

Bayesian Models for Machine Learning

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a) Denote $X = \{x_1, \dots, x_N\}$, $Y = \{y_1, \dots, y_N\}$, $\alpha = \{\alpha_1, \dots, \alpha_d\}$

$$P(Y, W, \alpha, \lambda | X) = P(Y | X, W, \lambda) P(W | \alpha) P(\lambda) P(\alpha)$$

$$\therefore \ln P(Y, W, \alpha, \lambda | X) = \ln P(Y | X, W, \lambda) + \ln P(W | \alpha) + \ln P(\lambda) + \ln P(\alpha)$$

$$= \sum_{i=1}^N \ln P(y_i | x_i, W, \lambda) + \ln P(W | \alpha) + \ln P(\lambda) + \sum_{i=1}^d \ln P(\alpha_i)$$

$$= \frac{N}{2} \ln \lambda - \frac{N}{2} \ln 2\pi - \frac{\lambda}{2} \sum_{i=1}^N (y_i^2 - 2W^T x_i y_i + W^T x_i x_i^T W) + \frac{1}{2} \sum_{i=1}^d \ln \alpha_i - \frac{1}{2} \ln 2\pi - \frac{1}{2} (W^T \text{diag}(\alpha_1, \dots, \alpha_d) W) \\ + e_0 \ln f_0 - \ln T(e_0) + (e_0 - 1) \ln \lambda - f_0 \lambda + \sum_{i=1}^d (a_0 \ln b_0 - \ln T(a_0) + (a_0 - 1) \ln \alpha_i - b_0 \alpha_i)$$

$$\cdot q(W) = \exp \left\{ E_{q(\alpha, \lambda)} [\ln P(Y, W, \alpha, \lambda | X)] \right\} \propto \exp \left\{ E_{q(\alpha, \lambda)} \left[-\frac{\lambda}{2} \sum_{i=1}^N (y_i^2 - 2W^T x_i y_i + W^T x_i x_i^T W) - \frac{1}{2} (W^T \text{diag}(\alpha_1, \dots, \alpha_d) W) \right] \right\}$$

$$= \exp \left\{ -\frac{\lambda}{2} \left[E_{q(\alpha)} \sum_{i=1}^N (y_i^2 - 2W^T x_i y_i + W^T x_i x_i^T W) + (W^T E_{q(\alpha)} [\text{diag}(\alpha_1, \dots, \alpha_d)] W) \right] \right\}$$

$$\propto \exp \left\{ -\frac{1}{2} [W^T E_{q(\alpha)} (\text{diag}(\alpha_1, \dots, \alpha_d)) + E_{q(\alpha)} \sum_{i=1}^N x_i x_i^T] W - 2W^T (E_{q(\alpha)} \sum_{i=1}^N x_i y_i) \right\}$$

$$\therefore q(W) = \text{Normal}(U', \Sigma') \text{ where } \Sigma' = (E_{q(\alpha)} [\text{diag}(\alpha_1, \dots, \alpha_d)] + E_{q(\alpha)} \sum_{i=1}^N x_i x_i^T)^{-1} \\ U' = \Sigma' (E_{q(\alpha)} \sum_{i=1}^N x_i y_i)$$

$$\cdot q(\lambda) = \exp \left\{ E_{q(W, \alpha)} [\ln P(Y, W, \alpha, \lambda | X)] \right\} \propto \exp \left\{ E_{q(W, \alpha)} \left[\frac{N}{2} \ln \lambda - \frac{\lambda}{2} \sum_{i=1}^N (y_i^2 - 2W^T x_i y_i + W^T x_i x_i^T W) + (e_0 - 1) \ln \lambda - f_0 \lambda \right] \right\}$$

$$= \exp \left\{ \frac{N}{2} \ln \lambda - \frac{\lambda}{2} \sum_{i=1}^N (y_i^2 - 2E_{q(W)}^T x_i y_i + E_{q(W)}^T x_i x_i^T E_{q(W)}) + (e_0 - 1) \ln \lambda - f_0 \lambda \right\}$$

