

# Supplementary files of “Goodness of fit of logistic models for random graphs” : Appendix

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# A Appendix

## A.1 Proof of Proposition 1

Let us start by showing that:

$$\log p(Y|Z, \alpha, \beta) \geq \log \sqrt{h(Z, \alpha, \beta, \xi)},$$

where  $\xi$  is an  $n \times n$  positive real matrix. We use the bound on the log-logistic function introduced by [1] from Taylor expansions:

$$\log g(x) \geq \log g(\xi) + \frac{x - \xi}{2} - \lambda(\xi)(x^2 - \xi^2), \forall (x, \xi) \in \mathbb{R} \times \mathbb{R}^+, \quad (1)$$

where  $\lambda(\xi) = (g(\xi) - 1/2)/(2\xi)$ . Note that (1) is an even function and therefore we can consider only positive values of  $x$  without loss of generality. Since

$$\log p(Y_{ij}|Z_i, Z_j, \alpha, \beta) = Y_{ij}(Z_i^\top \alpha Z_j + x_{ij}^\top \beta) + \log g(-Z_i^\top \alpha Z_j - x_{ij}^\top \beta),$$

then

$$\begin{aligned} \log p(Y_{ij}|Z_i, Z_j, \alpha, \beta) &\geq Y_{ij}(Z_i^\top \alpha Z_j + x_{ij}^\top \beta) + \log g(\xi_{ij}) - \frac{Z_i^\top \alpha Z_j + x_{ij}^\top \beta + \xi_{ij}}{2} \\ &\quad - \lambda(\xi_{ij})((Z_i^\top \alpha Z_j + x_{ij}^\top \beta)^2 - \xi_{ij}^2) \\ &= (Y_{ij} - \frac{1}{2})(Z_i^\top \alpha Z_j + x_{ij}^\top \beta) - \frac{\xi_{ij}}{2} + \log g(\xi_{ij}) \\ &\quad - \lambda(\xi_{ij})((Z_i^\top \alpha Z_j + x_{ij}^\top \beta)^2 - \xi_{ij}^2). \end{aligned}$$

Note that for undirected networks, the matrix  $\xi$  has to be symmetric, *i.e.*  $\xi_{ij} = \xi_{ji}, \forall i \neq j$ .

We then have

$$\log p(Y|Z, \alpha, \beta) = \frac{1}{2} \sum_{i \neq j}^n \log p(Y_{ij}|Z_i, Z_j, \alpha, \beta).$$

Therefore

$$\log p(Y|Z, \alpha, \beta) \geq \log \sqrt{h(Z, \alpha, \beta, \xi)}.$$

Finally,

$$\begin{aligned}
\mathcal{L}_K(q) &= \sum_Z \int q(Z, \pi, \alpha, \beta, \gamma, \eta) \log \frac{p(Y, Z, \pi, \alpha, \beta, \gamma, \eta)}{q(Z, \pi, \alpha, \beta, \gamma, \eta)} d\pi d\alpha d\beta d\gamma d\eta \\
&= \sum_Z \int q(Z, \pi, \alpha, \beta, \gamma, \eta) \log \frac{p(Y|Z, \alpha, \beta) p(Z, \pi, \alpha, \beta, \gamma, \eta)}{q(Z, \pi, \alpha, \beta, \gamma, \eta)} d\pi d\alpha d\beta d\gamma d\eta \\
&\geq \sum_Z \int q(Z, \pi, \alpha, \beta, \gamma, \eta) \log \frac{\sqrt{h(Z, \alpha, \beta, \xi)} p(Z, \pi, \alpha, \beta, \gamma, \eta)}{q(Z, \pi, \alpha, \beta, \gamma, \eta)} d\pi d\alpha d\beta d\gamma d\eta \\
&= \mathcal{L}_K(q; \xi).
\end{aligned}$$

## A.2 Proof of Proposition 2

$$\begin{aligned}
\log q(Z_i) &= \mathbb{E}_{Z \setminus i, \alpha, \beta, \pi} \left[ \frac{1}{2} \log h(Z, \alpha, \beta, \xi) + \log p(Z|\pi) \right] + \text{cst} \\
&= \mathbb{E}_{Z \setminus i, \alpha, \beta, \pi} \left[ \frac{1}{2} \sum_{i \neq j}^n \left\{ (Y_{ij} - \frac{1}{2}) Z_i^\top \alpha Z_j - \lambda(\xi_{ij}) ((Z_i^\top \alpha Z_j)^2 + 2Z_i^\top \alpha Z_i x_{ij}^\top \beta) \right\} \right. \\
&\quad \left. + \sum_{k=1}^n Z_{ik} \log \pi_k \right] + \text{cst}.
\end{aligned}$$

Note that

$$\begin{aligned}
\mathbb{E}_{Z_j, \alpha} [Z_i^\top \alpha Z_j] &= \mathbb{E}_{Z_j, \alpha} \left[ \sum_{k,l} Z_{ik} \alpha_{kl} Z_{jl} \right] \\
&= \sum_{k=1}^K Z_{ik} \left\{ \sum_{l=1}^K (m_\alpha)_{kl} \tau_{jl} \right\}.
\end{aligned}$$

