# $Text\ Mining-Assignment\ \#3$

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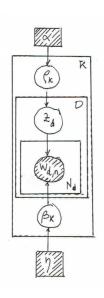
# Exercise 1

Text.

# Exercise 2

#### Part (a)

The directed graph is the following:



# Part (b)

The Markov blankets of these elemets of the model can be expressed as follows:

- Words in document d: topic assignments  $z_d$  (parent) and topics  $\beta_k$  (parent).
- Topic assignment  $z_d$ : topic probabilities  $\rho_k$  (parent), the set of words  $w_{d,n}$  (children) and topics  $\beta_k$  (children's parent).
- Topics  $\beta_k$ : hyperparameter  $\eta$  (parent), the set of words  $w_{d,n}$  (children) and topic assignment  $z_d$  (children's parent).

# Part (c)

An uncollapsed Gibbs algorithm could be the following:

- 1. Set values for  $\eta$ ,  $\alpha \in \mathbb{R}^K$
- 2. Draw for each topic  $k \in \{1, ..., K\}$  a sample  $\beta_k \sim \text{Dir}(\eta) \in \Delta^{V-1}$
- 3. Draw a sample  $\rho \sim \text{Dir}(\alpha) \in \Delta^{K-1}$  that specifies the likelihood of each topic
- 4. Draw for each document  $d \in \{1, ..., D\}$  a sample  $z_d \sim \text{multinom}(\rho)$
- 5. Draw for each word  $n \in \{1, ..., N_d\}$  in document d the word  $w_{d,n} \sim \text{multinom}\left(\boldsymbol{\beta}_{z_d}\right)$
- 6. Update for each k the vector  $\boldsymbol{\beta}_k \sim \mathrm{Dir}(\boldsymbol{\eta} + \mathbf{m}_k) \in \Delta^{V-1}$ , where element v of vector  $\mathbf{m}_k \in \mathbb{R}^V$  is  $m_{k,v}$ , the number of times topic k generates word v.
- 7. Update the vector  $\rho \sim \text{Dir}(\alpha + \delta) \in \Delta^{K-1}$ , where element k of vector  $\delta \in \mathbb{R}^K$  is  $\delta_k$ , the number of documents that are assigned topic k.
- 8. Return to step 4 and repeat until convergence.