Filtered Latent Dirichlet Allocation: Variational Bayes Algorithm

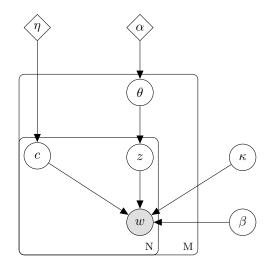
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Filtered latent Dirichlet allocation is a natural extension of latent Dirichlet allocation that organically captures and filters out corpus-specific stop-words which are not otherwise caught by generic stop-word lists.

Probabilistic Graphical Model:

θ_d	$\sim \text{Dirichlet}(\alpha)$
c_{d_n}	$\sim \operatorname{Binomial}(\eta)$
z_{d_n}	$\sim \text{Categorical}(\theta_d)$
w_{d_n}	$\sim \text{Categorical}(\kappa)$ if $c_{d_n} = 0$
w_{d_n}	$\sim \text{Categorical}(\beta_{z_{d_n}}) \text{ if } c_{d_n} = 1$



The marginal log-likelihood for this PGM, in expected value form, is:

$$\log p(w|\alpha,\beta,\kappa,\eta) \geq \sum_{d=1}^{M} \left[\mathbf{E}_{q}[\log p(w_{d},z_{d},\theta_{d}|\alpha,\beta,\kappa,\eta)] - \mathbf{E}_{q}[\log q(z_{d},c_{d},\theta_{d}|\phi_{d},\tau_{d},\gamma_{d})] \right]$$

$$= \sum_{d=1}^{M} \left[\operatorname{E}_{q}[\log p(w_{d}|z_{d}, c_{d}, \beta, \kappa)] + \operatorname{E}_{q}[\log p(z_{d}|\theta_{d})] + \operatorname{E}_{q}[\log p(c_{d}|\eta)] + \operatorname{E}_{q}[\log p(\theta_{d}|\alpha)] - \operatorname{E}_{q}[\log q(z_{d}|\phi_{d})] - \operatorname{E}_{q}[\log q(c_{d}|\tau_{d})] - \operatorname{E}_{q}[\log q(\theta_{d}|\gamma_{d})] \right]$$

In particular:

$$\begin{split} \bullet \quad \sum_{d=1}^{M} \mathbf{E}_{q}[\log p(w_{d}|z_{d},c_{d},\beta,\kappa)] &= \sum_{d=1}^{M} \sum_{n=1}^{N_{d}} \mathbf{E}_{q}[\log p(w_{d_{n}}|z_{d_{n}},c_{d_{n}},\beta,\kappa)] \\ &= \sum_{d=1}^{M} \sum_{n=1}^{N_{d}} \mathbf{E}_{q}[\log(\beta_{z_{d_{n}},w_{d_{n}}}^{c_{d_{n}}} \cdot \kappa_{w_{d_{n}}}^{1-c_{d_{n}}})] \\ &= \sum_{d=1}^{M} \sum_{n=1}^{N_{d}} \left[\mathbf{E}_{q}[c_{d_{n}}] \cdot \mathbf{E}_{q}[\log \beta_{z_{d_{n}},w_{d_{n}}}] + \mathbf{E}_{q}[1-c_{d_{n}}] \log \kappa_{w_{d_{n}}} \right] \\ &= \sum_{d=1}^{M} \sum_{n=1}^{N_{d}} \left[\tau_{d_{n}}(-\log \kappa_{w_{d_{n}}} + \sum_{i=1}^{K} \phi_{d_{in}} \log \beta_{iw_{d_{n}}}) + \log \kappa_{w_{d_{n}}} \right] \end{split}$$

•
$$\sum_{d=1}^{M} \mathbb{E}_{q}[\log p(c_{d}|\eta)] = \sum_{d=1}^{M} \left[N_{d} \log(1-\eta) + \log(\frac{\eta}{1-\eta}) \sum_{n=1}^{N_{d}} \tau_{n} \right]$$

•
$$\sum_{d=1}^{M} E_q[\log q(c_d|\tau_d)] = \sum_{d=1}^{M} \sum_{n=1}^{N_d} \left[\tau_{d_n} \log \tau_{d_n} + (1 - \tau_{d_n}) \log(1 - \tau_{d_n}) \right]$$

The remaining expected values are identical to those found in Blei's paper, Latent Dirichlet Allocation (2003).

The relevant update equations are as follows:

$$\bullet \quad \eta = \frac{\sum_{d=1}^{M} \sum_{n=1}^{N_d}}{\sum_{d=1}^{M} N_d}$$

$$\bullet \quad \kappa_j \propto \sum_{d=1}^M \sum_{n=1}^{N_d} (1 - \tau_{d_n}) w_{d_n}^j$$

$$\bullet \quad \tau_{d_n} = \frac{\eta}{\eta + (1 - \eta)\kappa_{w_{d_n}} \prod_{i=1}^K \beta_{iw_{d_n}}^{-\phi_{d_{in}}}}$$

•
$$\phi_{d_{in}} \propto \beta_{iw_{d_n}}^{\tau_{d_n}} \exp\left(\psi(\gamma_i) - \psi(\sum_{l=1}^K \gamma_l)\right)$$

$$\bullet \quad \beta_{ij} \propto \sum_{d=1}^{M} \sum_{n=1}^{N_d} \tau_{d_n} \phi_{d_{in}} w_{d_n}^j$$