Furthermore

$$\begin{aligned}
\mathbb{E}_{Z_j, \alpha} [(Z_i^\top \alpha Z_j)^2] &= \mathbb{E}_{Z_j, \alpha} \left[ \left( \sum_{k,l} Z_{ik} \alpha_{kl} Z_{jl} \right)^2 \right] \\
&= \mathbb{E}_{Z_j, \alpha} \left[ \sum_{k,k',l,l'} Z_{ik} Z_{ik'} \alpha_{kl} \alpha_{k'l'} Z_{jl} Z_{jl'} \right].
\end{aligned} \tag{2}$$

Because all vectors  $Z_i$  are sampled from a multinomial distribution with parameters  $(1, \pi)$ , all terms  $Z_{ik} Z_{ik'} = 0$  for  $k \neq k'$  and  $Z_{ik}^2 = Z_{ik}$  in (2). Therefore

$$(Z_i^\top \alpha Z_j)^2 = \sum_{k,l}^K Z_{ik} \alpha_{kl}^2 Z_{jl}. \tag{3}$$

This leads to

$$\mathbb{E}_{Z_j, \alpha} [(Z_i^\top \alpha Z_j)^2] = \sum_{k=1}^K Z_{ik} \left\{ \sum_{l=1}^K \mathbb{E}_{\alpha_{kl}} [\alpha_{kl}^2] \tau_{jl} \right\}.$$

Finally

$$\begin{aligned} \log q(Z_i) = & \sum_{k=1}^K Z_{ik} \left\{ \sum_{l=1}^K (m_\alpha)_{kl} \frac{1}{2} \sum_{j \neq i}^n \left( (Y_{ij} - \frac{1}{2}) - 2\lambda(\xi_{ij}) x_{ij}^\top m_\beta \right) \tau_{jl} - \sum_{l=1}^K \mathbb{E}_{\alpha_{kl}}[\alpha_{kl}^2] \frac{1}{2} \sum_{j \neq i}^n \lambda(\xi_{ij}) \tau_{jl} \right. \\ & + \sum_{l=1}^K (m_\alpha)_{lk} \frac{1}{2} \sum_{j \neq i}^n \left( (Y_{ji} - \frac{1}{2}) - 2\lambda(\xi_{ji}) x_{ji}^\top m_\beta \right) \tau_{jl} - \sum_{l=1}^K \mathbb{E}_{\alpha_{lk}}[\alpha_{lk}^2] \frac{1}{2} \sum_{j \neq i}^n \lambda(\xi_{ji}) \tau_{jl} \\ & \left. + \psi(e_k^n) - \psi\left(\sum_{l=1}^K e_l^n\right) \right\} + \text{cst.} \end{aligned}$$

Since  $(m_\alpha)_{kl} = (m_\alpha)_{lk}$ ,  $\mathbb{E}_{\alpha_{kl}}[\alpha_{kl}^2] = \mathbb{E}_{\alpha_{lk}}[\alpha_{lk}^2]$ ,  $Y_{ij} = Y_{ji}$ ,  $x_{ij} = x_{ji}$ ,  $\xi_{ij} = \xi_{ji}$ , then

$$\begin{aligned} \log q(Z_i) = & \sum_{k=1}^K Z_{ik} \left\{ \sum_{l=1}^K (m_\alpha)_{kl} \sum_{j \neq i}^n \left( (Y_{ij} - \frac{1}{2}) - 2\lambda(\xi_{ij}) x_{ij}^\top m_\beta \right) \tau_{jl} - \sum_{l=1}^K \mathbb{E}_{\alpha_{kl}}[\alpha_{kl}^2] \sum_{j \neq i}^n \lambda(\xi_{ij}) \tau_{jl} \right. \\ & \left. + \psi(e_k^n) - \psi\left(\sum_{l=1}^K e_l^n\right) \right\} + \text{cst.} \end{aligned}$$

Therefore

$$q(Z_i) = \mathcal{M}(Z_i; 1, \tau_i),$$

where  $\sum_{k=1}^K \tau_{ik} = 1$  and

$$\begin{aligned} \tau_{ik} \propto \exp \left\{ \sum_{l=1}^K (m_\alpha)_{kl} \sum_{j \neq i}^n \left( (Y_{ij} - \frac{1}{2}) - 2\lambda(\xi_{ij}) x_{ij}^\top m_\beta \right) \tau_{jl} - \sum_{l=1}^K \mathbb{E}_{\alpha_{kl}}[\alpha_{kl}^2] \sum_{j \neq i}^n \lambda(\xi_{ij}) \tau_{jl} \right. \\ \left. + \psi(e_k^n) - \psi\left(\sum_{l=1}^K e_l^n\right) \right\}. \end{aligned}$$

### A.3 Proof of Proposition 3

$$\begin{aligned} \log q(\pi) &= \mathbb{E}_Z [\log p(Z|\pi) + \log p(\pi)] + \text{cst} \\ &= \sum_{i=1}^n \sum_{k=1}^K \tau_{ik} \log \pi_k + \sum_{k=1}^K (e_0 - 1) \log \pi_k + \text{cst} \\ &= \sum_{k=1}^K \left( e_0 + \sum_{i=1}^n \tau_{ik} - 1 \right) \log \pi_k + \text{cst}. \end{aligned}$$

Therefore

$$q(\pi) = \text{Dir}(\pi; e^n),$$

where  $e_k^n = e_0 + \sum_{i=1}^n \tau_{ik}$ ,  $\forall k \in \{1, \dots, K\}$ .