$$= \exp \left\{ \ln \lambda \left(\frac{N}{2} + e_0 - 1 \right) - \lambda \left(\frac{1}{2} \sum_{i=1}^N (y_i^2 - 2E_{q(W)}^T x_i y_i + E_{q(W)}^T x_i x_i^T E_{q(W)}) + f_0 \right) \right\}$$

$$\therefore q(\lambda) = \text{Gamma}(e', f') \text{ where } e' = \frac{N}{2} + e_0 \quad f' = \frac{1}{2} \sum_{i=1}^N (y_i^2 - 2E_{q(W)}^T x_i y_i + E_{q(W)}^T x_i x_i^T E_{q(W)}) + f_0$$

$$\cdot q(\alpha) = \exp \left\{ E_{q(W, \lambda)} [\ln P(Y, W, \alpha, \lambda | X)] \right\} \propto \exp \left\{ E_{q(W, \lambda)} \left[\frac{1}{2} \sum_{i=1}^d \ln \alpha_i - \frac{1}{2} (W^T \text{diag}(\alpha_1, \dots, \alpha_d) W) + (a_0 - 1) \sum_{i=1}^d \ln \alpha_i - b_0 \sum_{i=1}^d \alpha_i \right] \right\}$$

$$\text{Since } q(\alpha) = \prod_{i=1}^d q(\alpha_i)$$

$$\cdot q(\alpha_i) \propto \exp \left\{ E_{q(W, \lambda)} \left[\frac{1}{2} \ln \alpha_i - \frac{1}{2} (W^T \text{diag}(\alpha_1, \dots, \alpha_d) W) + (a_0 - 1) \ln \alpha_i - b_0 \alpha_i \right] \right\}$$

$$= \exp \left\{ \frac{1}{2} \ln \alpha_i - \frac{1}{2} \text{Tr}(\text{diag}(\alpha_1, \dots, \alpha_d) E_{q(W)} W W^T) + (a_0 - 1) \ln \alpha_i - b_0 \alpha_i \right\} \propto \exp \left\{ \frac{1}{2} \ln \alpha_i - \frac{1}{2} E_{q(W)} W^T \alpha_i + (a_0 - 1) \ln \alpha_i - b_0 \alpha_i \right\}$$

$$\therefore q(\alpha_i) \propto \exp \left\{ \ln \alpha_i \left(\frac{1}{2} + a_0 - 1 \right) - \alpha_i \left(\frac{1}{2} E_{q(w)} [W_i^2] + b_0 \right) \right\}$$

$$\therefore q(\alpha_i) = \text{Gamma}(a'_i, b'_i) \text{ where } a'_i = \frac{1}{2} + a_0, \quad b'_i = \frac{1}{2} E_{q(w)} [W_i^2] + b_0.$$

$$\text{For this problem, } E_{q(\lambda)} [\lambda] = \frac{e'}{f'}, \quad E_{q(\alpha_i)} [\alpha_i] = \frac{a'}{b'}, \quad E_{q(w)} [w] = \mu', \quad E_{q(w)} [ww^T] = \Sigma' + \mu' \mu'^T$$

$$\text{Thus } q(w) = \text{Normal}(\mu', \Sigma') \text{ where } \Sigma' = \left(\text{diag} \left(\frac{a'_1}{b'_1}, \dots, \frac{a'_d}{b'_d} \right) + \frac{e'}{f'} \sum_{i=1}^N x_i x_i^T \right)^{-1}, \quad \mu' = \Sigma' \left(\frac{e'}{f'} \sum_{i=1}^N y_i x_i \right)$$

$$q(\lambda) = \text{Gamma}(e', f') \text{ where } e' = \frac{N}{2} + e_0, \quad f' = \frac{1}{2} \sum_{i=1}^N [(y_i - \mu'^T x_i)^2 + x_i^T \Sigma' x_i] + f_0.$$

$$q(\alpha_i) = \text{Gamma}(a'_i, b'_i) \text{ where } a'_i = \frac{1}{2} + a_0, \quad b'_i = \frac{1}{2} (\Sigma' + \mu' \mu'^T)_{[i, i]} + b_0.$$

(b) Pseudo-code.

