

Gibbs sampling for B-LDA

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Algorithm 1 Gibbs Sampling for B-LDA.

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1: procedure GIBBSAMPLING
2:   for each user  $u = 1, \dots, U$  do
3:     for  $u$ 's  $n$ -th tweet,  $n = 1, \dots, N_u$  do
4:       Randomly assign a topic to  $z_{u,n}$ 
5:       for each word  $l = 1, \dots, L_{u,n}$  do
6:         Randomly assign 0 or 1 to  $y_{u,n,l}$ 
7:       end for
8:     end for
9:   end for
10:  for each Gibbs Sampling iteration do
11:    for each user  $u = 1, \dots, U$  do
12:      for  $u$ 's  $n$ -th tweet,  $n = 1, \dots, N_u$  do
13:        Draw a topic  $z_{u,n}$  according to Eqn. 2.1
14:        for each word  $l = 1, \dots, L_{u,n}$  do
15:          Draw  $y_{u,n,l}$  according to Eqn. 2.2
16:        end for
17:      end for
18:    end for
19:  end for
20:  Estimate model parameters  $\theta, \varphi, \phi', \phi$  and  $\psi$ 
21: end procedure

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Notations	Descriptions
U	the total number of users
N_u	the total number of tweets in user u
$L_{u,n}$	the total number of words in u 's n -th tweet
T	the total number of topics
b	a behavior in $\mathcal{B} = \{post, retweet, reply, mention\}$
y	a switch
z	a topic label
w	a word label
<hr/>	
ϕ_t	topic-specific word distribution
ψ_t	topic-specific behavior distribution
ϕ'	background word distribution
θ_u	user-specific topic distribution
φ	Bernoulli distribution
$\alpha, \eta, \beta', \beta, \gamma$	Dirichlet priors

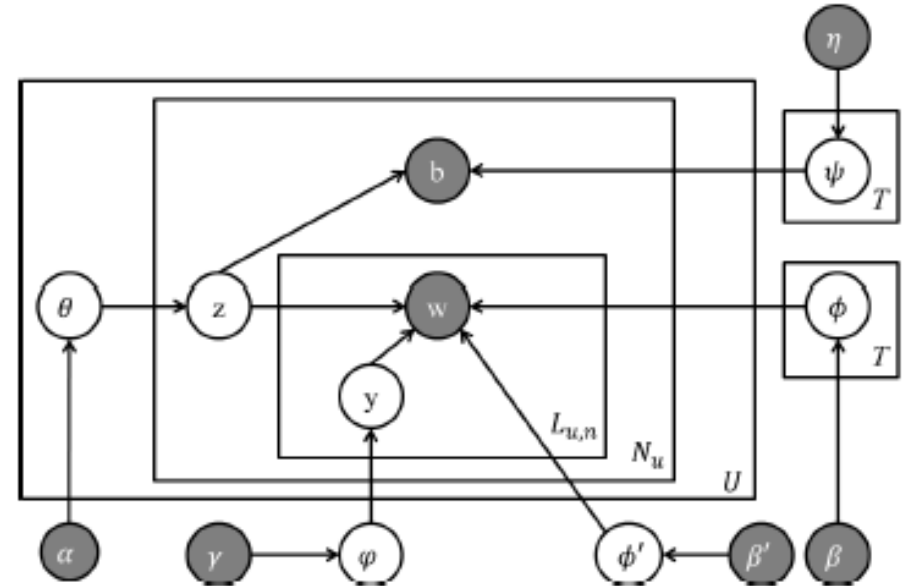
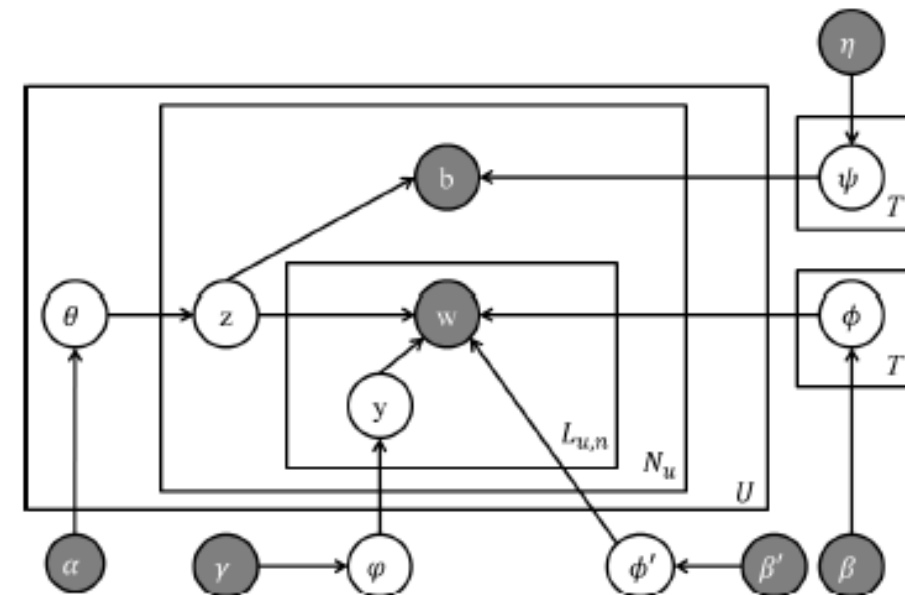


Figure 1: LDA-based behavior-topic model (B-LDA)

Hence, the problem is to compute the following two updating rules.

- To sample label $y_{u,n,l}$:



□ Sampling topic $z_{u,n}$:

$$p(\vec{Z}, \vec{W}, \vec{Y}, \vec{B} | \vec{\eta}, \vec{\beta}, \vec{\beta}', \vec{\gamma}, \vec{\alpha}) = p(\vec{Z} | \vec{\alpha}) p(\vec{W} | \vec{Z}, \vec{Y}, \vec{\beta}, \vec{\beta}') p(\vec{Y} | \vec{\gamma}) p(\vec{B} | \vec{Z}, \vec{\eta})$$

$$\begin{aligned} p(\vec{Z} | \vec{\alpha}) &= \int p(\vec{Z} | \vec{\theta}) p(\vec{\theta} | \vec{\alpha}) d\vec{\theta} = \int \prod_{u=1}^U \prod_{t=1}^T p_t^{n_u^t} \text{Dirichlet}(\vec{\alpha}) d\vec{\theta}_u \\ &= \prod_{u=1}^U \int \prod_{t=1}^T p_t^{n_u^t} \frac{1}{\Delta(\vec{\alpha})} \prod_{t=1}^T p_t^{\alpha_t - 1} d\vec{\theta}_u \\ &= \prod_{u=1}^U \int \frac{1}{\Delta(\vec{\alpha})} \prod_{t=1}^T p_t^{n_u^t + \alpha_t - 1} d\vec{\theta}_u \\ &= \prod_{u=1}^U \frac{\Delta(\vec{n}_u + \vec{\alpha})}{\Delta(\vec{\alpha})} \end{aligned}$$

