

1. **Algorithm 1:** Use the (collapsed) Gibbs sampling algorithm to approximate (using samples) the posterior distribution  $p(\mathbf{Z}, \Theta \mid \mathbf{X})$ . Derive the posterior.

$$\underline{p(z_n = k \mid \mathbf{x}_n, \mathbf{Z}_{-n}, \Theta) :}$$

$$\begin{aligned} p(z_n = k \mid \mathbf{x}_n, \mathbf{Z}_{-n}, \Theta) &\propto p(z_n = k, \mathbf{x}_n \mid \mathbf{Z}_{-n}, \Theta) \\ &= p(z_n = k \mid \mathbf{Z}_{-n}, \Theta) p(\mathbf{x}_n \mid z_n = k, \mathbf{Z}_{-n}, \Theta) \end{aligned}$$

$$\underline{1 \leq k \leq K^+ :}$$

From Lecture 4:

$$p(z_n = k \mid \mathbf{Z}_{-n}) = \frac{m_k}{N - 1 + \alpha} \quad , \text{ where } m_k = \sum_{i=1}^{n-1} [z_i = k]$$

$$p(\mathbf{x}_n \mid z_n = k, \mathbf{Z}_{-n}, \Theta) = p(\mathbf{x}_n \mid \theta_k) = \prod_{i=1}^{W_n} \text{Cat}(x_{ni} \mid \theta_k)$$

$$\begin{aligned} \Rightarrow p(z_n = k \mid \mathbf{x}_n, \mathbf{Z}_{-n}, \Theta) &\propto \frac{m_k}{N - 1 + \alpha} \cdot \prod_{i=1}^{W_n} \text{Cat}(x_{ni} \mid \theta_k) \\ &= \frac{m_k}{N - 1 + \alpha} \cdot \prod_{m=1}^{|I|} \theta_{km}^{\sum_{i=1}^{W_n} [x_{ni}=m]} \end{aligned}$$

$$\begin{aligned} \Rightarrow \log p(z_n = k \mid \mathbf{x}_n, \mathbf{Z}_{-n}, \Theta) &\propto \log p(z_n = k \mid \mathbf{Z}_{-n}, \Theta) + \log p(\mathbf{x}_n \mid z_n = k, \mathbf{Z}_{-n}, \Theta) \\ &= \log m_k - \log(N - 1 + \alpha) + \sum_{m=1}^{|I|} \sum_{i=1}^{W_n} [x_{ni} = m] \log \theta_{km} \end{aligned}$$

$$\underline{k = K_{new} :}$$

From Lecture 4:

$$p(z_n = K_{new} \mid \mathbf{Z}_{-n}) = \frac{\alpha}{N - 1 + \alpha}$$

$$p(\mathbf{x}_n \mid z_n = K_{new}, \mathbf{Z}_{-n}, \Theta) = p(\mathbf{x}_n \mid z_n = K_{new}),$$

since  $\mathbf{Z}_{-n}$  contains no assignments to  $K_{new}$  and  $\Theta$  does not contain  $\theta_{K_{new}}$

$$\begin{aligned} &= \int p(\mathbf{x}_n, \theta \mid z_n = K_{new}) d\theta \\ &= \int p(\mathbf{x}_n \mid z_n = K_{new}, \theta) p(\theta \mid \gamma) d\theta \end{aligned}$$

$$\begin{aligned}
p(\mathbf{x}_n \mid z_n = K_{new}) &= \int p(\mathbf{x}_n \mid z_n = K_{new}, \boldsymbol{\theta}) p(\boldsymbol{\theta} \mid \boldsymbol{\gamma}) d\boldsymbol{\theta} \\
&= \int \text{Dir}(\boldsymbol{\theta} \mid \boldsymbol{\gamma}) \prod_{i=1}^{W_n} \text{Cat}(x_{ni} \mid \boldsymbol{\theta}) d\boldsymbol{\theta} \\
&= \int \frac{1}{B(\boldsymbol{\gamma})} \left( \prod_{m=1}^{|I|} \theta_m^{\gamma_m-1} \right) \left( \prod_{m'=1}^{|I|} \theta_{m'}^{\sum_{i=1}^{W_n} [x_{ni}=m']} \right) d\boldsymbol{\theta} \\
&= \int \frac{B(\boldsymbol{\gamma} + \mathbf{s})}{B(\boldsymbol{\gamma})} \frac{1}{B(\boldsymbol{\gamma} + \mathbf{s})} \prod_{m=1}^{|I|} \theta_m^{\gamma_m+s_m-1} d\boldsymbol{\theta} \quad , \text{ where } s_m = \sum_{i=1}^{W_n} [x_{ni} = m] \\
&= \frac{B(\boldsymbol{\gamma} + \mathbf{s})}{B(\boldsymbol{\gamma})} \int \text{Dir}(\boldsymbol{\theta} \mid \boldsymbol{\gamma} + \mathbf{s}) d\boldsymbol{\theta} \\
&= \frac{B(\boldsymbol{\gamma} + \mathbf{s})}{B(\boldsymbol{\gamma})} \\
&= \frac{\prod_{m=1}^{|I|} \prod_{i=0}^{s_m-1} (\gamma_m + i)}{\prod_{j=0}^{W_n-1} \left( i + \sum_{m'=1}^{|I|} \gamma_{m'} \right)}, \quad (\text{as shown in Tutorial 3})
\end{aligned}$$