1. Initialize $a_0, b_0, e_0, f_0, \mu_0$ and Σ_0 in some way.

2. For iteration $t=1, \dots, T$:

— for $k=1, \dots, d$, update $q(\alpha_k)$ by setting:

$$a'_{kt} = a_0 + \frac{1}{2}, \quad b'_{kt} = \frac{1}{2} (\Sigma'_{t-1} + \mu'_{t-1} \mu'_{t-1}^T)_{[k, k]} + b_0.$$

— update $q(\lambda)$ by setting:

$$e'_t = e_0 + \frac{N}{2}, \quad f'_t = \frac{1}{2} \sum_{i=1}^N [(y_i - \mu'_{t-1}^T x_i)^2 + x_i^T \Sigma'_{t-1} x_i] + f_0.$$

— update $q(w)$ by setting

$$\Sigma'_t = \left(\text{diag} \left(\frac{a'_{1t}}{b'_{1t}}, \dots, \frac{a'_{dt}}{b'_{dt}} \right) + \frac{e'_t}{f'_t} \sum_{i=1}^N x_i x_i^T \right)^{-1}, \quad \mu'_t = \Sigma'_t \left(\frac{e'_t}{f'_t} \sum_{i=1}^N y_i x_i \right)$$

— Evaluate $L(a_t, b_t, e_t, f_t, \Sigma_t, \mu_t)$ to assess convergence.

$$\begin{aligned}
(c). L &= \int q(w, \lambda, \alpha) \ln \frac{p(Y, w, \alpha, \lambda | X)}{q(w, \lambda, \alpha)} dw d\lambda d\alpha = \int q(w) q(\lambda) q(\alpha) \ln \frac{p(Y, w, \alpha, \lambda | X)}{q(w) q(\lambda) q(\alpha)} dw d\lambda d\alpha \\
&= \int q(w, \lambda, \alpha) \ln p(Y, w, \alpha, \lambda | X) dw d\lambda d\alpha - \int q(w) \ln q(w) dw - \int q(\lambda) \ln q(\lambda) d\lambda - \int q(\alpha) \ln q(\alpha) d\alpha \\
&= E_q[\ln p(Y, w, \alpha, \lambda | X)] - E_q[\ln q(w)] - E_q[\ln q(\lambda)] - E_q[\ln q(\alpha)] \\
&= E_q\left[\sum_{i=1}^N \ln p(y_i | w, \lambda, \alpha)\right] + E_q[\ln p(\lambda | e, f_0)] + E_q[\ln p(w | \alpha)] + E_q\left[\sum_{k=1}^d \ln p(\alpha_k | a_k, b_k)\right] \\
&\quad - E_q[\ln q(w | w', \Sigma')] - E_q[\ln q(\lambda | e', f')] - E_q\left[\sum_{k=1}^d \ln q(\alpha_k | a'_k, b'_k)\right] \\
&= (E_q[\ln p(\lambda | e, f_0)] - E_q[\ln q(\lambda | e', f')]) + (E_q[\ln p(w | \alpha)] - E_q[\ln q(w | w', \Sigma')]) \\
&\quad + (E_q\left[\sum_{k=1}^d \ln p(\alpha_k | a_k, b_k)\right] - E_q\left[\sum_{k=1}^d \ln q(\alpha_k | a'_k, b'_k)\right]) + E_q\left[\sum_{i=1}^N \ln p(y_i | w, \lambda, \alpha)\right]
\end{aligned}$$

