

$$P(y_{\text{act}}^i = 1 | x_{\text{act}}, \beta_n) = \frac{\exp(\beta_n^T x_{\text{act}}^i)}{\sum_{j'} \exp(\beta_n^T x_{\text{act}}^{ij'})}$$

$$\beta_n | \mathcal{G}, \Omega \stackrel{\text{iid}}{\sim} \mathcal{N}_K(\mathcal{G}, \Omega)$$

$$\mathcal{G} | \beta_0, \Omega_0 \sim \mathcal{N}_K(\beta_0, \Omega_0)$$

$$\Omega | S, r \sim W^{-1}(S^{-1}, r)$$

$$\begin{aligned} \text{KL}[q||p] &= \int \log \frac{q(\beta_n | \lambda)}{p(\beta_n | D, \mathcal{G}, \Omega)} = E \left[ \log \frac{q(\beta_n | \lambda)}{p(\beta_n | D, \mathcal{G}, \Omega)} \right] \\ &= -\mathcal{L}(\lambda, \mathcal{G}, \Omega) + \log p(D | \mathcal{G}, \Omega) \end{aligned}$$

$$P(\beta_n, \mathcal{G}, \Omega | D) = \frac{p(\mathcal{G}) p(\Omega) \prod_{h=1}^H p(\beta_n | \mathcal{G}, \Omega) \prod_{t=1}^{T_h} P(y_{\text{act}} | x_{\text{act}}, \beta_n)}{\int p(\mathcal{G}) p(\Omega) \prod_{h=1}^H p(\beta_n | \mathcal{G}, \Omega) \prod_{t=1}^{T_h} P(y_{\text{act}} | x_{\text{act}}, \beta_n) d\beta_n d\mathcal{G} d\Omega}$$

$$\mathcal{L}(\lambda; \mathcal{G}, \Omega) := -E_{q_n} \log \left[ \frac{q(\beta_n | \lambda)}{p(\beta_n, D | \mathcal{G}, \Omega)} \right]$$

$$q(\beta_n | \mu_n, \Sigma_n) \stackrel{\text{iid}}{\sim} \mathcal{N}(\mu_n, \Sigma_n)$$

$$H(q) + \sum_{h=1}^H E_q \log P(\beta_n | \mathcal{G}, \Omega) + \sum_{h=1}^H \sum_{t=1}^{T_h} E_q \log P(y_{\text{act}} | x_{\text{act}}, \beta_n)$$

$$P(y_{\text{act}} | x_{\text{act}}, \beta_n) = \prod_j \left[ \frac{\exp(x_{\text{act}}^T \beta_n)}{\sum_{j'} \exp(x_{\text{act}}^T \beta_n)} \right]^{y_{\text{act}}^j}$$

$$\begin{aligned} H(q) + \sum_{h=1}^H E_q [\log P(\beta_n | \mathcal{G}, \Omega)] + \sum_{h=1}^H \sum_{t=1}^{T_h} E_q \left[ \sum_{j=1}^J y_{\text{act}}^j (x_{\text{act}}^T \beta_n) - \log \left( \sum_{j=1}^J \exp(x_{\text{act}}^T \beta_n) \right) \right] \\ H(q) + \sum_{h=1}^H E_q [\log P(\beta_n | \mathcal{G}, \Omega)] + \sum_{h=1}^H \sum_{t=1}^{T_h} \left[ \sum_{j=1}^J y_{\text{act}}^j x_{\text{act}}^T \mu_n - E_q \left[ \log \left( \sum_{j=1}^J \exp(x_{\text{act}}^T \beta_n) \right) \right] \right] \end{aligned}$$

has no closed form



Delta method:  $Eg(v) \approx g(E(v)) + \frac{1}{2} \text{tr} \left[ \left( \frac{\partial g(E(v))}{\partial v \partial v^T} \right) \text{cov}(v) \right]$

$$\begin{aligned} E_q \left[ \log \left( \sum_{j=2}^J \exp(X_{\text{att},j}^T \beta_a) \right) \right] &= \log \left( \sum_{j=2}^J \exp(X_{\text{att},j}^T \mu_a) \right) + \frac{1}{2} \text{tr} \left[ \frac{\partial^2 \left[ \log \left( \sum_{j=2}^J \exp(X_{\text{att},j}^T \mu_a) \right) \right]}{\partial \mu_a \partial \mu_a^T} \Sigma_a \right] \\ &= \log \left( \sum_{j=2}^J \exp(X_{\text{att},j}^T \mu_a) \right) + \frac{1}{2} \text{tr} \left[ \frac{\partial^2 \left[ \log \left( \sum_{j=2}^J \exp(X_{\text{att},j}^T \mu_a) \right) \right]}{\partial \mu_a \partial \mu_a^T} \exp(0) \right] \\ &= \log \left( \sum_{j=2}^J \exp(X_{\text{att},j}^T \mu_a) \right) + \frac{1}{2} \left[ \left( \sum_{j=2}^J \exp(X_{\text{att},j}^T \mu_a) \right)^{-1} \left( \sum_{j=2}^J X_{\text{att},j} X_{\text{att},j}^T \right) \exp(X_{\text{att},j}^T \mu_a) - \left( \sum_{j=2}^J X_{\text{att},j}^T \exp(X_{\text{att},j}^T \mu_a) \right)^2 \right] \odot \left( \sum_{j=2}^J X_{\text{att},j}^T \exp(X_{\text{att},j}^T \mu_a) \right) \end{aligned}$$

Jensen inequality  $\varphi(E[X]) \leq E[\varphi(X)]$

$$\begin{aligned} H(a) + \sum_{h=2}^H E_q [\log p(\beta_a | \varphi, \Sigma)] + \sum_{h=2}^H \sum_{t=1}^{T_h} \left[ \sum_{i=1}^J y_{\text{att}}^i X_{\text{att},i}^T \mu_h - E_q \left[ \log \left( \sum_{j=2}^J \exp(X_{\text{att},j}^T \beta_a) \right) \right] \right] \\ = H(a) + \sum_{h=2}^H E_q [\log p(\beta_a | \varphi, \Sigma)] + \sum_{h=2}^H \sum_{t=1}^{T_h} \left[ \sum_{i=1}^J y_{\text{att}}^i X_{\text{att},i}^T \mu_h - \log \left( \sum_{j=2}^J \exp(X_{\text{att},j}^T \mu_h + \frac{1}{2} X_{\text{att},j}^T \Sigma_h X_{\text{att},j}) \right) \right] \\ = \frac{1}{2} \sum_{h=2}^H \log[(2\pi e)^K |\Sigma_h|] - \frac{H}{2} \log((2\pi)^K |\Sigma|) - \frac{1}{2} \text{tr} \left[ \Sigma^{-1} \sum_{h=2}^H \{ \Sigma_h + (\mu_h - \varphi)(\mu_h - \varphi)^T \} \right] \\ + \sum_{h=2}^H \sum_{t=1}^{T_h} \left[ \sum_{i=1}^J y_{\text{att}}^i X_{\text{att},i}^T \mu_h - \log \left( \sum_{j=2}^J \exp(X_{\text{att},j}^T \mu_h + \frac{1}{2} X_{\text{att},j}^T \Sigma_h X_{\text{att},j}) \right) \right] \end{aligned}$$