#### A.4 Proof of Proposition 4

$$\begin{aligned}
\log q(\beta) &= \mathbb{E}_{Z, \alpha, \eta} \left[ \frac{1}{2} \log h(Z, \alpha, \beta, \xi) + \log p(\beta|\eta) \right] + \text{cst} \\
&= \mathbb{E}_{Z, \alpha, \eta} \left[ \frac{1}{2} \sum_{i \neq j}^n \left\{ (Y_{ij} - 1/2) x_{ij}^\top \beta - \lambda(\xi_{ij}) \left( (x_{ij}^\top \beta)^2 + 2Z_i^\top \alpha Z_j x_{ij}^\top \beta \right) \right\} - \frac{\eta}{2} \beta^\top \beta \right] + \text{cst} \\
&= \beta^\top \left\{ \frac{1}{2} \sum_{i \neq j}^n \left( Y_{ij} - \frac{1}{2} - 2\lambda(\xi_{ij}) \tau_i^\top m_\alpha \tau_j \right) x_{ij} \right\} - \frac{1}{2} \beta^\top \left\{ \frac{c_n}{d_n} I_d + \sum_{i \neq j}^n \lambda(\xi_{ij}) x_{ij} x_{ij}^\top \right\} \beta + \text{cst}.
\end{aligned}$$

Therefore

$$q(\beta) = \mathcal{N}(\beta; m_\beta, S_\beta),$$

where

$$S_\beta^{-1} = \frac{c_n}{d_n} I_d + \sum_{i \neq j}^n \lambda(\xi_{ij}) x_{ij} x_{ij}^\top,$$

and

$$m_\beta = S_\beta \frac{1}{2} \sum_{i \neq j}^n \left( Y_{ij} - \frac{1}{2} - 2\lambda(\xi_{ij}) \tau_i^\top m_\alpha \tau_j \right) x_{ij}.$$

#### A.5 Proof of Proposition 5

$$\begin{aligned}
\log q(\gamma) &= \mathbb{E}_\alpha [\log p(\alpha|\gamma) + \log p(\gamma)] + \text{cst} \\
&= \mathbb{E}_\alpha \left[ \sum_{k \leq l}^K \frac{1}{2} \log(\gamma) - \sum_{k \leq l}^K \frac{\gamma}{2} \alpha_{kl}^2 \right] + (a_0 - 1) \log \gamma - b_0 \gamma + \text{cst} \\
&= (a_0 + \frac{K(K+1)}{4} - 1) \log \gamma - \left( b_0 + \frac{1}{2} \sum_{k \leq l}^K \mathbb{E}_{\alpha_{kl}} [\alpha_{kl}^2] \right) \gamma + \text{cst}.
\end{aligned}$$

Therefore

$$q(\gamma) = \text{Gam}(\gamma; a_n, b_n),$$

where  $a_n = a_0 + \frac{K(K+1)}{4}$  and  $b_n = b_0 + \frac{1}{2} \sum_{k \leq l}^K \mathbb{E}_{\alpha_{kl}} [\alpha_{kl}^2]$ .

#### A.6 Proof of Proposition 6

$$\begin{aligned}
\log q(\eta) &= \mathbb{E}_\beta [\log p(\beta|\eta) + \log p(\eta)] + \text{cst} \\
&= \mathbb{E}_\beta \left[ \frac{d}{2} \log \eta - \frac{\eta}{2} \beta^\top \beta \right] + (c_0 - 1) \log \eta - d_0 \eta + \text{cst} \\
&= (c_0 + \frac{d}{2} - 1) \log \eta - (d_0 + \frac{1}{2} \text{Tr}(S_\beta) + \frac{1}{2} m_\beta^\top m_\beta) \eta + \text{cst}.
\end{aligned}$$

Therefore

$$q(\eta) = \text{Gam}(\eta; c_n, d_n),$$

where  $c_n = c_0 + \frac{d}{2}$  and  $d_n = d_0 + \frac{1}{2}\text{Tr}(S_\beta) + \frac{1}{2}m_\beta^\top m_\beta$ .