$$\begin{aligned}
\text{Tr } ① \quad E_q[\ln p(\lambda | e, f_0)] - E_q[\ln p(\lambda | e', f')] &= E_q[(e_0 \ln f_0 - \ln \Gamma(e_0) + (e_0 - 1) \ln \lambda - f_0 \lambda) - (e' \ln f' - \ln \Gamma(e') + (e' - 1) \ln \lambda - f' \lambda)] \\
&= e_0 \ln f_0 - \ln \Gamma(e_0) + (e_0 - 1) E_q[\ln \lambda] - f_0 E_q[\lambda] - e' \ln f' + \ln \Gamma(e') - (e' - 1) E_q[\ln \lambda] + f' E_q[\lambda] \\
&= e_0 \ln f_0 - \ln \Gamma(e_0) + (e_0 - 1) (\psi(e') - \ln f') - f_0 \frac{e'}{f'} - e' (\ln f' + \ln \Gamma(e') - (e' - 1) (\psi(e') - \ln f') + f' \frac{e'}{f'}) \quad \text{where } \psi \text{ is the Digamma function} \\
&= e_0 \ln f_0 - \ln \Gamma(e_0) + (e_0 - e') (\psi(e') - \ln f') - (f_0 - f') \frac{e'}{f'} - e' \ln f' + \ln \Gamma(e')
\end{aligned}$$

$$\begin{aligned}
② \quad E_q[\ln p(w | \alpha)] - E_q[\ln q(w | w', \Sigma')] &= \frac{1}{2} \sum_{k=1}^d E_q[\ln \alpha_k] - \frac{1}{2} \ln 2\pi - \frac{1}{2} (E_q[w^T \text{diag}(\alpha_1, \dots, \alpha_d) w]) + \frac{1}{2} \ln |\Sigma'| + \frac{1}{2} \ln 2\pi + \frac{1}{2} E_q[(w - w')^T \Sigma'^{-1} (w - w')] \\
&= \frac{1}{2} \sum_{k=1}^d E_q[\ln \alpha_k] - \frac{1}{2} \sum_{k=1}^d E_q(w_k^2) E_q(\alpha_k) + \frac{1}{2} \ln |\Sigma'| + \frac{1}{2} E_q[w^T \Sigma'^{-1} w - 2 w^T \Sigma'^{-1} w' + w'^T \Sigma'^{-1} w'] \\
&= \frac{1}{2} \sum_{k=1}^d (\psi(\alpha'_k) - \ln b'_k) - \frac{1}{2} \sum_{k=1}^d [\Sigma' + w' w'^T]_{(k,k)} \frac{\alpha'_k}{b'_k} + \frac{1}{2} \ln |\Sigma'| + \frac{1}{2} (E_q(w^T \Sigma'^{-1} w) - w'^T \Sigma'^{-1} w') \\
\text{Since } E_q(w^T \Sigma'^{-1} w) - w'^T \Sigma'^{-1} w' &= E_q[\text{Tr}(\Sigma'^{-1} w w^T)] - \text{Tr}(\Sigma'^{-1} w' w'^T) = \sum_{k=1}^d [\Sigma'_{(k,k)} (E_q(w_k^2) - (w'_k)^2)] \\
&= \sum_{k=1}^d [\Sigma'^{-1}_{(k,k)} (\Sigma'_{(k,k)} + w' w'^T)_{(k,k)} - (w' w'^T)_{(k,k)}] = \sum_{k=1}^d [\Sigma'^{-1}_{(k,k)} \Sigma'_{(k,k)}] = d \\
\therefore E_q[\ln p(w | \alpha)] - E_q[\ln q(w | w', \Sigma')] &= \frac{1}{2} \sum_{k=1}^d (\psi(\alpha'_k) - \ln b'_k) - \frac{1}{2} \sum_{k=1}^d [\Sigma' + w' w'^T]_{(k,k)} \frac{\alpha'_k}{b'_k} + \frac{1}{2} \ln |\Sigma'| + \frac{d}{2}
\end{aligned}$$

