aligned}$$

$$q^*(\pi, \mu, \lambda) = q^*(\pi) \prod_k q^*(\lambda_k, \mu_k)$$

$$\ln q^*(\pi) = \sum_k \ln \pi_k \cdot (\alpha_0 - 1 + \sum_n z_{nk})$$

$$q^*(\pi) = \text{Dir}(\pi | \bar{\alpha}), \quad \alpha_k = \alpha_0 + \sum_n z_{nk}$$

$$\begin{aligned}
\ln q^*(\lambda_k, \mu_k) &= (\alpha_0 - 1) \ln \lambda_k - b_0 \lambda_k + \frac{1}{2} \ln \lambda_k - \frac{\lambda_k \beta_0}{2} (\mu_k - \mu_0)^2 \\
&\quad + \sum_n z_{nk} \left( \frac{1}{2} \ln \lambda_k - \frac{\lambda_k}{2} (x_n - \mu_k)^2 \right)
\end{aligned}$$

$$- \frac{\lambda_k}{2} \left( \beta_0 + \sum_n z_{nk} \right) \mu_k^2$$

$$\lambda_k \left( \beta_0 \mu_0 + \sum_n z_{nk} x_n \right) \mu_k$$

$$q^*(\lambda_k, \mu_k) = \text{Gam}(\lambda_k | a, b) \cdot \mathcal{N}(\mu_k | m, (\lambda_k \beta)^{-1})$$

$$\beta = \beta_0 + \sum_n z_{nk}$$

$$a = \alpha_0 + \frac{1}{2} \sum_n z_{nk}$$

$$m = \frac{\beta_0 \mu_0 + \sum_n z_{nk} x_n}{\beta_0 + \sum_n z_{nk}}$$

$$b = b_0 + \dots$$

$$p(d|...) = \prod p(w|...)$$