## A.7 Proof of Proposition 7

$$\begin{aligned} \log q(\alpha) &= \mathbb{E}_{Z, \beta, \gamma} \left[ \frac{1}{2} \log h(Z, \alpha, \beta, \xi) + \log p(\alpha | \gamma) \right] + \text{cst} \\ &= \mathbb{E}_{Z, \beta, \gamma} \left[ \frac{1}{2} \sum_{i \neq j}^n \left\{ (Y_{ij} - \frac{1}{2}) Z_i^\top \alpha Z_j - \lambda(\xi_{ij}) \left( (Z_i^\top \alpha Z_j)^2 + 2 Z_i^\top \alpha Z_j x_{ij}^\top \beta \right) \right\} \right. \\ &\quad \left. - \sum_{k \leq l}^K \frac{\gamma}{2} \alpha_{kl}^2 \right] + \text{cst}. \end{aligned} \quad (4)$$

We have  $Z_i^\top \alpha Z_j = \sum_{k,l}^K Z_{ik} \alpha_{kl} Z_{jl}$  and  $(Z_i^\top \alpha Z_j)^2 = \sum_{k,l}^K Z_{ik} \alpha_{kl}^2 Z_{jl} = Z_i^\top A Z_j$  (see Eq. 3) with  $A$  the  $K \times K$  matrix such that  $A_{kl} = \alpha_{kl}^2$ . Moreover, any expression of the form  $(1/2) \sum_{i \neq j}^n c_{ij} Z_i^\top B Z_j$  where  $B$  is a symmetric  $K \times K$  matrix and  $c_{ij} = c_{ji}$  can be written

$$\begin{aligned} \frac{1}{2} \sum_{i \neq j}^n c_{ij} Z_i^\top B Z_j &= \frac{1}{2} \sum_{i \neq j}^n c_{ij} \sum_{k,l}^K Z_{ik} B_{kl} Z_{jl} \\ &= \frac{1}{2} \sum_{i \neq j}^n c_{ij} \left( \sum_{k=1}^K Z_{ik} B_{kk} Z_{jk} + \sum_{k < l}^K Z_{ik} B_{kl} Z_{jl} + \sum_{k < l}^K Z_{jk} B_{lk} Z_{il} \right) \\ &= \sum_{k=1}^K B_{kk} \frac{1}{2} \sum_{i \neq j}^n c_{ij} Z_{ik} Z_{jk} + \sum_{k < l}^K B_{kl} \left( \frac{1}{2} \sum_{i \neq j}^n c_{ij} Z_{ik} Z_{jl} + \frac{1}{2} \sum_{i \neq j}^n c_{ij} Z_{jk} Z_{il} \right). \end{aligned}$$

By exchanging the role of  $i$  and  $j$  in the sum of the last term and since  $c_{ij} = c_{ji}$ , we obtain

$$\frac{1}{2} \sum_{i \neq j}^n c_{ij} Z_i^\top B Z_j = \sum_{k=1}^K B_{kk} \frac{1}{2} \sum_{i \neq j}^n c_{ij} Z_{ik} Z_{jk} + \sum_{k < l}^K B_{kl} \sum_{i \neq j}^n c_{ij} Z_{ik} Z_{jl}. \quad (5)$$

Using (5) in (4) leads to

$$\begin{aligned}
\log q(\alpha) &= \sum_{k=1}^K \alpha_{kk} \sum_{i \neq j}^n \left( \frac{1}{2} (Y_{ij} - \frac{1}{2}) - \lambda(\xi_{ij}) x_{ij}^\top m_\beta \right) \tau_{ik} \tau_{jk} \\
&\quad + \sum_{k < l}^K \alpha_{kl} \sum_{i \neq j}^n \left( (Y_{ij} - \frac{1}{2}) - 2\lambda(\xi_{ij}) x_{ij}^\top m_\beta \right) \tau_{ik} \tau_{jl} \\
&\quad - \sum_{k=1}^K \alpha_{kk}^2 \sum_{i \neq j}^n \frac{1}{2} \lambda(\xi_{ij}) \tau_{ik} \tau_{jk} - \sum_{k < l}^K \alpha_{kl}^2 \sum_{i \neq j}^n \lambda(\xi_{ij}) \tau_{ik} \tau_{jl} \\
&\quad - \sum_{k \leq l}^K \frac{a_n}{2b_n} \alpha_{kl}^2.
\end{aligned}$$

Therefore

$$q(\alpha) = \prod_{k \neq l}^K \mathcal{N}(\alpha_{kl}; (m_\alpha)_{kl}, (\sigma_\alpha^2)_{kl}),$$

where

$$\begin{aligned}
(\sigma_\alpha^2)_{kk}^{-1} &= \frac{a_n}{b_n} + \sum_{i \neq j}^n \lambda(\xi_{ij}) \tau_{ik} \tau_{jk}, \forall k, \\
(\sigma_\alpha^2)_{kl}^{-1} &= \frac{a_n}{b_n} + 2 \sum_{i \neq j}^n \lambda(\xi_{ij}) \tau_{ik} \tau_{jl}, \forall k \neq l, \\
(m_\alpha)_{kk} &= (\sigma_\alpha^2)_{kk} \sum_{i \neq j}^n \left( \frac{1}{2} (Y_{ij} - \frac{1}{2}) - \lambda(\xi_{ij}) x_{ij}^\top m_\beta \right) \tau_{ik} \tau_{jk} \\
(m_\alpha)_{kl} &= (\sigma_\alpha^2)_{kl} \sum_{i \neq j}^n \left( (Y_{ij} - \frac{1}{2}) - 2\lambda(\xi_{ij}) x_{ij}^\top m_\beta \right) \tau_{ik} \tau_{jl}.
\end{aligned}$$

