

Cluster LDA

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1 Standard Generative Process

For this reduced case of IPTM, we assume that the document-cluster assignments $c_d \in \{1, \dots, C\}$ for $d \in [D]$ are known and fixed. Following the cluster-based topic model, document d has the document-topic distribution

$$\boldsymbol{\theta}_d \sim \text{Dirichlet}(\alpha, \boldsymbol{\xi}_{c_d}), \quad (1)$$

where α are the concentration parameter and $\boldsymbol{\xi}_{c_d}$ is the base measure corresponding to the cluster of document d . In order to capture the overall prevalence of each topic in the corpus, we assume that each $\boldsymbol{\xi}_c$ is given Dirichlet priors with a single corpus-level base measure \mathbf{m}

$$\boldsymbol{\xi}_c \sim \text{Dirichlet}(\alpha_1, \mathbf{m}), \quad (2)$$

where α_1 is the concentration parameter determining the extent to which the group-specific base measures are affected by the corpus-level base measure. Finally, the corpus-level base measure is assumed to have Dirichlet prior with uniform base measure

$$\mathbf{m} \sim \text{Dirichlet}\left(\alpha_0, \left(\frac{1}{K}, \dots, \frac{1}{K}\right)\right). \quad (3)$$

Given that $\bar{N}_d = \max(1, N_d)$ where N_d is known, a topic z_{dn} is drawn from the document-topic distribution for each $n \in [\bar{N}_d]$ —i.e.,

$$z_{dn} \sim \text{Multinomial}(\boldsymbol{\theta}_d). \quad (4)$$

2 Current Derivatoin

The predictive probability

$$\begin{aligned}
\Pr(z_{N_d+1} = k | \mathbf{z}, \alpha, \boldsymbol{\xi}_{c_d}) &= \int_{\boldsymbol{\theta}_d} \Pr(k | \boldsymbol{\theta}_d) \times \Pr(\boldsymbol{\theta}_d | \mathbf{z}, \alpha, \boldsymbol{\xi}_{c_d}) d\boldsymbol{\theta}_d \\
&= \int_{\boldsymbol{\theta}_d} \prod_{k=1}^K (\theta_{dk}) \times \left(\prod_{n=1}^{N_d} \Pr(z_{dn} | \boldsymbol{\theta}_d) \times \Pr(\boldsymbol{\theta}_d | \alpha, \boldsymbol{\xi}_{c_d}) \right) d\boldsymbol{\theta}_d \\
&= \int_{\boldsymbol{\theta}_d} \prod_{k=1}^K (\theta_{dk}) \times \prod_{k=1}^K (\theta_{dk})^{N_{dk}} \times \frac{\Gamma(\sum_{k=1}^K \alpha \xi_{c_d k})}{\prod_{k=1}^K \Gamma(\alpha \xi_{c_d k})} \prod_{k=1}^K (\theta_{dk})^{(\alpha \xi_{c_d k}) - 1} d\boldsymbol{\theta}_d \\
&\propto \int_{\boldsymbol{\theta}_d} \prod_{k=1}^K (\theta_{dk}) \times \text{Dir}(N_{dk} + \alpha \xi_{c_d k}) d\boldsymbol{\theta}_d \quad (\text{Expectation!}) \\
&= \frac{N_{dk} + \alpha \xi_{c_d k}}{N_d + \alpha}.
\end{aligned} \tag{5}$$

Then, we move to the next cluster-level hierarchy:

$$\begin{aligned}
\Pr(z_{N_d+1} = k | \mathbf{z}, \alpha, \alpha_1, \mathbf{m}) &= \int_{\boldsymbol{\xi}_{c_d}} \Pr(z_{N_d+1} = k | \mathbf{z}, \alpha, \boldsymbol{\xi}_{c_d}) \times \Pr(\boldsymbol{\xi}_{c_d} | \mathbf{z}, \alpha_1, \mathbf{m}) d\boldsymbol{\xi}_{c_d} \\
&= \int_{\boldsymbol{\xi}_{c_d}} \prod_{k=1}^K \left(\frac{N_{dk} + \alpha \xi_{c_d k}}{N_d + \alpha} \right) \times \left(\prod_{d': c'_d = c_d} \prod_{n=1}^{N_d} \Pr(z_{dn} | \boldsymbol{\xi}_{c_d}) \times \Pr(\boldsymbol{\xi}_{c_d} | \alpha_1, \mathbf{m}) \right) d\boldsymbol{\xi}_{c_d} \\
&= \frac{N_{dk}}{N_d + \alpha} + \frac{\alpha}{N_d + \alpha} \int_{\boldsymbol{\xi}_{c_d}} \prod_{k=1}^K (\xi_{c_d k}) \times \prod_{k=1}^K (\xi_{c_d k})^{N_{c_d k}} \times \frac{\Gamma(\sum_{k=1}^K \alpha_1 m_k)}{\prod_{k=1}^K \Gamma(\alpha_1 m_k)} \prod_{k=1}^K (\xi_{c_d k})^{(\alpha_1 m_k) - 1} d\boldsymbol{\xi}_{c_d} \\
&\propto \frac{N_{dk}}{N_d + \alpha} + \frac{\alpha}{N_d + \alpha} \int_{\boldsymbol{\xi}_{c_d}} \prod_{k=1}^K (\xi_{c_d k}) \times \text{Dir}(N_{c_d k} + \alpha_1 m_k) d\boldsymbol{\xi}_{c_d} \quad (\text{Expectation!}) \\
&= \frac{N_{dk} + \alpha \frac{N_{c_d k} + \alpha_1 m_k}{N_{c_d} + \alpha_1}}{N_d + \alpha}.
\end{aligned} \tag{6}$$

NOTE: I am not sure if we can directly use $\Pr(z_{dn} | \boldsymbol{\xi}_{c_d})$ as Multinomial when we move from the first line to the second one.

Finally, we move to the next corpus-level hierarchy:

$$\begin{aligned}
& \Pr(z_{N_d+1} = k | \mathbf{z}, \alpha, \alpha_1, \alpha_0) \\
&= \int_{\mathbf{m}} \Pr(z_{N_d+1} = k | \mathbf{z}, \alpha, \alpha_1, \mathbf{m}) \times \Pr(\mathbf{m} | \mathbf{z}, \alpha_0, \mathbf{u}) d\mathbf{m} \\
&= \int_{\mathbf{m}} \prod_{k=1}^K \left(\frac{N_{dk} + \alpha \frac{N_{c_d k} + \alpha_1 m_k}{N_{c_d} + \alpha_1}}{N_d + \alpha} \right) \times \left(\prod_{d=1}^D \prod_{n=1}^{N_d} \Pr(z_{dn} | \mathbf{m}) \times \Pr(\mathbf{m} | \alpha_0, \mathbf{u}) \right) d\mathbf{m} \\
&= \frac{N_{dk}}{N_d + \alpha} + \frac{\alpha}{N_d + \alpha} \times \\
&\left(\frac{N_{c_d k}}{N_{c_d} + \alpha_1} + \frac{\alpha_1}{N_{c_d} + \alpha_1} \int_{\mathbf{m}} \prod_{k=1}^K (m_k) \times \prod_{k=1}^K (m_k)^{N_k} \times \frac{\Gamma(\sum_{k=1}^K \alpha_0 / K)}{\prod_{k=1}^K \Gamma(\alpha_0 / K)} \prod_{k=1}^K (m_k)^{(\alpha_0 / K) - 1} d\mathbf{m} \right) \\
&\propto \frac{N_{dk}}{N_d + \alpha} + \frac{\alpha}{N_d + \alpha} \times \\
&\left(\frac{N_{c_d k}}{N_{c_d} + \alpha_1} + \frac{\alpha_1}{N_{c_d} + \alpha_1} \int_{\mathbf{m}} \prod_{k=1}^K (m_k) \times \text{Dir}(N_k + \alpha_0 / K) d\mathbf{m} \right) \quad (\text{Expectation!}) \\
&= \frac{N_{dk}}{N_d + \alpha} + \frac{\alpha}{N_d + \alpha} \left(\frac{N_{c_d k}}{N_{c_d} + \alpha_1} + \frac{\alpha_1}{N_{c_d} + \alpha_1} \frac{N_k + \alpha_0 / K}{N + \alpha_0} \right) \\
&= \frac{N_{dk} + \alpha \frac{N_{c_d k} + \alpha_1 \frac{N_k + \alpha_0 / K}{N + \alpha_0}}{N_{c_d} + \alpha_1}}{N_d + \alpha}.
\end{aligned} \tag{7}$$

