

Problem 2

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$$p(\gamma, z, w | x) = \prod_{d=1}^D q(w_d) \prod_{n=1}^N q(z_n) \prod_{d=1}^D q(\gamma_d)$$

$$p(\gamma, z, w, x) = p(x | \gamma, z, w) \cdot \underbrace{p(\gamma, z, w)}_{p(x) \cdot p(\gamma) \cdot p(z) \cdot p(w)} = p(x) \cdot p(\gamma) \cdot p(z) \cdot p(w)$$

$$= p(x | \gamma, z, w) \cdot p(\gamma) \cdot p(z) \cdot p(w).$$

$$\Rightarrow \log(p(\gamma, z, w, x)) = \log(p(x | \gamma, z, w)) + \log(p(\gamma)) + \log(p(z)) + \log(p(w))$$

$$\log(p(x | \gamma, z, w)) = \sum_{n=1}^N \log N_D(x_n | w_{z_n}, \text{diag}(\gamma))$$

$$\log N_D(x_n | w_{z_n}, \text{diag}(\gamma)) = -\frac{D}{2} \log 2\pi + \frac{1}{2} \log(\det(\text{diag}(\gamma))) - \frac{1}{2} (x_n - w_{z_n})^T \text{diag}(\gamma) (x_n - w_{z_n})$$

$$\Rightarrow \log p(x | \gamma, z, w) = -\frac{ND}{2} \log 2\pi + \frac{N}{2} \log(\det(\text{diag}(\gamma))) + -\frac{1}{2} \sum_{n=1}^N (x_n - w_{z_n})^T \text{diag}(\gamma) (x_n - w_{z_n})$$

$$\log p(\gamma) = \sum_{d=1}^D \log \text{Gamma}(\gamma_d | a, b)$$

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$$= \sum_{d=1}^D a \log b - \log \Gamma(a) + (a-1) \log \gamma_d - b \gamma_d$$

$$= D(a \log b - \log \Gamma(a)) + \sum_{d=1}^D (a-1) \log \gamma_d - b \gamma_d$$

$$\log p(z) = \sum_{n=1}^N \log N_k(z_n | 0, I) = \underbrace{\frac{N}{2} \cdot \log(1)}_{=0}$$

$$= -\frac{NK}{2} \log 2\pi + \frac{N}{2} \log (\det(I))$$

$$- \frac{1}{2} \sum_{n=1}^N (z_n - 0)^T \cdot I (z_n - 0)$$

$$= -\frac{NK}{2} \log 2\pi - \frac{1}{2} \sum_{n=1}^N z_n^T \cdot I \cdot z_n$$

$$\log p(w) = \sum_{d=1}^D \log N_k(w_d | 0, \alpha I) \quad I \in \mathbb{R}^{K \times K}$$

$$= -\frac{DK}{2} \log (2\pi) + \frac{D}{2} \log (\underbrace{\det(\alpha I)}_{= \alpha^K})$$

$$- \frac{1}{2} \sum_{d=1}^D w_d^T (\alpha I)^{-1} w_d$$

$$= -\frac{DK}{2} \log 2\pi + \frac{DK}{2} \log \alpha$$

$$-\frac{D}{2\alpha} \sum_{d=1}^D w_d^T w_d$$

$$\Rightarrow \log p(\gamma, z, w, x)$$

$$\begin{aligned}
 &= -\frac{ND}{2} \log 2\pi + \frac{N}{2} \log \det(\text{diag}(\gamma)) - \frac{1}{2} \sum_{n=1}^N (x_n - Wz_n)^T \text{diag}(\gamma) \\
 &\quad + D(a \log b - \log I(a) + \sum_{d=1}^D (a-1) \log \gamma_d - b \gamma_d) \\
 &- \frac{NK}{2} \log 2\pi - \frac{1}{2} \sum_{n=1}^N z_n^T z_n \\
 &- \frac{NK}{2} \log 2\pi + \frac{NK}{2} \log a - \frac{1}{2a} \sum_{d=1}^D w_d^T w_d
 \end{aligned}$$

part 2

only those terms which depend on  $z$  denoted by:

$$\begin{aligned}
 \log p_z(\gamma | \log w(z)) &= -\frac{1}{2} \sum_{n=1}^N (x_n - Wz_n)^T \text{diag}(\gamma) (x_n - Wz_n) \\
 &\quad - \frac{1}{2} \sum_{n=1}^N z_n^T z_n
 \end{aligned}$$

$$= -\frac{1}{2} \sum_{n=1}^N (x_n - Wz_n)^T \text{diag}(\gamma) (x_n - Wz_n) - \frac{1}{2} z_n^T z_n$$