$$\Rightarrow p(z_n = K_{new} \mid \mathbf{x}_n, \mathbf{Z}_{-n}, \boldsymbol{\Theta}) \propto \frac{\alpha}{N-1+\alpha} \cdot \frac{\prod_{m=1}^{|I|} \prod_{i=0}^{s_m-1} (\gamma_m + i)}{\prod_{j=0}^{W_n-1} \left( i + \sum_{m'=1}^{|I|} \gamma_{m'} \right)}$$

$$\underline{p(\boldsymbol{\theta} \mid \mathbf{x}_n)} :$$

$$\begin{aligned}
p(\boldsymbol{\theta} \mid \mathbf{x}_n) &\propto p(\mathbf{x}_n \mid \boldsymbol{\theta}) p(\boldsymbol{\theta}) = p(\mathbf{x}_n \mid \boldsymbol{\theta}) p(\boldsymbol{\theta} \mid \boldsymbol{\gamma}) \\
&= \prod_{i=1}^{W_n} \text{Cat}(x_{ni} \mid \boldsymbol{\theta}) \text{Dir}(\boldsymbol{\theta} \mid \boldsymbol{\gamma}) \\
&= \prod_{m=1}^{|I|} \theta_m^{\sum_{i=1}^{W_n} [x_{ni}=m]} \frac{1}{B(\boldsymbol{\gamma})} \prod_{m'=1}^{|I|} \theta_{m'}^{\gamma_{m'}-1} \\
&= \frac{1}{B(\boldsymbol{\gamma})} \prod_{m=1}^{|I|} \theta_m^{\gamma_m+s_m-1}, \quad \text{ where } s_m = \sum_{i=1}^{W_n} [x_{ni} = m] \\
&\propto \text{Dir}(\boldsymbol{\theta} \mid \boldsymbol{\gamma} + \mathbf{s})
\end{aligned}$$

$$\Rightarrow p(\boldsymbol{\theta} \mid \mathbf{x}_n) = \text{Dir}(\boldsymbol{\theta} \mid \boldsymbol{\gamma} + \mathbf{s})$$

$p(\boldsymbol{\theta}_k \mid \mathbf{X}, \mathbf{Z}) :$

$$\begin{aligned}
 p(\boldsymbol{\theta}_k \mid \mathbf{X}, \mathbf{Z}) &\propto p(\boldsymbol{\theta}_k, \mathbf{X} \mid \mathbf{Z}) = p(\boldsymbol{\theta}_k) p(\mathbf{X} \mid \boldsymbol{\theta}_k, \mathbf{Z}) \\
 &= \text{Dir}(\boldsymbol{\theta}_k \mid \boldsymbol{\gamma}) \prod_{n=1}^N p(\mathbf{x}_n \mid \boldsymbol{\theta}_k, z_n) \\
 &= \text{Dir}(\boldsymbol{\theta}_k \mid \boldsymbol{\gamma}) \prod_{n=1}^N p(\mathbf{x}_n \mid \boldsymbol{\theta}_k)^{[z_n=k]} \\
 &= \text{Dir}(\boldsymbol{\theta}_k \mid \boldsymbol{\gamma}) \prod_{n=1}^N \prod_{i=1}^{W_n} \text{Cat}(x_{ni} \mid \boldsymbol{\theta}_k)^{[z_n=k]} \\
 &= \text{Dir}(\boldsymbol{\theta}_k \mid \boldsymbol{\gamma}) \prod_{n=1}^N \prod_{i=1}^{W_n} \prod_{m=1}^{|I|} \theta_{km}^{[x_{ni}=m][z_n=k]} \\
 &= \text{Dir}(\boldsymbol{\theta}_k \mid \boldsymbol{\gamma}) \prod_{m=1}^{|I|} \theta_{km}^{\sum_{n=1}^N [z_n=k] \sum_{i=1}^{W_n} [x_{ni}=m]} \\
 &= \prod_{m=1}^{|I|} \theta_{km}^{\gamma_m + c_{km} - 1}, \quad \text{where, } c_{km} = \sum_{n=1}^N [z_n = k] \sum_{i=1}^{W_n} [x_{ni} = m] \\
 &\propto \text{Dir}(\boldsymbol{\theta}_k \mid \boldsymbol{\gamma} + \mathbf{c}_k)
 \end{aligned}$$