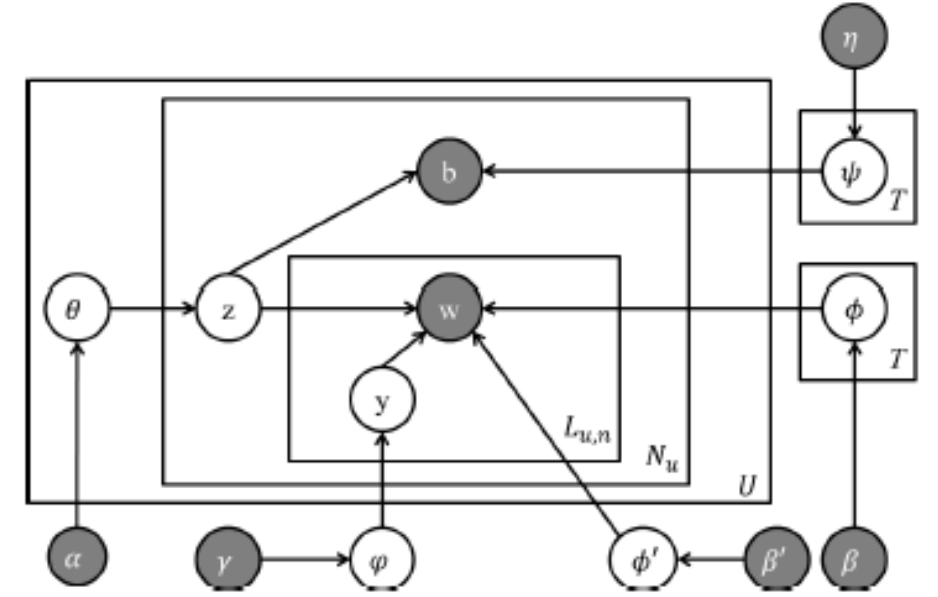


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□ Sampling topic $z_{u,n}$:

$$p(\vec{Z}, \vec{W}, \vec{Y}, \vec{B} | \vec{\eta}, \vec{\beta}, \vec{\beta}', \vec{\gamma}, \vec{\alpha}) = p(\vec{Z} | \vec{\alpha}) p(\vec{W} | \vec{Z}, \vec{Y}, \vec{\beta}, \vec{\beta}') p(\vec{Y} | \vec{\gamma}) p(\vec{B} | \vec{Z}, \vec{\eta})$$

$$\begin{aligned} p(\vec{B} | \vec{Z}, \vec{\eta}) &= \prod_{t=1}^T \int p(\vec{B} | \vec{Z}_t, \vec{\psi}_t) p(\vec{\psi}_t | \vec{\eta}) d\vec{\psi}_t \\ &= \prod_{t=1}^T \int \prod_{b=1}^B p_{t,b}^{n_t^b} \text{Dirichlet}(\vec{\eta}) d\vec{\psi}_t \\ &= \prod_{t=1}^T \int \frac{1}{\Delta(\vec{\eta})} \prod_{b=1}^B p_{t,b}^{n_t^b + \eta_b - 1} d\vec{\psi}_t \\ &= \prod_{t=1}^T \frac{\Delta(\vec{n}_t + \vec{\eta})}{\Delta(\vec{\eta})} \end{aligned}$$

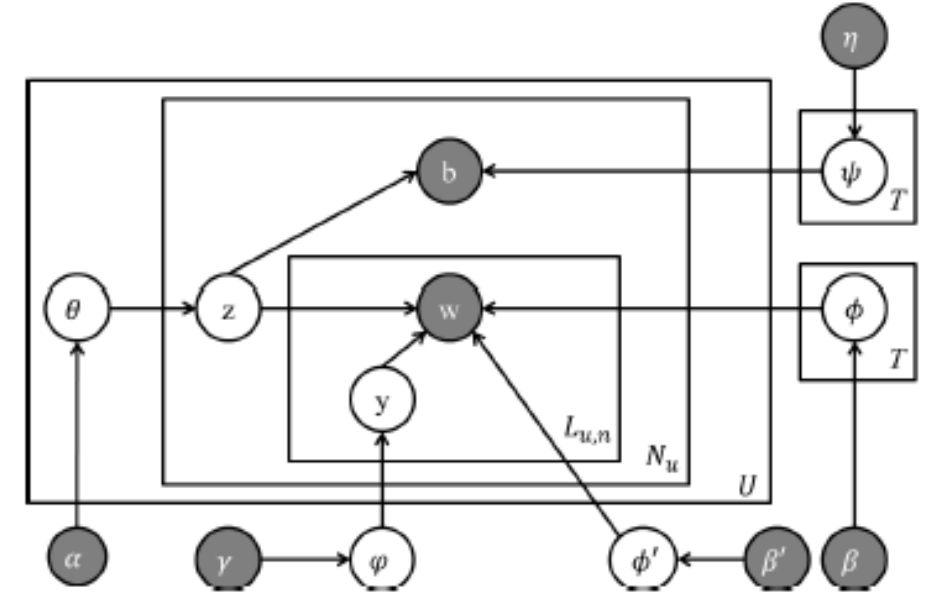


Figure 1: LDA-based behavior-topic model (B-LDA)

□ Sampling topic $z_{u,n}$:

$$p(\vec{Z}, \vec{W}, \vec{Y}, \vec{B} | \vec{\eta}, \vec{\beta}, \vec{\beta}', \vec{\gamma}, \vec{\alpha}) = p(\vec{Z} | \vec{\alpha}) p(\vec{W} | \vec{Z}, \vec{Y}, \vec{\beta}, \vec{\beta}') p(\vec{Y} | \vec{\gamma}) p(\vec{B} | \vec{Z}, \vec{\eta})$$

$$\begin{aligned} p(\vec{Y} | \vec{\gamma}) &= \int p(\vec{Y} | \vec{\varphi}) p(\vec{\varphi} | \vec{\gamma}) d\vec{\varphi} = \int \prod_{y=0}^1 p_y^{n_y} \text{Dirichlet}(\vec{\gamma}) d\vec{\varphi} \\ &= \int \prod_{y=0}^1 p_y^{n_y} \frac{1}{\Delta(\vec{\gamma})} \prod_{y=0}^1 p_y^{\gamma_y - 1} d\vec{\varphi} \\ &= \int \frac{1}{\Delta(\vec{\gamma})} \prod_{y=0}^1 p_y^{n_y + \gamma_y - 1} d\vec{\varphi} \\ &= \frac{\Delta(n_{(\cdot)}^y + \vec{\gamma})}{\Delta(\vec{\gamma})} \end{aligned}$$