## A.8 Proof of Proposition 8

Keeping only the terms that do depend on  $\xi_{ij}$  in (7), the lower bound is given by

$$\begin{aligned}
\mathcal{L}_K(q; \xi) &= \frac{1}{2} \sum_{i \neq j}^n \left\{ \log g(\xi_{ij}) - \frac{\xi_{ij}}{2} - \lambda(\xi_{ij}) \left( \mathbb{E}_{Z_i, Z_j, \alpha} [(Z_i^\top \alpha Z_j)^2] + 2 \mathbb{E}_{Z_i, Z_j, \alpha} [Z_i^\top \alpha Z_j] x_{ij}^\top \mathbb{E}_\beta [\beta] \right. \right. \\
&\quad \left. \left. + \mathbb{E}_\beta [(x_{ij}^\top \beta)^2] - \xi_{ij}^2 \right) \right\} + \text{cst} \\
&= \frac{1}{2} \sum_{i \neq j}^n \left\{ \log g(\xi_{ij}) - \frac{\xi_{ij}}{2} - \lambda(\xi_{ij}) \left( \sum_{k, l}^K \tau_{ik} \tau_{jl} \mathbb{E}_{\alpha_{kl}} [\alpha_{kl}^2] + 2 \sum_{k, l}^K \tau_{ik} \tau_{jl} (m_\alpha)_{kl} x_{ij}^\top m_\beta \right. \right. \\
&\quad \left. \left. + \text{Tr}(x_{ij} x_{ij}^\top (S_\beta + m_\beta m_\beta^\top)) - \xi_{ij}^2 \right) \right\} + \text{cst}.
\end{aligned}$$

The partial derivative of the lower bound with respect to  $\xi_{ij}$  is

$$\begin{aligned} \frac{\partial \mathcal{L}_K}{\partial \xi_{ij}}(q; \xi) = g(-\xi_{ij}) - \frac{1}{2} - \lambda'(\xi_{ij}) & \left( \sum_{k,l}^K \tau_{ik} \tau_{jl} \mathbb{E}_{\alpha_{kl}}[\alpha_{kl}^2] + 2 \sum_{k,l}^K \tau_{ik} \tau_{jl} (m_\alpha)_{kl} x_{ij}^\top m_\beta \right. \\ & \left. + \text{Tr}(x_{ij} x_{ij}^\top (S_\beta + m_\beta m_\beta^\top)) - \xi_{ij}^2 \right) + 2\lambda(\xi_{ij}) \xi_{ij}, \end{aligned}$$

and

$$\begin{aligned} \frac{\partial \mathcal{L}_K}{\partial \xi_{ij}}(q; \xi) = -\lambda'(\xi_{ij}) & \left( \sum_{k,l}^K \tau_{ik} \tau_{jl} \mathbb{E}_{\alpha_{kl}}[\alpha_{kl}^2] + 2 \sum_{k,l}^K \tau_{ik} \tau_{jl} (m_\alpha)_{kl} x_{ij}^\top m_\beta \right. \\ & \left. + \text{Tr}(x_{ij} x_{ij}^\top (S_\beta + m_\beta m_\beta^\top)) - \xi_{ij}^2 \right). \quad (6) \end{aligned}$$

Finally,  $\lambda(\xi_{ij})$  is a strictly decreasing function for positive values of  $\xi_{ij}$ . Thus,  $\lambda'(\xi_{ij}) \neq 0$  and therefore if we set (6) to zero, we obtain

$$\xi_{ij}^2 = \sum_{k,l}^K \tau_{ik} \tau_{jl} \mathbb{E}_{\alpha_{kl}}[\alpha_{kl}^2] + 2 \sum_{k,l}^K \tau_{ik} \tau_{jl} (m_\alpha)_{kl} x_{ij}^\top m_\beta + \text{Tr}(x_{ij} x_{ij}^\top (S_\beta + m_\beta m_\beta^\top)).$$

Note that  $\xi_{ij} = \xi_{ji}$ .

