

Intigrate over x

$$\begin{aligned}
& \int p_{ij}(x_{ij}) \mathcal{N}(x_{ij}; \tilde{\mu}_{ij}(z_i), w_j^2) dx_{ij} \\
&= \int \exp(-\lambda x_{ij}^2) \frac{1}{\sqrt{2\pi}w_j} \exp\left(-\frac{(x_{ij} - \tilde{\mu}_{ij})^2}{2w_j^2}\right) dx_{ij} \\
&= \frac{1}{\sqrt{1 + 2\lambda w_j^2}} \exp\left(-\frac{\lambda \tilde{\mu}_{ij}^2}{1 + 2\lambda w_j^2}\right)
\end{aligned} \tag{1}$$

The conditional distribution of  $y$  is

$$\begin{aligned}
\log p(\mathbf{y}_i | \mathbf{z}_i) &= \sum_{j: y_{ij}=0} \log \left[ \frac{1}{\sqrt{1 + 2\lambda w_j^2}} \exp\left(-\frac{\lambda \tilde{\mu}_{ij}^2}{1 + 2\lambda w_j^2}\right) \right] \\
&+ \sum_{j: y_{ij} \neq 0} \log \left[ (1 - \exp(-\lambda y_{ij}^2)) \frac{1}{\sqrt{2\pi}w_j} \exp\left(-\frac{(y_{ij} - \tilde{\mu}_{ij})^2}{2w_j^2}\right) \right].
\end{aligned} \tag{2}$$

Our target is to get  $p(z_i | y_i)$ , so we can use VAE.

Encoder:

$$\begin{aligned}
\boldsymbol{\mu}_i, \boldsymbol{\sigma}_i^2 &= f_\phi(\mathbf{y}_i) \\
q_\phi(\mathbf{z}_i | \mathbf{y}_i) &= \mathcal{N}(\boldsymbol{\mu}_i, \boldsymbol{\sigma}_i^2)
\end{aligned} \tag{3}$$

Decoeder:

$$p(\mathbf{z}_i) = \mathcal{N}(\mathbf{0}, \mathbf{I}) \tag{4}$$

$$\tilde{\mathbf{x}}_i = f_\theta(\mathbf{z}_i) \tag{5}$$

The evidence lower bound (ELBO):

$$elbo = \mathbb{E}_{q_\phi(\mathbf{z}_i | \mathbf{y}_i)} [\log p_{\boldsymbol{\theta}, \mathbf{w}, \boldsymbol{\lambda}}(\mathbf{y}_i | \mathbf{z}_i)] - D_{KL}(q_\phi(\mathbf{z}_i | \mathbf{y}_i) \| p(\mathbf{z}_i)) \tag{6}$$