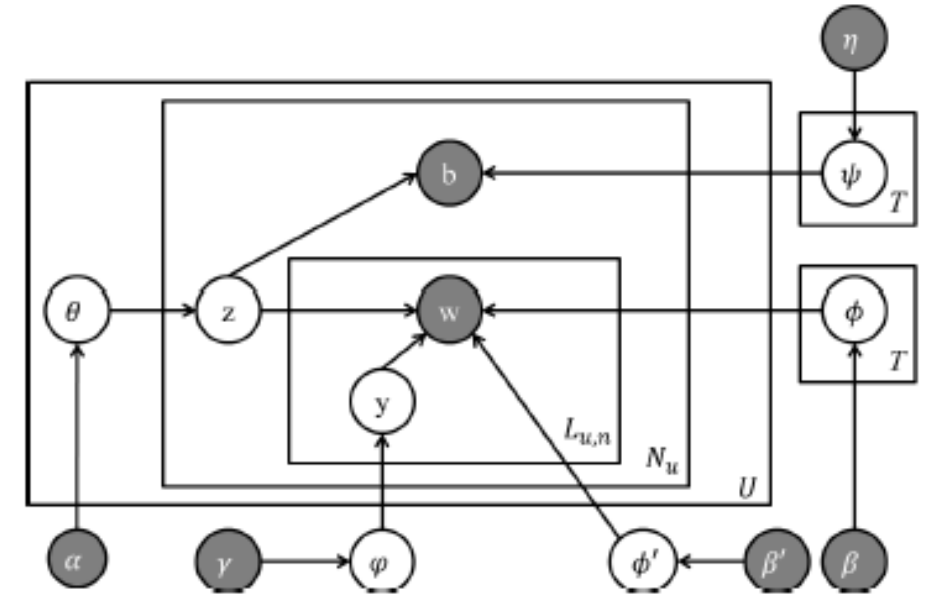


Figure 1: LDA-based behavior-topic model (B-LDA)

□ Sampling topic $z_{u,n}$:

$$p(\vec{Z}, \vec{W}, \vec{Y}, \vec{B} | \vec{\eta}, \vec{\beta}, \vec{\beta}', \vec{\gamma}, \vec{\alpha}) = p(\vec{Z} | \vec{\alpha}) p(\vec{W} | \vec{Z}, \vec{Y}, \vec{\beta}, \vec{\beta}') p(\vec{Y} | \vec{\gamma}) p(\vec{B} | \vec{Z}, \vec{\eta})$$

We assume each word has a corresponding label y that indicates which model it is sampled from. Specifically, if $y = 0$, the word is sampled from the background model; if $y = 1$, it is from a topic specific model. To derive $p(\vec{W} | \vec{Z}, \vec{Y}, \vec{\beta}, \vec{\beta}')$, we then need to consider two types of word distributions ϕ and ϕ' .

$$\begin{aligned} p(\vec{W} | \vec{Z}, \vec{Y}, \vec{\beta}, \vec{\beta}') &= \int \int p(\vec{W} | \vec{Z}, \vec{Y}, \vec{\phi}, \vec{\phi}') p(\vec{\phi} | \vec{\beta}) p(\vec{\phi}' | \vec{\beta}') d\vec{\phi}' d\vec{\phi} \\ &= \int \int \prod_{w=1}^V p_{w,y=0}^{n_{y=0}^w} \prod_{t=1}^T \prod_{w=1}^V p_{t,w,y=1}^{n_{t,w,y=1}^w} \frac{1}{\Delta(\vec{\beta})} \text{Dirichlet}(\vec{\beta}) \frac{1}{\Delta(\vec{\beta}')} \text{Dirichlet}(\vec{\beta}') d\vec{\phi}' d\vec{\phi} \\ &= \int \int \prod_{w=1}^V p_{w,y=0}^{n_{y=0}^w} \prod_{t=1}^T \prod_{w=1}^V p_{t,w,y=1}^{n_{t,w,y=1}^w} \frac{1}{\Delta(\vec{\beta})} \prod_{w=1}^V p_{t,w,y=1}^{\beta^w - 1} \frac{1}{\Delta(\vec{\beta}')} \prod_{w=1}^V p_{w,y=0}^{\beta'^w - 1} d\vec{\phi}' d\vec{\phi} \\ &= \int \int \frac{1}{\Delta(\vec{\beta})} \prod_{t=1}^T \prod_{w=1}^V p_{t,w,y=1}^{n_{t,w,y=1}^w + \beta^w - 1} \frac{1}{\Delta(\vec{\beta}')} \prod_{w=1}^V p_{w,y=0}^{n_{y=0}^w + \beta'^w - 1} d\vec{\phi}' d\vec{\phi} \\ &= \frac{\Delta(n_{y=0}^w + \beta')}{\Delta(\vec{\beta}')} \prod_{t=1}^T \frac{\Delta(n_{t,w,y=1}^w + \beta)}{\Delta(\vec{\beta})} \end{aligned}$$

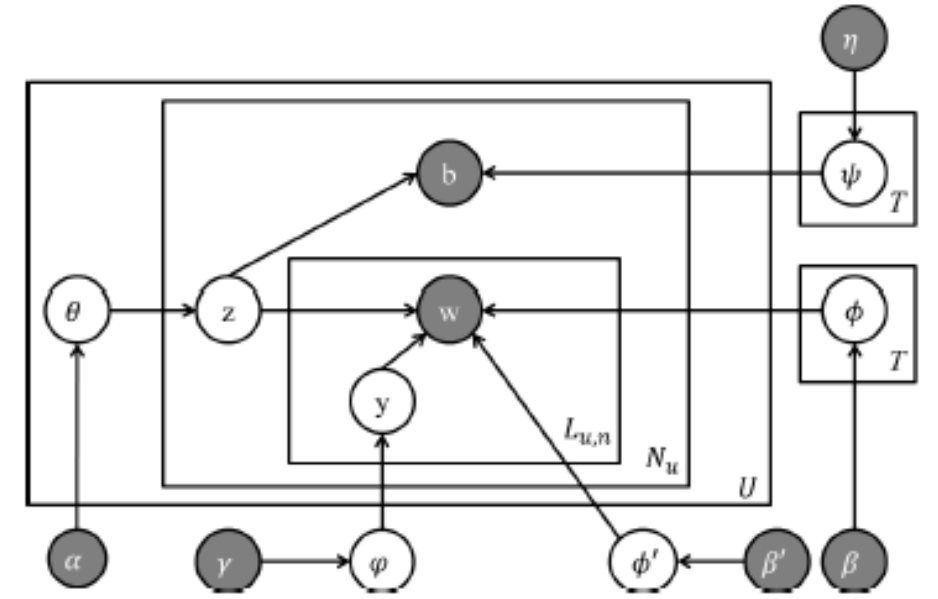


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因此：

$$p(\vec{Z}, \vec{W}, \vec{Y}, \vec{B} | \vec{\eta}, \vec{\beta}, \vec{\beta}', \vec{\gamma}, \vec{\alpha})$$

$$= p(\vec{Y} | \vec{\gamma}) p(\vec{W} | \vec{Z}, \vec{Y}, \vec{\beta}, \vec{\beta}') p(\vec{B} | \vec{Z}, \vec{\eta}) p(\vec{Z} | \vec{\alpha})$$

$$= \frac{\Delta(n_{(\cdot)}^y + \vec{\gamma})}{\Delta(\vec{\gamma})} \cdot \frac{\Delta(n_{y=0}^w + \beta')}{\Delta(\vec{\beta}')} \prod_{t=1}^T \frac{\Delta(n_{t,y=1}^w + \beta)}{\Delta(\vec{\beta})} \cdot \prod_{t=1}^T \frac{\Delta(\vec{n}_t^b + \vec{\eta})}{\Delta(\vec{\eta})} \cdot \prod_{u=1}^U \frac{\Delta(\vec{n}_u^t + \vec{\alpha})}{\Delta(\vec{\alpha})}$$