lets look at the summands:

$$-\frac{1}{2} (x_n - Wz_n)^T \text{diag}(\gamma) (x_n - Wz_n) - \frac{1}{2} z_n^T z_n$$

$$\underbrace{-\frac{1}{2} x_n^T \text{diag}(\gamma) x_n}_{\text{discard since does not depend on } z} + x_n^T \text{diag}(\gamma) Wz_n - \frac{1}{2} z_n^T W^T \text{diag}(\gamma) Wz_n - \frac{1}{2} z_n^T z_n$$

$$= \frac{1}{2} z_n^T \underbrace{\left( W^T \text{diag}(\gamma) W + I \right)}_A z_n + \underbrace{x_n^T \text{diag}(\gamma) W z_n}_{= b^T}$$

$$= -\frac{1}{2} z_n^T A z_n + b^T z_n$$

apply the completing square technique since  
indeed  $A^T = A$

$$\Rightarrow -\frac{1}{2} z_n^T A z_n + b^T z_n = -\frac{1}{2} (z_n - A^{-1}b)^T A (z_n - A^{-1}b)$$

$$-\underbrace{\frac{1}{2} b^T A^{-1}b}_{\text{does not depend on } z} \\ \Rightarrow \text{discard}$$

$$\Rightarrow \log p_z(z) = \sum_{n=1}^N -\frac{1}{2} (z_n - A^{-1}b)^T A (z_n - A^{-1}b)$$

where  $A = W^T \text{diag}(\gamma) W + I$ ;  $b = W^T \text{diag}(\gamma) x_n$

how lets look at

$$\log q(z_n) = -\frac{1}{2} (z_n - A^{-1}b)^T A (z_n - A^{-1}b)$$

we notice that this is the form of a normal distribution. The covariance matrix is  $A^{-1}$  and the mean is  $A^{-1}b$ .

$$\Rightarrow q(z_n) \sim N_k(A^{-1}b, A^{-1})$$

but since now we need to replace  $\gamma$  and  $W$  with their expectations since we assumed that they were given  
 (sorry for the shitty notation)

$$\Rightarrow A = \mathbb{E}_{w, \gamma} [W^T \text{diag}(\gamma) W + I], \quad b = \mathbb{E}_{w, \gamma} [W^T \text{diag}(\gamma) X_n]$$

$$A = \mathbb{E}_{w, \gamma} [[W^T \text{diag}(\gamma) W] + I]$$

using the hint

$$= \mathbb{E}_{w, \gamma} \left[ \sum_{d=1}^D \gamma_d w_d w_d^T \right] + I$$

linearity

$$= I + \sum_{d=1}^D \mathbb{E}_{w, \gamma} [\gamma_d w_d w_d^T] + I$$

independence assumption between  $\gamma$  and  $W$

$$= I + \sum_{d=1}^D \underbrace{\mathbb{E}_\gamma [\gamma_d]}_{\langle \gamma_d \rangle} \underbrace{\mathbb{E}_w [w_d w_d^T]}_{\langle w_d w_d^T \rangle} + I$$

$$= I + \sum_{d=1}^D \langle \gamma_d \rangle \cdot \langle w_d w_d^T \rangle$$

$$\Rightarrow A^{-1} = \left( I + \sum_{d=1}^D \langle \gamma_d \rangle \cdot \langle w_d w_d^T \rangle \right)^{-1}$$

covariance matrix

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how the mean similarly, independent "constant"

$$b = \mathbb{E}_{w, \gamma} [w^T \text{diag}(\gamma) x_n] = \mathbb{E}_{w, \gamma} [\underbrace{w^T \text{diag}(\gamma)}_{\downarrow \downarrow}] x_n$$

$$= \underbrace{\mathbb{E}_w [w^T]}_{\langle w^T \rangle} \cdot \underbrace{\mathbb{E}_\gamma [\text{diag}(\gamma)]}_{\text{diag}(\langle \gamma \rangle)} x_n$$

$$= \langle w^T \rangle \text{diag}(\langle \gamma \rangle) x_n \Rightarrow A^{-1} b = A^{-1} \langle w^T \rangle \text{diag}(\langle \gamma \rangle) x_n$$

via independence assumption

thus we have shown that the update factor is equal to:

$$q(z_n) = \mathcal{N}_K (\mu_n, K_n)$$

$$\text{where } K_n = A^{-1} = \left( I + \sum_{d=1}^D \langle \gamma_d \rangle \langle w_d w_d^T \rangle \right)^{-1}$$

$$\mu_n = A^{-1} b = K_n \langle w^T \rangle \text{diag}(\langle \gamma \rangle) x_n$$

□

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