

# Bonus Homework 5

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## Problem description

In **Probabilistic Latent Semantic Analysis**, we want to calculate parameters  $\theta, \pi$  so that

$$\text{maximum } l(\theta, \pi; N) = \sum_{d,w} n(d, w) \log \left( \sum_{z=1}^K P(w|z; \theta) P(z|d; \pi) \right)$$

Apply **Expectation Maximization** to estimate  $\theta, \pi$ .

## Solution

The **EM** method is an iterative method and can be divided into 2 steps.

Actually we want to estimate two matrices  $[N, K]$  and  $[K, M]$ , denote them as  $P(z_k|d_i)$  and  $P(w_j|z_k)$ . At very beginning, assume we have known these parameters (use random value).

**E-step:** Use the current results to calculate the posterior probability of latent variables.

$$P(z_k|d_i, w_j) = \frac{P(w_j|z_k) P(z_k|d_i)}{\sum_{l=1}^K P(w_j|z_l) P(z_l|d_i)}$$

**M-step:**

1. We use the posterior in the previous step to "modify" the expectation:

$$\begin{aligned} E(d_i, w_j, z_k) &= \sum_{i=1}^N \sum_{j=1}^M n(d_i, w_j) \sum_{k=1}^K P(z_k|d_i, w_j) \log[P(w_j|z_k) P(z_k|d_i)] \\ \text{s.t. } &\sum_{j=1}^M p(w_j|z_k) = 1, \sum_{k=1}^K p(z_k|d_i) = 1 \end{aligned}$$

2. This function has  $N \times K + K \times M$  variables. Use Lagrange Multiplier Approach:

$$\mathcal{H} = E(d_i, w_j, z_k) + \sum_{k=1}^K \tau_k \left( 1 - \sum_{j=1}^M P(w_j|z_k) \right) + \sum_{i=1}^N \rho_i \left( 1 - \sum_{k=1}^K P(z_k|d_i) \right)$$

3. Derivative them, get the equations:

$$\begin{aligned} \sum_{i=1}^N n(d_i, w_j) P(z_k|d_i, w_j) - \alpha_k P(w_j|z_k) &= 0 \\ \sum_{j=1}^M n(d_i, w_j) P(z_k|d_i, w_j) - \beta_i P(z_k|d_i) &= 0 \end{aligned}$$

4. Finally we can acquire the new value of parameters:

$$\begin{aligned} P(w_j|z_k) &= \frac{\sum_{i=1}^N n(d_i, w_j) P(z_k|d_i, w_j)}{\sum_{m=1}^M \sum_{i=1}^N n(d_i, w_m) P(z_k|d_i, w_m)} \\ P(z_k|d_i) &= \frac{\sum_{j=1}^M n(d_i, w_j) P(z_k|d_i, w_j)}{n(d_i)} \end{aligned}$$

Repeat the above 2 steps until results converge.