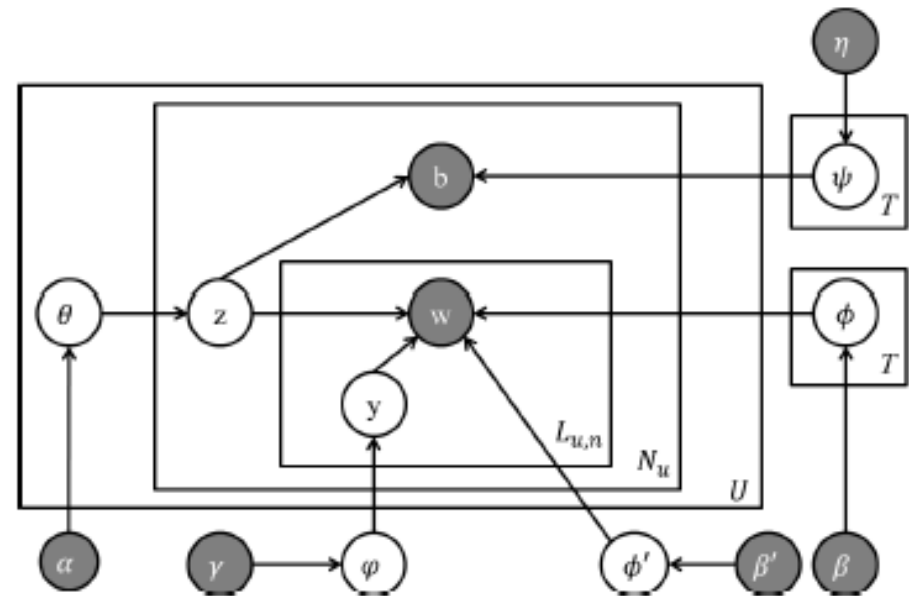


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let c denote $\{u, n\}$:

$$\begin{aligned}
p(\vec{Z}_c | \vec{Z}_{\neg c}, \vec{W}, \vec{Y}, \vec{B}) &= \frac{p(\vec{Z}, \vec{W}, \vec{Y}, \vec{B} | \vec{\eta}, \vec{\beta}, \vec{\beta}', \vec{\gamma}, \vec{\alpha})}{p(\vec{Z}_{\neg c}, \vec{W}, \vec{Y}, \vec{B} | \vec{\eta}, \vec{\beta}, \vec{\beta}', \vec{\gamma}, \vec{\alpha})} \\
&\propto \frac{p(\vec{Z}, \vec{W}, \vec{Y}, \vec{B} | \vec{\eta}, \vec{\beta}, \vec{\beta}', \vec{\gamma}, \vec{\alpha})}{p(\vec{Z}_{\neg c}, \vec{W}, \vec{Y}, \vec{B}_{\neg c} | \vec{\eta}, \vec{\beta}, \vec{\beta}', \vec{\gamma}, \vec{\alpha})} \\
&= \frac{\frac{\cancel{\Delta(n_{(\cdot)}^y + \vec{\gamma})}}{\cancel{\Delta(\vec{\gamma})}} \cdot \frac{\cancel{\Delta(n_{y=0}^w + \beta')}}{\cancel{\Delta(\vec{\beta}')}} \prod_{t=1}^T \frac{\Delta(n_{t,y=1}^w + \beta)}{\cancel{\Delta(\vec{\beta})}} \cdot \prod_{t=1}^T \frac{\Delta(\vec{n}_t^b + \vec{\eta})}{\cancel{\Delta(\vec{\eta})}} \cdot \prod_{u=1}^U \frac{\Delta(\vec{n}_u^t + \vec{\alpha})}{\cancel{\Delta(\vec{\alpha})}}}{\frac{\cancel{\Delta(n_{(\cdot)}^y + \vec{\gamma})}}{\cancel{\Delta(\vec{\gamma})}} \cdot \frac{\cancel{\Delta(n_{y=0}^w + \beta')}}{\cancel{\Delta(\vec{\beta}')}} \prod_{t=1}^T \frac{\Delta(n_{t,y=1,\neg c}^w + \beta)}{\cancel{\Delta(\vec{\beta})}} \cdot \prod_{t=1}^T \frac{\Delta(\vec{n}_{t,\neg c}^b + \vec{\eta})}{\cancel{\Delta(\vec{\eta})}} \cdot \prod_{u=1}^U \frac{\Delta(\vec{n}_{u,\neg c}^t + \vec{\alpha})}{\cancel{\Delta(\vec{\alpha})}}} \\
&= \prod_{t=1}^T \frac{\Delta(n_{t,y=1}^w + \vec{\beta})}{\Delta(n_{t,y=1,\neg c}^w + \vec{\beta})} \cdot \prod_{t=1}^T \frac{\Delta(\vec{n}_t^b + \vec{\eta})}{\Delta(\vec{n}_{t,\neg c}^b + \vec{\eta})} \cdot \prod_{u=1}^U \frac{\Delta(\vec{n}_u^t + \vec{\alpha})}{\Delta(\vec{n}_{u,\neg c}^t + \vec{\alpha})} \\
\Rightarrow p(\vec{Z}_c = z | \vec{Z}_{\neg c}, \vec{W}, \vec{Y}, \vec{B}) &= \frac{\Delta(n_{t,y=1}^w + \vec{\beta})}{\Delta(n_{t,y=1,\neg c}^w + \vec{\beta})} \cdot \frac{\Delta(\vec{n}_z^b + \vec{\eta})}{\Delta(\vec{n}_{z,\neg c}^b + \vec{\eta})} \cdot \frac{\Delta(\vec{n}_u^t + \vec{\alpha})}{\Delta(\vec{n}_{u,\neg c}^t + \vec{\alpha})}
\end{aligned}$$