$$\Rightarrow p(\boldsymbol{\theta}_k \mid \mathbf{X}, \mathbf{Z}) = \text{Dir}(\boldsymbol{\theta}_k \mid \boldsymbol{\gamma} + \mathbf{c}_k)$$

2. **Algorithm 2:** Use the (collapsed) Gibbs sampling algorithm to approximate (using samples) the posterior distribution  $p(\mathbf{Z} \mid \mathbf{X})$ . Derive the posterior.

$$\underline{p(z_n = k \mid \mathbf{x}_n, \mathbf{Z}_{-n})} :$$

$$\begin{aligned} p(z_n = k \mid \mathbf{x}_n, \mathbf{Z}_{-n}) &\propto p(z_n = k, \mathbf{x}_n \mid \mathbf{X}_{-n}, \mathbf{Z}_{-n}) \\ &= p(z_n = k \mid \mathbf{Z}_{-n}) p(\mathbf{x}_n \mid z_n = k, \mathbf{X}_{-n}, \mathbf{Z}_{-n}) \end{aligned}$$

$$\underline{1 \leq k \leq K^+} :$$

From Algorithm 1:

$$p(z_n = k \mid \mathbf{Z}_{-n}) = \frac{m_k}{N - 1 + \alpha} \quad , \text{ where } m_k = \sum_{i=1}^{n-1} [z_i = k]$$

From Tutorial 3:

$$p(\mathbf{x}_n \mid z_n = k, \mathbf{X}_{-n}, \mathbf{Z}_{-n}) = \frac{\prod_{m=1}^{|I|} \prod_{i=0}^{s_m-1} (\gamma'_m + i)}{\prod_{i=0}^{W_n-1} (i + \sum_{m'=1}^{|I|} \gamma'_{m'})} , \text{ where } \gamma'_m = \gamma_m + \sum_{i \neq n} \sum_{j=1}^{W_n} [x_{ij} = m] [z_i = k]$$

$$\Rightarrow p(z_n = k \mid \mathbf{x}_n, \mathbf{Z}_{-n}) \propto \frac{m_k}{N - 1 + \alpha} \cdot \frac{\prod_{m=1}^{|I|} \prod_{i=0}^{s_m-1} (\gamma'_m + i)}{\prod_{i=0}^{W_n-1} (i + \sum_{m'=1}^{|I|} \gamma'_{m'})}$$

$$\underline{k = K_{new} :}$$

From Algorithm 1:

$$p(z_n = K_{new} \mid \mathbf{Z}_{-n}) = \frac{\alpha}{N - 1 + \alpha}$$

$$p(\mathbf{x}_n \mid z_n = K_{new}, \mathbf{X}_{-n}, \mathbf{Z}_{-n}) = p(\mathbf{x}_n \mid z_n = K_{new}),$$

since  $\mathbf{Z}_{-n}$  does not contain assignments to  $K_{new}$

$$= \frac{\prod_{m=1}^{|I|} \prod_{i=0}^{s_m-1} (\gamma_m + i)}{\prod_{j=0}^{W_n-1} (i + \sum_{m'=1}^{|I|} \gamma_{m'})}$$

$$\Rightarrow p(z_n = k \mid \mathbf{x}_n, \mathbf{Z}_{-n}) \propto \frac{\alpha}{N - 1 + \alpha} \cdot \frac{\prod_{m=1}^{|I|} \prod_{i=0}^{s_m-1} (\gamma_m + i)}{\prod_{j=0}^{W_n-1} (i + \sum_{m'=1}^{|I|} \gamma_{m'})}$$