## A.9 Proof of Proposition 9

$$\mathcal{L}_K(q; \xi) = \mathbb{E}_{Z, \pi, \alpha, \beta, \gamma, \eta} \left[ \log \sqrt{h(Z, \alpha, \beta, \xi)} + \log p(Z, \pi, \alpha, \beta, \gamma, \eta) \right] - \mathbb{E}_{Z, \pi, \alpha, \beta, \gamma, \eta} [\log q(Z, \pi, \alpha, \beta, \gamma, \eta)]$$



$$\begin{aligned}
\mathcal{L}_K(q; \xi) = & \frac{1}{2} \sum_{i \neq j}^n \left\{ (Y_{ij} - \frac{1}{2})(\mathbb{E}_{Z_i, Z_j, \alpha}[Z_i^\top \alpha Z_j] + x_{ij}^\top \mathbb{E}_\beta[\beta]) + \log g(\xi_{ij}) - \frac{\xi_{ij}}{2} \right. \\
& \left. - \lambda(\xi_{ij}) \left( \mathbb{E}_{Z_i, Z_j, \alpha}[(Z_i^\top \alpha Z_j)^2] + 2\mathbb{E}_{Z_i, Z_j, \alpha}[Z_i^\top \alpha Z_j] x_{ij}^\top \mathbb{E}_\beta[\beta] + \mathbb{E}_\beta[(x_{ij}^\top \beta)^2] - \xi_{ij}^2 \right) \right\} \\
& + \sum_{i=1}^n \sum_{k=1}^K \mathbb{E}_{Z_i}[Z_{ik}] \mathbb{E}_\pi[\log \pi_k] - \log C(e) + \sum_{k=1}^K (e_0 - 1) \mathbb{E}_\pi[\log \pi_k] - \frac{K(K+1)}{4} \log(2\pi) \\
& + \frac{K(K+1)}{4} \mathbb{E}_\gamma[\log \gamma] - \frac{\mathbb{E}_\gamma[\gamma]}{2} \sum_{k \leq l}^K \mathbb{E}_{\alpha_{kl}}[\alpha_{kl}^2] - \frac{d}{2} \log(2\pi) + \frac{d}{2} \mathbb{E}_\eta[\log \eta] - \frac{\mathbb{E}_\eta[\eta]}{2} \mathbb{E}_\beta[\beta^\top \beta] \\
& - \log \Gamma(a_0) + a_0 \log b_0 + (a_0 - 1) \mathbb{E}_\gamma[\log \gamma] - b_0 \mathbb{E}_\gamma[\gamma] - \log \Gamma(c_0) \\
& + c_0 \log d_0 + (c_0 - 1) \mathbb{E}_\eta[\log \eta] - d_0 \mathbb{E}_\eta[\eta] - \sum_{i=1}^n \sum_{k=1}^K \mathbb{E}_{Z_i}[Z_{ik}] \log \tau_{ik} + \log C(e^n) \\
& - \sum_{k=1}^K (e_k^n - 1) \mathbb{E}_\pi[\log \pi_k] + \frac{K(K+1)}{4} \log(2\pi) + \frac{1}{2} \sum_{k \leq l}^K \log(\sigma_\alpha^2)_{kl} + \frac{1}{2} \sum_{k \leq l}^K (\sigma_\alpha^2)_{kl}^{-1} \mathbb{E}_{\alpha_{kl}}[\alpha_{kl}^2] \\
& - \sum_{k \leq l}^K (\sigma_\alpha^2)_{kl}^{-1} \mathbb{E}_{\alpha_{kl}}[\alpha_{kl}] (m_\alpha)_{kl} + \frac{1}{2} \sum_{k \leq l}^K (\sigma_\alpha^2)_{kl}^{-1} (m_\alpha)_{kl}^2 + \frac{d}{2} \log(2\pi) + \frac{1}{2} \log |S_\beta| \\
& + \frac{1}{2} \mathbb{E}_\beta[\beta^\top S_\beta^{-1} \beta] - \mathbb{E}_\beta[\beta]^\top S_\beta^{-1} m_\beta + \frac{1}{2} m_\beta^\top S_\beta^{-1} m_\beta + \log \Gamma(a_n) - a_n \log b_n \\
& - (a_n - 1) \mathbb{E}_\gamma[\log \gamma] + b_n \mathbb{E}_\gamma[\gamma] + \log \Gamma(c_n) - c_n \log d_n - (c_n - 1) \mathbb{E}_\eta[\log \eta] \\
& + d_n \mathbb{E}_\eta[\eta], \quad (7)
\end{aligned}$$

where  $C(x) = \frac{\prod_{k=1}^K \Gamma(x_k)}{\Gamma(\sum_{k=1}^K x_k)}$  and  $\Gamma(\cdot)$  is the gamma function. The terms in  $\mathbb{E}_\gamma[\log \gamma]$ ,  $\mathbb{E}_\eta[\log \eta]$ ,  $\mathbb{E}_\pi[\log \pi]$  and  $\log(2\pi)$  do simplify in (7) after the VBEM update step. This leads to