$$\begin{aligned}
\frac{\Delta(n_{t,y=1}^w + \vec{\beta})}{\Delta(n_{t,y=1,\neg c}^w + \vec{\beta})} &= \frac{\frac{\prod_{w=1}^V \Gamma(n_{z,y=1}^w + \beta)}{\Gamma(\sum_{w=1}^V n_{z,y=1}^w + V\beta)}}{\frac{\prod_{w=1}^V \Gamma(n_{z,y=1,\neg c}^w + \beta)}{\Gamma(\sum_{w=1}^V n_{z,y=1,\neg c}^w + V\beta)}} \\
&= \frac{\prod_{w=1}^V \Gamma(n_{z,y=1}^w + \beta)}{\prod_{w=1}^V \Gamma(n_{z,y=1,\neg c}^w + \beta)} \cdot \frac{\Gamma(\sum_{w=1}^V n_{z,y=1,\neg c}^w + V\beta)}{\Gamma(\sum_{w=1}^V n_{z,y=1}^w + V\beta)} \quad (2)
\end{aligned}$$

$$\begin{aligned}
&= \frac{\prod_{w=1}^V \prod_{i=1}^{n_{c,y=1}^w} (n_{z,y=1,\neg c}^w + \beta + i - 1)}{\prod_{j=1}^{n_{c,y=1}^w} (\sum_{w=1}^V n_{z,y=1,\neg c}^w + V\beta + j - 1)} \quad (3)
\end{aligned}$$

注: (2)->(3):

$$\begin{aligned}
\frac{\prod_{w=1}^V \Gamma(n_{z,y=1}^w + \beta)}{\prod_{w=1}^V \Gamma(n_{z,y=1,\neg c}^w + \beta)} &= \frac{\Gamma(n_{z,y=1}^1 + \beta) \cdot \Gamma(n_{z,y=1}^2 + \beta) \cdots \Gamma(n_{z,y=1}^V + \beta)}{\Gamma(n_{z,y=1,\neg c}^1 + \beta) \cdot \Gamma(n_{z,y=1,\neg c}^2 + \beta) \cdots \Gamma(n_{z,y=1,\neg c}^V + \beta)} \\
&= \prod_{w=1}^V \prod_{i=1}^{n_{c,y=1}^w} (n_{z,y=1,\neg c}^w + \beta + i - 1)
\end{aligned}$$

where $n_{c,y=1}^w$ denotes the number of times word w occurs as topical words ;

$n_{c,y=1}^w$ is the total number of topical words in user u 's n - th tweets.

$$\begin{aligned}
\frac{\Delta(\vec{n}_z^b + \vec{\eta})}{\Delta(\vec{n}_{z,\neg c}^b + \vec{\eta})} &= \frac{\frac{\prod_{b=1}^B \Gamma(n_z^b + \eta)}{\Gamma(\sum_{b=1}^B n_z^b + B\eta)}}{\frac{\prod_{b=1}^B \Gamma(n_{z,\neg c}^b + \eta)}{\Gamma(\sum_{b=1}^B n_{z,\neg c}^b + B\eta)}} = \frac{\prod_{b=1}^B \Gamma(n_z^b + \eta)}{\prod_{b=1}^B \Gamma(n_{z,\neg c}^b + \eta)} \frac{\Gamma(\sum_{b=1}^B n_{z,\neg c}^b + B\eta)}{\Gamma(\sum_{b=1}^B n_z^b + B\eta)} \\
&= \frac{\Gamma(n_z^b + \eta)}{\Gamma(n_{z,\neg c}^b + \eta)} \frac{1}{\sum_{b=1}^B n_{z,\neg c}^b + B\eta} \\
&= \frac{n_{z,\neg c}^b + \eta}{\sum_{b=1}^B n_{z,\neg c}^b + B\eta}
\end{aligned}$$

同理： $\frac{\Delta(\vec{n}_u^t + \vec{\alpha})}{\Delta(\vec{n}_{u,\neg c}^t + \vec{\alpha})} = \frac{n_{u,\neg c}^Z + \alpha}{\sum_{t=1}^T n_{u,\neg c}^Z + T\alpha}$

where $n_{z,\neg c}^b$ denotes number of times topic z co-occurs with behavior b without considering the current tweet, $n_{u,\neg c}^Z$ denotes number of times topic z is sampled in user u 's tweets without considering the current tweet.

因此，

$$p(\vec{z}_c = z | \vec{Z}_{\neg c}, \vec{W}, \vec{Y}, \vec{B}) = \frac{\prod_{w=1}^V \prod_{i=1}^{n_{c,y=1}^w} (n_{z,y=1,\neg c}^w + \beta + i - 1)}{\prod_{j=1}^{n_{c,y=1}^{\textcolor{red}{w}}} (\sum_{w=1}^V n_{z,y=1,\neg c}^w + V\beta + j - 1)} \cdot \frac{n_{z,\neg c}^b + \eta}{\sum_{b=1}^B n_{z,\neg c}^b + B\eta} \cdot \frac{n_{u,\neg c}^z + \alpha}{\sum_{t=1}^T n_{u,\neg c}^z + T\alpha}$$

■ To sample label $y_{u,n,l}$:
 Let d be $\{u, n, l\}$,

$$p(y_d | \vec{Y}_{\neg d}, \vec{Z}, \vec{W}, \vec{B}) = \frac{p(\vec{Z}, \vec{W}, \vec{Y}, \vec{B} | \vec{\eta}, \vec{\beta}, \vec{\beta}', \vec{\gamma}, \vec{\alpha})}{p(\vec{Y}_{\neg d}, \vec{Z}, \vec{W}, \vec{B} | \vec{\eta}, \vec{\beta}, \vec{\beta}', \vec{\gamma}, \vec{\alpha})}$$

$$\propto \frac{p(\vec{Z}, \vec{W}, \vec{Y}, \vec{B} | \vec{\eta}, \vec{\beta}, \vec{\beta}', \vec{\gamma}, \vec{\alpha})}{p(\vec{Z}, \vec{W}_{\neg d}, \vec{Y}_{\neg d}, \vec{B} | \vec{\eta}, \vec{\beta}, \vec{\beta}', \vec{\gamma}, \vec{\alpha})}$$

