
Text Mining – Assignment #3

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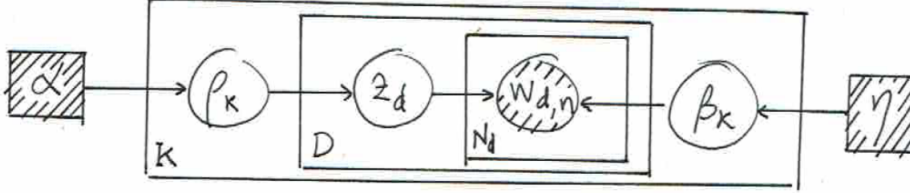
Exercise 1

Text.

Exercise 2

Part (a)

The directed graph is the following:



Part (b)

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

- Words in document d : topic assignments z_d (parent) and topics β_k (parent).
- Topic assignment z_d : topic probabilities ρ_k (parent), the set of words $w_{d,n}$ (children) and topics β_k (children's parent).
- Topics β_k : hyperparameter η (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 $\eta \in \mathbb{R}^V$ and $\alpha \in \mathbb{R}^K$
2. Draw for each topic $k \in \{1, \dots, K\}$ 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 $d \in \{1, \dots, D\}$ a sample $z_d \sim \text{multinom}(\rho)$
5. Draw for each word $n \in \{1, \dots, N_d\}$ in document d the word $w_{d,n} \sim \text{multinom}(\beta_{z_d})$
6. Update for each k the vector $\beta_k \sim \text{Dir}(\eta + \mathbf{m}_k) \in \Delta^{V-1}$, 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 δ_k , the number of documents that are assigned topic k .
8. Return to step 4 and repeat until convergence.