$$\begin{aligned}
\mathcal{L}_K(q; \xi) = & \frac{1}{2} \sum_{i \neq j}^n \left\{ (Y_{ij} - \frac{1}{2})(\mathbb{E}_{Z_i, Z_j, \alpha}[Z_i^\top \alpha Z_j] + x_{ij}^\top \mathbb{E}_\beta[\beta]) + \log g(\xi_{ij}) - \frac{\xi_{ij}}{2} \right. \\
& \left. - \lambda(\xi_{ij}) (\mathbb{E}_{Z_i, Z_j, \alpha}[(Z_i^\top \alpha Z_j)^2] + 2\mathbb{E}_{Z_i, Z_j, \alpha}[Z_i^\top \alpha Z_j] x_{ij}^\top \mathbb{E}_\beta[\beta] + \mathbb{E}_\beta[(x_{ij}^\top \beta)^2] - \xi_{ij}^2) \right\} \\
& - \log C(e) - \frac{a_n}{2b_n} \sum_{k \leq l}^K \mathbb{E}_{\alpha_{kl}}[\alpha_{kl}^2] - \frac{c_n}{2d_n} \text{Tr}(S_\beta + m_\beta m_\beta^\top) \\
& - \log \Gamma(a_0) + a_0 \log b_0 - b_0 \frac{a_n}{b_n} - \log \Gamma(c_0) + c_0 \log d_0 - d_0 \frac{c_n}{d_n} \\
& - \sum_{i=1}^n \sum_{k=1}^K \tau_{ik} \log \tau_{ik} + \log C(e^n) + \frac{1}{2} \sum_{k \leq l}^K \log(\sigma_\alpha^2)_{kl} + \frac{1}{2} \sum_{k \leq l}^K (\sigma_\alpha^2)_{kl}^{-1} \mathbb{E}_{\alpha_{kl}}[\alpha_{kl}^2] \\
& - \frac{1}{2} \sum_{k \leq l}^K (\sigma_\alpha^2)_{kl}^{-1} (m_\alpha)_{kl}^2 + \frac{1}{2} \log |S_\beta| + \frac{1}{2} \text{Tr}(S_\beta^{-1}(S_\beta + m_\beta(m_\beta)^\top)) \\
& - \frac{1}{2} m_\beta^\top S_\beta^{-1} m_\beta + \log \Gamma(a_n) - a_n \log b_n + b_n \frac{a_n}{b_n} + \log \Gamma(c_n) \\
& - c_n \log d_n + d_n \frac{c_n}{d_n}.
\end{aligned}$$

Moreover, using (5), note that

$$\begin{aligned}
\frac{1}{2} \sum_{i \neq j}^n (Y_{ij} - \frac{1}{2}) \mathbb{E}_{Z_i, Z_j, \alpha}[Z_i^\top \alpha Z_j] &= \sum_{k=1}^K \mathbb{E}_{\alpha_{kk}}[\alpha_{kk}] \frac{1}{2} \sum_{i \neq j}^n (Y_{ij} - \frac{1}{2}) \tau_{ik} \tau_{jk} \\
&+ \sum_{k < l}^K \mathbb{E}_{\alpha_{kl}}[\alpha_{kl}] \sum_{i \neq j}^n (Y_{ij} - \frac{1}{2}) \tau_{ik} \tau_{jl}, \\
\frac{1}{2} \sum_{i \neq j}^n 2\lambda(\xi_{ij}) \mathbb{E}_{Z_i, Z_j, \alpha}[Z_i^\top \alpha Z_j] x_{ij}^\top \mathbb{E}_\beta[\beta] &= \sum_{k=1}^K \mathbb{E}_{\alpha_{kk}}[\alpha_{kk}] \sum_{i \neq j}^n \lambda(\xi_{ij}) x_{ij}^\top \mathbb{E}_\beta[\beta] \tau_{ik} \tau_{jk} \\
&+ \sum_{k < l}^K \mathbb{E}_{\alpha_{kl}}[\alpha_{kl}] 2 \sum_{i \neq j}^n \lambda(\xi_{ij}) x_{ij}^\top \mathbb{E}_\beta[\beta] \tau_{ik} \tau_{jl}.
\end{aligned}$$

Using (3) and (5),

$$\begin{aligned} \frac{1}{2} \sum_{i \neq j}^n \lambda(\xi_{ij}) \mathbb{E}_{Z_i, Z_j, \alpha} [(Z_i^\top \alpha Z_j)^2] &= \sum_{k=1}^K \mathbb{E}_{\alpha_{kk}} [\alpha_{kk}^2] \frac{1}{2} \sum_{i \neq j}^n \lambda(\xi_{ij}) \tau_{ik} \tau_{jk} \\ &\quad + \sum_{k < l}^K \mathbb{E}_{\alpha_{kl}} [\alpha_{kl}^2] \sum_{i \neq j}^n \lambda(\xi_{ij}) \tau_{ik} \tau_{jl} \end{aligned}$$

Finally

$$\begin{aligned} \mathbb{E}_\beta [(x_{ij}^\top \beta)^2] &= \mathbb{E}_\beta [x_{ij}^\top \beta x_{ij}^\top \beta] \\ &= \mathbb{E}_\beta [x_{ij}^\top \beta \beta^\top x_{ij}] \\ &= \text{Tr} (x_{ij} x_{ij}^\top \mathbb{E}_\beta [\beta \beta^\top]) \\ &= \text{Tr} (x_{ij} x_{ij}^\top (S_\beta + m_\beta m_\beta^\top)) . \end{aligned}$$