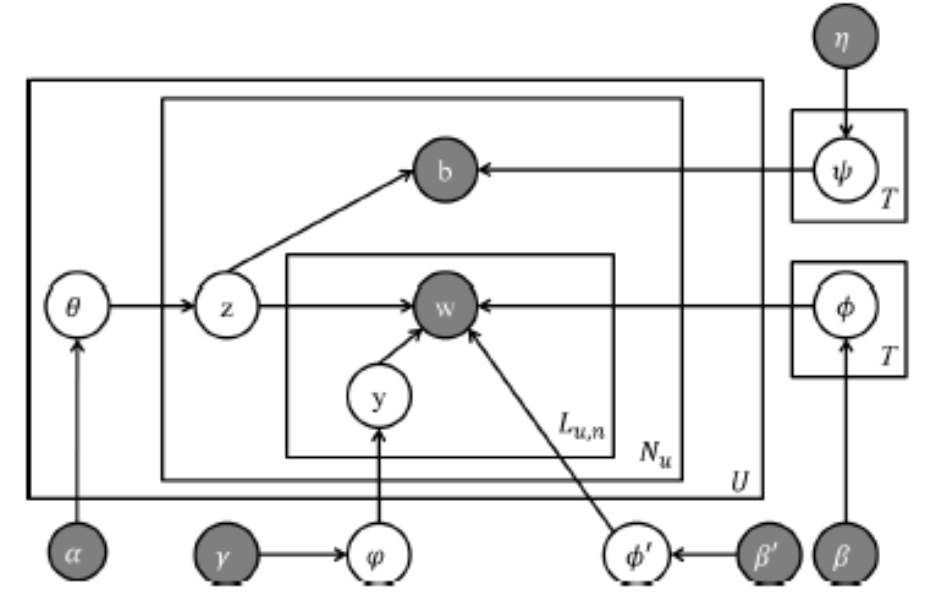


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$$\frac{\frac{\Delta(n_{(\cdot)}^y + \vec{\gamma})}{\cancel{\Delta(\vec{\gamma})}} \cdot \frac{\Delta(n_{y=0}^w + \vec{\beta}')}{\cancel{\Delta(\vec{\beta}')}} \prod_{t=1}^T \frac{\Delta(n_{t,y=1}^w + \vec{\beta})}{\cancel{\Delta(\vec{\beta})}} \cdot \prod_{t=1}^T \frac{\Delta(\vec{n}_t^b + \vec{\eta})}{\cancel{\Delta(\vec{\eta})}} \cdot \prod_{u=1}^U \frac{\Delta(\vec{n}_u^t + \vec{\alpha})}{\cancel{\Delta(\vec{\alpha})}}}{\frac{\Delta(n_{\neg d}^y + \vec{\gamma})}{\cancel{\Delta(\vec{\gamma})}} \cdot \frac{\Delta(n_{y=0,\neg d}^w + \vec{\beta}')}{\cancel{\Delta(\vec{\beta}')}} \prod_{t=1}^T \frac{\Delta(n_{t,y=1,\neg d}^w + \vec{\beta})}{\cancel{\Delta(\vec{\beta})}} \cdot \prod_{t=1}^T \frac{\Delta(\vec{n}_t^b + \vec{\eta})}{\cancel{\Delta(\vec{\eta})}} \cdot \prod_{u=1}^U \frac{\Delta(\vec{n}_u^t + \vec{\alpha})}{\cancel{\Delta(\vec{\alpha})}}}$$

$$= \frac{\Delta(n_{(\cdot)}^y + \vec{\gamma})}{\Delta(n_{\neg d}^y + \vec{\gamma})} \cdot \frac{\Delta(n_{y=0}^w + \vec{\beta}')}{\Delta(n_{y=0,\neg d}^w + \vec{\beta}')} \prod_{t=1}^T \frac{\Delta(n_{t,y=1}^w + \vec{\beta})}{\Delta(n_{t,y=1,\neg d}^w + \vec{\beta})}$$

$$p(y_d = 0 | \vec{Y}_{\neg d}, \vec{Z}, \vec{W}, \vec{B}) = \frac{\Delta(n_{(\cdot)}^y + \vec{\gamma})}{\Delta(n_{\neg d}^y + \vec{\gamma})} \cdot \frac{\Delta(n_{y=0}^w + \vec{\beta}')}{\Delta(n_{y=0, \neg d}^w + \vec{\beta}')}$$

$$\frac{\Delta(n_{(\cdot)}^y + \vec{\gamma})}{\Delta(n_{\neg d}^y + \vec{\gamma})} = \frac{\frac{\prod_{y=0}^1 \Gamma(n_{(\cdot)}^y + \gamma)}{\Gamma(\sum_{y=0}^1 n_{(\cdot)}^y + 2\gamma)}}{\frac{\prod_{y=0}^1 \Gamma(n_{\neg d}^y + \gamma)}{\Gamma(\sum_{y=0}^1 n_{\neg d}^y + 2\gamma)}}$$

$$= \frac{\prod_{y=0}^1 \Gamma(n_{(\cdot)}^y + \gamma)}{\prod_{y=0}^1 \Gamma(n_{\neg d}^y + \gamma)} \cdot \frac{1}{\sum_{y=0}^1 n_{\neg d}^y + 2\gamma}$$

$$= \frac{n_{\neg d}^{y_d=0} + \gamma}{\sum_{y=0}^1 n_{\neg d}^y + 2\gamma}$$

$$\Rightarrow p(y_d = 0 | \vec{Y}_{\neg d}, \vec{Z}, \vec{W}, \vec{B}) = \frac{n_{\neg d}^{y_d=0} + \gamma}{\sum_{y=0}^1 n_{\neg d}^y + 2\gamma} \cdot \frac{n_{y=0, \neg d}^{w_d} + \beta'}{\sum_{w=1}^V n_{y=0, \neg d}^w + V\beta'}$$

$$\frac{\Delta(n_{y=0}^w + \vec{\beta}')}{\Delta(n_{y=0, \neg d}^w + \vec{\beta}')} = \frac{\frac{\prod_{w=1}^V \Gamma(n_{y=0}^w + \beta')}{\Gamma(\sum_{w=1}^V n_{y=0}^w + V\beta')}}{\frac{\prod_{w=1}^V \Gamma(n_{y=0, \neg d}^w + \beta')}{\Gamma(\sum_{w=1}^V n_{y=0, \neg d}^w + V\beta')}} = \frac{n_{y=0, \neg d}^{w_d} + \beta'}{\sum_{w=1}^V n_{y=0, \neg d}^w + V\beta'}$$

$$\text{同理: } p(y_d = 1 | \vec{Y}_{\neg d}, \vec{Z}, \vec{W}, \vec{B}) = \frac{n_{\neg d}^{y_d=1} + \gamma}{\sum_{y=0}^1 n_{\neg d}^y + 2\gamma} \cdot \frac{n_{z_c, y=1, \neg d}^{w_d} + \beta}{\sum_{w=1}^V n_{z_c, y=1, \neg d}^w + V\beta}$$