NOTE: Again, I am not sure if we can directly use $\Pr(z_{dn} | \mathbf{m})$ as Multinomial when we move from the first line to the second one.

3 Integrating out Θ, Ξ, \mathbf{m}

The big joint distribution is:

$$\begin{aligned}
& \prod_{d=1}^D \prod_{n=1}^{N_d} \Pr(z_{dn} | \boldsymbol{\theta}_d) \times \prod_{d=1}^D \Pr(\boldsymbol{\theta}_d | \alpha, \boldsymbol{\xi}_{c_d}) \times \prod_{c=1}^C \Pr(\boldsymbol{\xi}_c | \alpha_1, \mathbf{m}) \times \Pr(\mathbf{m} | \alpha_0) \\
&= \prod_{d=1}^D \prod_{n=1}^{N_d} \text{Multinom}(z_{dn} | \boldsymbol{\theta}_d) \times \prod_{d=1}^D \text{Dir}(\boldsymbol{\theta}_d | \alpha, \boldsymbol{\xi}_{c_d}) \times \prod_{c=1}^C \text{Dir}(\boldsymbol{\xi}_c | \alpha_1, \mathbf{m}) \times \text{Dir}(\mathbf{m} | \alpha_0).
\end{aligned} \tag{8}$$

We want to integrate out $\boldsymbol{\theta}$, $\boldsymbol{\xi}$, and \mathbf{m} :

$$\begin{aligned}
& \int_{\mathbf{m}} \int_{\Xi} \int_{\Theta} \prod_{d=1}^D \prod_{n=1}^{N_d} \Pr(z_{dn} | \boldsymbol{\theta}_d) \times \prod_{d=1}^D \Pr(\boldsymbol{\theta}_d | \alpha, \boldsymbol{\xi}_{c_d}) \times \prod_{c=1}^C \Pr(\boldsymbol{\xi}_c | \alpha_1, \mathbf{m}) \times \Pr(\mathbf{m} | \alpha_0) d\Theta d\Xi dM \\
&= \int_{\mathbf{m}} \int_{\Xi} \prod_{c=1}^C \Pr(\boldsymbol{\xi}_c | \alpha_1, \mathbf{m}) \times \Pr(\mathbf{m} | \alpha_0) \times \left(\int_{\Theta} \prod_{d=1}^D \prod_{n=1}^{N_d} \Pr(z_{dn} | \boldsymbol{\theta}_d) \times \prod_{d=1}^D \Pr(\boldsymbol{\theta}_d | \alpha, \boldsymbol{\xi}_{c_d}) d\Theta \right) d\Xi dM.
\end{aligned} \tag{9}$$

First, we can work on the integration of $\boldsymbol{\theta}$ as below:

$$\begin{aligned}
& \int_{\Theta} \prod_{d=1}^D \prod_{n=1}^{N_d} \Pr(z_{dn} | \boldsymbol{\theta}_d) \times \prod_{d=1}^D \Pr(\boldsymbol{\theta}_d | \alpha, \boldsymbol{\xi}_{c_d}) d\Theta \\
&= \prod_{d=1}^D \int_{\boldsymbol{\theta}_d} \prod_{n=1}^{N_d} \text{Multinom}(z_{dn} | \boldsymbol{\theta}_d) \times \text{Dir}(\boldsymbol{\theta}_d | \alpha, \boldsymbol{\xi}_{c_d}) d\boldsymbol{\theta}_d \\
&= \prod_{d=1}^D \int_{\boldsymbol{\theta}_d} \prod_{k=1}^K (\theta_{dk})^{N_{dk}} \times \frac{\Gamma(\sum_{k=1}^K \alpha \xi_{c_d k})}{\prod_{k=1}^K \Gamma(\alpha \xi_{c_d k})} \prod_{k=1}^K (\theta_{dk})^{(\alpha \xi_{c_d k}) - 1} d\boldsymbol{\theta}_d \\
&= \prod_{d=1}^D \frac{\Gamma(\sum_{k=1}^K \alpha \xi_{c_d k})}{\prod_{k=1}^K \Gamma(\alpha \xi_{c_d k})} \int_{\boldsymbol{\theta}_d} \prod_{k=1}^K (\theta_{dk})^{(\alpha \xi_{c_d k}) + N_{dk} - 1} d\boldsymbol{\theta}_d \\
&= \prod_{d=1}^D \frac{\Gamma(\sum_{k=1}^K \alpha \xi_{c_d k})}{\prod_{k=1}^K \Gamma(\alpha \xi_{c_d k})} \frac{\prod_{k=1}^K \Gamma(\alpha \xi_{c_d k} + N_{dk})}{\Gamma(\sum_{k=1}^K \alpha \xi_{c_d k} + N_{dk})} \int_{\boldsymbol{\theta}_d} \frac{\Gamma(\sum_{k=1}^K \alpha \xi_{c_d k} + N_{dk})}{\prod_{k=1}^K \Gamma(\alpha \xi_{c_d k} + N_{dk})} \prod_{k=1}^K (\theta_{dk})^{(\alpha \xi_{c_d k}) + N_{dk} - 1} d\boldsymbol{\theta}_d \\
&= \prod_{d=1}^D \frac{\Gamma(\sum_{k=1}^K \alpha \xi_{c_d k})}{\Gamma(\sum_{k=1}^K \alpha \xi_{c_d k} + N_{dk})} \frac{\prod_{k=1}^K \Gamma(\alpha \xi_{c_d k} + N_{dk})}{\prod_{k=1}^K \Gamma(\alpha \xi_{c_d k})} \\
&= \prod_{d=1}^D \frac{\Gamma(\alpha)}{\Gamma(\alpha + N_d)} \frac{\prod_{k=1}^K \Gamma(\alpha \xi_{c_d k} + N_{dk})}{\prod_{k=1}^K \Gamma(\alpha \xi_{c_d k})},
\end{aligned} \tag{10}$$

where N_{dk} is the number of times topic k is assigned in document d . Since $\frac{\Gamma(\alpha)}{\Gamma(\alpha + N_d)}$ is only a function of hyperparameters, we drop the term and then

Equation (6) can be re-written as:

$$\begin{aligned}
& \int_{\mathbf{m}} \int_{\Xi} \prod_{c=1}^C \Pr(\boldsymbol{\xi}_c | \alpha_1, \mathbf{m}) \times \Pr(\mathbf{m} | \alpha_0) \times \left(\int_{\Theta} \prod_{d=1}^D \prod_{n=1}^{N_d} \Pr(z_{dn} | \boldsymbol{\theta}_d) \times \prod_{d=1}^D \Pr(\boldsymbol{\theta}_d | \alpha, \boldsymbol{\xi}_{c_d}) d\Theta \right) d\Xi dM \\
& \propto \int_{\mathbf{m}} \int_{\Xi} \prod_{c=1}^C \Pr(\boldsymbol{\xi}_c | \alpha_1, \mathbf{m}) \times \Pr(\mathbf{m} | \alpha_0) \times \prod_{d=1}^D \prod_{k=1}^K \frac{\Gamma(\alpha \xi_{cdk} + N_{dk})}{\Gamma(\alpha \xi_{cdk})} d\Xi dM \\
& = \int_{\mathbf{m}} \Pr(\mathbf{m} | \alpha_0) \left(\int_{\Xi} \prod_{c=1}^C \Pr(\boldsymbol{\xi}_c | \alpha_1, \mathbf{m}) \times \prod_{d=1}^D \prod_{k=1}^K \frac{\Gamma(\alpha \xi_{cdk} + N_{dk})}{\Gamma(\alpha \xi_{cdk})} d\Xi \right) dM.
\end{aligned} \tag{11}$$