Therefore

$$\begin{aligned} \mathcal{L}_K(q; \xi) &= \frac{1}{2} \sum_{i \neq j}^n \left\{ \log g(\xi_{ij}) - \frac{\xi_{ij}}{2} + \lambda(\xi_{ij}) \xi_{ij}^2 \right\} + \log \frac{C(e^n)}{C(e)} + \log \frac{\Gamma(a_n)}{\Gamma(a_0)} + \log \frac{\Gamma(c_n)}{\Gamma(c_0)} \\ &\quad + a_0 \log b_0 + a_n (1 - \frac{b_0}{b_n} - \log b_n) + c_0 \log d_0 + c_n (1 - \frac{d_0}{d_n} - \log d_n) \\ &\quad + \frac{1}{2} \sum_{k \leq l}^K \log(\sigma_\alpha^2)_{kl} + \frac{1}{2} \log |S_\beta| - \sum_{i=1}^n \sum_{k=1}^K \tau_{ik} \log \tau_{ik} - \frac{1}{2} \sum_{k \leq l}^K (\sigma_\alpha^2)_{kl}^{-1} (m_\alpha)_{kl}^2 - \frac{1}{2} m_\beta^\top S_\beta^{-1} m_\beta \\ &\quad + \sum_{k=1}^K (m_\alpha)_{kk} \sum_{i \neq j}^n \left( \frac{1}{2} (Y_{ij} - \frac{1}{2}) - \lambda(\xi_{ij}) x_{ij}^\top m_\beta \right) \tau_{ik} \tau_{jk} \\ &\quad + \sum_{k < l}^K (m_\alpha)_{kl} \sum_{i \neq j}^n \left( (Y_{ij} - \frac{1}{2}) - 2\lambda(\xi_{ij}) x_{ij}^\top m_\beta \right) \tau_{ik} \tau_{jl} \\ &\quad - \sum_{k=1}^K \mathbb{E}_{\alpha_{kk}} [\alpha_{kk}^2] \frac{1}{2} \left( \frac{a_n}{b_n} - (\sigma_\alpha^2)_{kk}^{-1} + \sum_{i \neq j}^n \lambda(\xi_{ij}) \tau_{ik} \tau_{jk} \right) \\ &\quad - \sum_{k < l}^K \mathbb{E}_{\alpha_{kl}} [\alpha_{kl}^2] \frac{1}{2} \left( \frac{a_n}{b_n} - (\sigma_\alpha^2)_{kl}^{-1} + 2 \sum_{i \neq j}^n \lambda(\xi_{ij}) \tau_{ik} \tau_{jl} \right) \\ &\quad + \frac{1}{2} m_\beta^\top \sum_{i \neq j}^n (Y_{ij} - \frac{1}{2}) x_{ij} \\ &\quad - \frac{1}{2} \text{Tr} \left( \left( 2 \sum_{i \neq j}^n \lambda(\xi_{ij}) x_{ij} x_{ij}^\top + \frac{c_n}{d_n} I_d - S_\beta^{-1} \right) (S_\beta + m_\beta m_\beta^\top) \right) . \end{aligned}$$

Finally, since the terms at the fourth and fifth line correspond exactly to  $\sum_{k \leq l}^K (m_\alpha)_{kl} (\sigma_\alpha^2)_{kl}^{-1} (m_\alpha)_{kl}$ , and after the VBEM update step

$$\begin{aligned} \mathcal{L}_K(q; \xi) = & \frac{1}{2} \sum_{i \neq j}^n \left\{ \log g(\xi_{ij}) - \frac{\xi_{ij}}{2} + \lambda(\xi_{ij}) \xi_{ij}^2 \right\} + \log \frac{C(e^n)}{C(e)} + \log \frac{\Gamma(a_n)}{\Gamma(a_0)} + \log \frac{\Gamma(c_n)}{\Gamma(c_0)} \\ & + a_0 \log b_0 + a_n \left(1 - \frac{b_0}{b_n} - \log b_n\right) + c_0 \log d_0 + c_n \left(1 - \frac{d_0}{d_n} - \log d_n\right) \\ & + \frac{1}{2} \sum_{k \leq l}^K \log(\sigma_\alpha^2)_{kl} + \frac{1}{2} \log |S_\beta| - \sum_{i=1}^n \sum_{k=1}^K \tau_{ik} \log \tau_{ik} + \frac{1}{2} \sum_{k \leq l}^K (\sigma_\alpha^2)_{kl}^{-1} (m_\alpha)_{kl}^2 - \frac{1}{2} m_\beta^\top S_\beta^{-1} m_\beta \\ & + \frac{1}{2} m_\beta^\top \sum_{i \neq j}^n \left(Y_{ij} - \frac{1}{2}\right) x_{ij}. \end{aligned}$$

## References

- [1] T.S. Jaakkola and M.I. Jordan. Bayesian parameter estimation via variational methods. *Statistics and Computing*, 10:25–37, 2000.