参数估计:

$$\phi'_w = \frac{n_{y=0}^w + \beta'}{\sum_{w=1}^V n_{y=0}^w + V\beta'}$$

$$\phi_{t,w} = \frac{n_{t,y=1}^w + \beta}{\sum_{w=1}^V n_{t,y=1}^w + V\beta}$$

$$\psi_{t,b} = \frac{n_t^b + \eta}{\sum_{b=1}^B n_t^b + B\eta}$$

$$\varphi_y = \frac{n_{(.)}^y + \gamma}{\sum_{y=0}^1 n_{(.)}^y + 2\gamma}$$

$$\theta_{u,t} = \frac{n_u^t + \alpha}{\sum_{t=1}^T n_u^t + T\alpha}$$

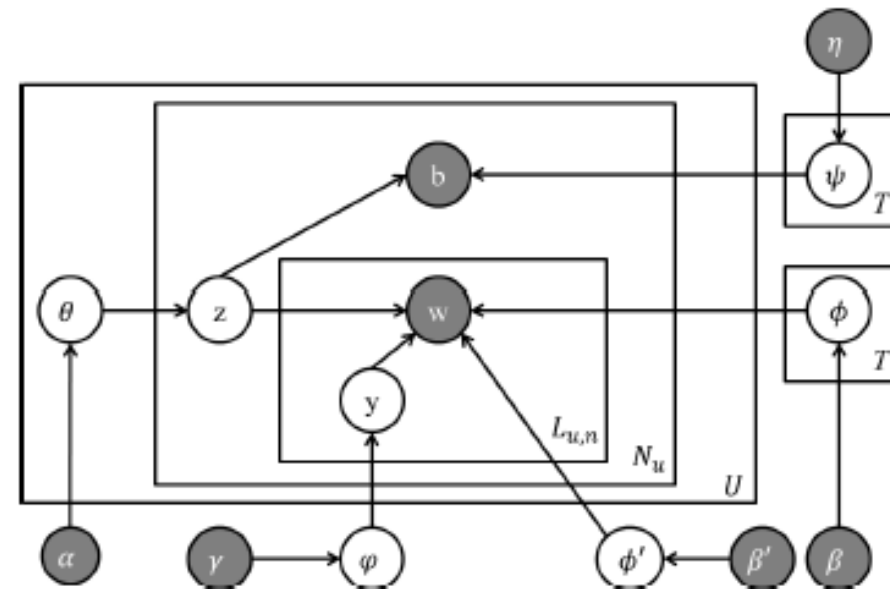


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where $n_{y=0}^w$ is the number of times w appears as background word, $n_{t,y=1}^w$ is the number of times w is sampled as topical word specific to topic t , n_t^b is number of time posting behavior b co-occurs with topic t , $n_{(.)}^y$ is number of times y appears, where $n_{\mathcal{U}}^t$ is, when given the user u , number of times t is sampled.

关于TOIS这篇：
关于这部分推导：

$$\begin{aligned}
& p(z_{u,s} = a | \mathbf{z}_{\neg\{u,s\}}, \mathbf{y}, \mathbf{l}, \Delta) \\
&= \frac{p(z_{u,s} = a, \mathbf{z}_{\neg\{u,s\}}, \mathbf{y}, \mathbf{l}, \Delta)}{p(\mathbf{z}_{\neg\{u,s\}}, \mathbf{y}, \mathbf{l}, \Delta)} \\
&= \frac{p(z_{u,s} = a, \mathbf{z}_{\neg\{u,s\}}, \mathbf{y}, \mathbf{l}, \Delta)}{p(\mathbf{z}_{\neg\{u,s\}}, \mathbf{y}_{\neg\{u,s\}}, \mathbf{l}_{\neg\{u,s\}}, \Delta_{\neg\{u,s\}})} \cdot \frac{1}{p(\mathbf{y}_{\{u,s\}}, \Delta_{\{u,s\}}, \mathbf{l}_{\{u,s\}} | \mathbf{z}_{\neg\{u,s\}}, \mathbf{y}_{\neg\{u,s\}}, \mathbf{l}_{\neg\{u,s\}}, \Delta_{\neg\{u,s\}})} \\
&\propto \frac{p(z_{u,s} = a, \mathbf{z}_{\neg\{u,s\}}, \mathbf{y}, \mathbf{l}, \Delta)}{p(\mathbf{z}_{\neg\{u,s\}}, \mathbf{y}_{\neg\{u,s\}}, \mathbf{l}_{\neg\{u,s\}}, \Delta_{\neg\{u,s\}})} \tag{1} \\
&= \frac{p(z_{u,s} = a, \mathbf{z}_{\neg\{u,s\}})}{p(\mathbf{z}_{\neg\{u,s\}})} \\
&\quad \times \frac{p(\mathbf{y} | z_{u,s} = a, \mathbf{z}_{\neg\{u,s\}})}{p(\mathbf{y}_{\neg\{u,s\}} | \mathbf{z}_{\neg\{u,s\}})} \\
&\quad \times \frac{p(\mathbf{l} | z_{u,s} = a, \mathbf{z}_{\neg\{u,s\}}, \mathbf{y})}{p(\mathbf{l}_{\neg\{u,s\}} | \mathbf{z}_{\neg\{u,s\}}, \mathbf{y}_{\neg\{u,s\}})} \\
&\quad \times \frac{p(\Delta | z_{u,s} = a, \mathbf{z}_{\neg\{u,s\}}, \mathbf{y})}{p(\Delta_{\neg\{u,s\}} | \mathbf{z}_{\neg\{u,s\}}, \mathbf{y}_{\neg\{u,s\}})}, \tag{2}
\end{aligned}$$

我自己的想法是：

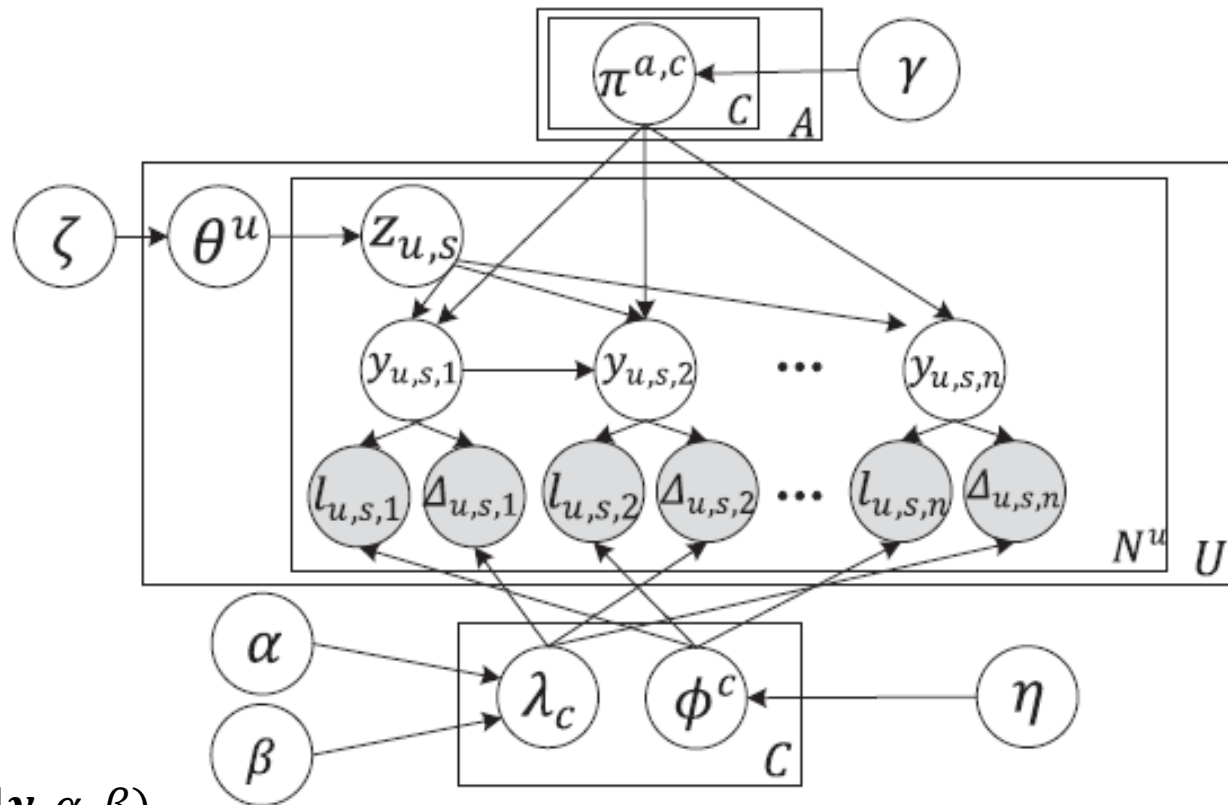
$$p(z_{u,s} = a | \mathbf{z}_{\neg\{u,s\}}, \mathbf{y}, \mathbf{l}, \Delta)$$