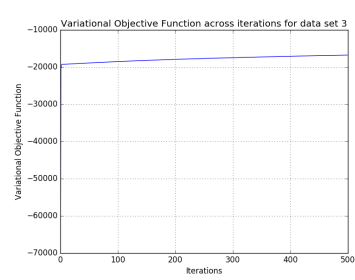
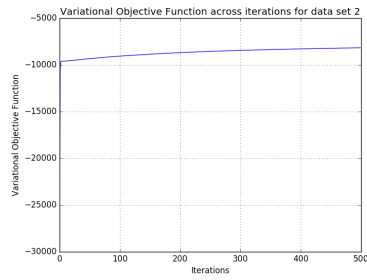
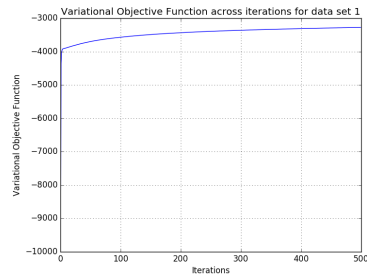
$$\begin{aligned}
 \textcircled{3} \quad E_q \left[\sum_{k=1}^d \ln \phi(\alpha_k | a_0, b_0) \right] - E_q \left[\sum_{k=1}^d \ln \phi(\alpha_k | a'_k, b'_k) \right] &= d a_0 \ln b_0 - d \ln \Gamma(a_0) + (a_0 - 1) \sum_{k=1}^d E_q(\ln a_k) - b_0 \sum_{k=1}^d E_q(a_k) \\
 &\quad - \sum_{k=1}^d (a'_k \ln b'_k - \ln \Gamma(a'_k) + (a'_k - 1) E_q[\ln \alpha_k] - b'_k E_q[\alpha_k]) = d(a_0 \ln b_0 - \ln \Gamma(a_0)) - \left(\sum_{k=1}^d a'_k \ln b'_k - \sum_{k=1}^d \ln \Gamma(a'_k) \right) \\
 &\quad + \sum_{k=1}^d (a_0 - a'_k) (\psi(a'_k) - \ln b'_k) - b_0 \sum_{k=1}^d \frac{a'_k}{b_k} + \sum_{k=1}^d a'_k
 \end{aligned}$$

$$\begin{aligned}
 \textcircled{4} \quad E_q \left[\sum_{i=1}^N \ln \phi(y_i | x_i, w_i, \lambda) \right] &= \frac{N}{2} E_q[\ln \lambda] - \frac{N}{2} \ln 2\pi - \frac{E_q[\lambda]}{2} \sum_{i=1}^N (y_i^2 - 2 E_q(W^T) x_i y_i + x_i^T E[W W^T] x_i) \\
 &= \frac{N}{2} (\psi(e') - \ln f') - \frac{N}{2} \ln 2\pi - \frac{f'}{2e'} \sum_{i=1}^N (y_i^2 - 2 \mu^T x_i y_i + x_i^T [\Sigma' + \mu \mu^T] x_i)
 \end{aligned}$$

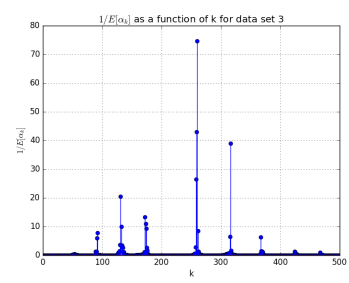
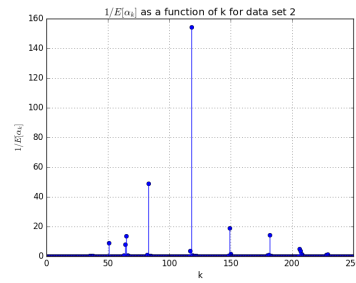
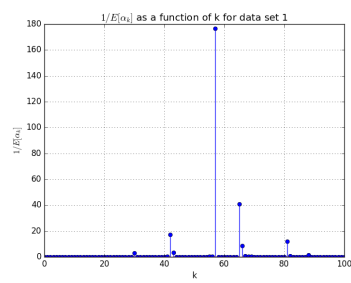
Thus we evaluate L by $\textcircled{1} + \textcircled{2} + \textcircled{3} + \textcircled{4}$

Problem 2

a)



b)



c)

$$\frac{1}{E_q[\lambda]} = 1.08002996$$

$$\frac{1}{E_q[\lambda]} = 0.89946298$$

$$\frac{1}{E_q[\lambda]} = 0.97814359$$

d)