Then, we can work on the integration of Ξ as below:

$$\begin{aligned}
& \int_{\Xi} \prod_{c=1}^C \Pr(\boldsymbol{\xi}_c | \alpha_1, \mathbf{m}) \times \prod_{d=1}^D \prod_{k=1}^K \frac{\Gamma(\alpha \xi_{cdk} + N_{dk})}{\Gamma(\alpha \xi_{cdk})} d\Xi \\
& = \prod_{c=1}^C \int_{\xi_c} \text{Dir}(\boldsymbol{\xi}_c | \alpha_1, \mathbf{m}) \times \prod_{d:c_d=c} \prod_{k=1}^K \frac{\Gamma(\alpha \xi_{cdk} + N_{dk})}{\Gamma(\alpha \xi_{cdk})} d\xi_c \\
& = \prod_{c=1}^C \int_{\xi_c} \frac{\Gamma(\sum_{k=1}^K \alpha_1 m_k)}{\prod_{k=1}^K \Gamma(\alpha_1 m_k)} \prod_{k=1}^K \left((\xi_{ck})^{(\alpha_1 m_k) - 1} \times \prod_{d:c_d=c} \frac{\Gamma(\alpha \xi_{cdk} + N_{dk})}{\Gamma(\alpha \xi_{cdk})} \right) d\xi_c \\
& = \prod_{c=1}^C \frac{\Gamma(\sum_{k=1}^K \alpha_1 m_k)}{\prod_{k=1}^K \Gamma(\alpha_1 m_k)} \int_{\xi_c} \prod_{k=1}^K \left((\xi_{ck})^{(\alpha_1 m_k) - 1} \times \prod_{d:c_d=c} \frac{\Gamma(\alpha \xi_{cdk} + N_{dk})}{\Gamma(\alpha \xi_{cdk})} \right) d\xi_c.
\end{aligned} \tag{12}$$

Here, we need to know from Equation (36) of Hierarchical Dirichlet Process (<http://qwone.com/~jason/trg/papers/teh-hierarchical-04.pdf>) that

$$\begin{aligned}
\frac{\Gamma(\alpha \xi_{cdk} + N_{dk})}{\Gamma(\alpha \xi_{cdk})} & = \prod_{i_{dk}=1}^{N_{dk}} (i_{dk} - 1 + \alpha \xi_{cdk}) \\
& = \sum_{i_{dk}=0}^{N_{dk}} s(N_{dk}, i_{dk}) (\alpha \xi_{cdk})^{i_{dk}},
\end{aligned} \tag{13}$$

where $s(i_{dk}, N_{dk})$ is the coefficient of $(\alpha\xi_{ck})^{i_{dk}}$, which are unsigned Stirling numbers of the first kind. Plugging this into Equation (9), we have

$$\prod_{c=1}^C \frac{\Gamma(\sum_{k=1}^K \alpha_1 m_k)}{\prod_{k=1}^K \Gamma(\alpha_1 m_k)} \int_{\xi_c} \prod_{k=1}^K \left((\xi_{ck})^{(\alpha_1 m_k) - 1} \times \prod_{d:c_d=c} \sum_{i_{dk}=0}^{N_{dk}} s(N_{dk}, i_{dk}) (\alpha\xi_{ck})^{i_{dk}} \right) d\xi_c. \quad (14)$$

Maybe we need to introduce auxiliary variable $\mathbf{i} = (i_{dk} : \forall d, k)$ and move forward, but I have no idea....