$$= \frac{p(\mathbf{z}, \mathbf{y}, \mathbf{l}, \Delta | \zeta, \alpha, \beta, \eta, \gamma)}{p(\mathbf{z}_{\neg\{u,s\}}, \mathbf{y}, \mathbf{l}, \Delta | \zeta, \alpha, \beta, \eta, \gamma)}$$

$$\propto \frac{p(\mathbf{z}, \mathbf{y}, \mathbf{l}, \Delta | \zeta, \alpha, \beta, \eta, \gamma)}{p(\mathbf{z}_{\neg\{u,s\}}, \mathbf{y}_{\neg\{u,s\}}, \mathbf{l}_{\neg\{u,s\}}, \Delta_{\neg\{u,s\}} | \zeta, \alpha, \beta, \eta, \gamma)}$$

$$p(\mathbf{z}, \mathbf{y}, \mathbf{l}, \Delta | \zeta, \alpha, \beta, \eta, \gamma) = p(\mathbf{z} | \zeta) p(\mathbf{y} | \mathbf{z}, \gamma) p(\mathbf{l} | \mathbf{y}, \eta) p(\Delta | \mathbf{y}, \alpha, \beta)$$

注：关于这个式子，我自己根据盘子图是这样想的，也不知道对不对



$$P(z|\xi) = \int P(z|\theta) P(\theta|\xi) d\theta$$

$$= \int \prod_{u=1}^U \prod_{a=1}^A P_{u,a}^{N_a^u} \cdot \text{Dirichlet}(\xi) d\theta$$

$$= \int \prod_{u=1}^U \prod_{a=1}^A P_{u,a}^{N_a^u + \xi_a - 1} \cdot \frac{1}{\phi(\xi)} d\theta$$

$$P(z_{us}=a|z_{\setminus us}) = \frac{\prod_{u=1}^U \frac{\phi(N_a^u + \xi)}{\phi(\xi)}}{\prod_{u=1}^U \phi(\xi)}$$

$$\frac{P(z)}{P(z_{us})} = \frac{P(z|\xi)}{P(z_{us}|\xi)} = \frac{\prod_{u=1}^U \frac{\phi(N_a^u + \xi)}{\phi(\xi)}}{\frac{\prod_{u=1}^U \phi(N_{a,us}^u + \xi)}{\phi(\xi)}} = \frac{\phi(N_a^u + \xi)}{\phi(N_{a,us}^u + \xi)}$$

$$\frac{\phi(N_a^u + \xi)}{\phi(N_{a,us}^u + \xi)} = \frac{\prod_{a=1}^A T(N_a^u + \xi)}{T(\sum_{a=1}^A N_a^u + A\xi)} = \frac{N_{a,us}^u + \xi}{\sum_{a=1}^A N_{a,us}^u + A\xi} \propto \frac{N_a^u + \xi}{N_{(\cdot)}^u + A\xi}$$

$$\frac{p(\mathbf{y}|z_{u,s} = a, \mathbf{z}_{\neg\{u,s\}})}{p(\mathbf{y}_{\neg\{u,s\}}|\mathbf{z}_{\neg\{u,s\}})} \propto \prod_{c_1=0}^C \left\{ \frac{\Gamma(N_{(\cdot)}^{a,c_1} + C\gamma)}{\Gamma(N_{(\cdot)}^{a,c_1} + N_{(\cdot)}^{u,s,c_1} + C\gamma)} \prod_{c_2=1}^C \frac{\Gamma(N_{c_2}^{a,c_1} + N_{c_2}^{c_1} + \gamma)}{\Gamma(N_{c_2}^{t,c_1} + \gamma)} \right\}, \quad (2)$$

$$\frac{p(\mathbf{l}|z_{u,s} = a, \mathbf{z}_{\neg\{u,s\}}, \mathbf{y})}{p(\mathbf{l}_{\neg\{u,s\}}|\mathbf{z}_{\neg\{u,s\}}, \mathbf{y}_{\neg\{u,s\}})} \propto \prod_{c=1}^C \left\{ \frac{\Gamma(N_{(\cdot)}^c + L\eta)}{\Gamma(N_{(\cdot)}^c + N_{(\cdot)}^{u,s,c} + L\eta)} \prod_{l=1}^L \frac{\Gamma(N_l^c + N_l^{u,s,c} + \eta)}{\Gamma(N_l^c + \eta)} \right\}, \quad (3)$$

$$\frac{p(\Delta|z_{u,s} = a, \mathbf{z}_{\neg\{u,s\}}, \mathbf{y})}{p(\Delta_{\neg\{u,s\}}|\mathbf{z}_{\neg\{u,s\}}, \mathbf{y}_{\neg\{u,s\}})} \propto \prod_{c=1}^C \lambda_c^{N_{(\cdot)}^{u,s,c}} e^{-\lambda_c N^{u,s} \overline{\Delta^{u,s}}}. \quad (4)$$

但是这三个式子我尝试了很多遍还是没推导出来，主要感觉奇怪的地方是：比如（2）式，按照基本LDA，Twitter-LDA，B-LDA的思路，我觉得不应该再出现gamma函数了，因为就像第一个式子那种应该可以约掉的。还有就是 $(z_{u,s} = a, \mathbf{z}_{\neg\{u,s\}})$ 应该就是 \mathbf{z} 吧？根据盘子图， \mathbf{l} 在给定 \mathbf{y} 的时候我认为就和 \mathbf{z} 无关了，因此第（3）个式子中， $p(\mathbf{l}|\mathbf{z}, \mathbf{y})$ 中 \mathbf{z} 应该可以忽略，就为 $p(\mathbf{l}|\mathbf{y})$ 。第（4）个也有相同的疑惑。