
Algorithm 1: Latent Dirichlet Allocation using Variational Expectation-Maximization

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input : A corpus collection of  $D$  documents
output:  $\gamma$  ( $D \times K$  matrix),  $\lambda$  ( $K \times V$  matrix),  $\phi$  ( $D \times N \times K$ )
1 initialize parameters;
2 Expectation-step: Optimize the variational parameters ( $\gamma_d, \phi_d \in D$ ),  $\lambda$ ,
  and likelihood;
3 while the ELBO has not converged do
4   loglikelihood := 0;
5   for  $d = 1$  to  $D$  do
6     repeat
7       for  $n = 1$  to  $N$  do
8         for  $k = 1$  to  $K$  do
9            $\phi_{dnk}^{t+1} := \exp(\Psi(\gamma_{dk}) + \Psi(\lambda_{k,n^j}) - \Psi(\sum_i \lambda_{k,n^j}))$ ;
10          end
11          normalize  $\phi_{dnk}^{t+1}$  to sum 1;
12        end
13         $\gamma^{t+1} := \alpha + \sum_{n=1}^N \phi_{dn}$ 
14      until converge of  $\phi_d$  and  $\gamma_d$ ;
15    end
16    for  $k = 1$  to  $K$  do
17      for  $j = 1$  to  $V$  do
18         $\lambda_{kj} := \eta + \sum_{d=1}^D \sum_{n=1}^N \phi_{dnk} w_{dn}^j$ ;
19      end
20      normalize  $\lambda_k$  to sum 1;
21    end
22    loglikelihood := loglikelihood +  $L(\lambda, \phi, \gamma; \alpha, \eta)$  ;
23  end
24  Maximization Step for lower bound on the likelihood;
25  for  $d = 1$  to  $D$  do
26    for  $k = 1$  to  $K$  do
27      for  $j = 1$  to  $V$  do
28        update  $\eta$  ;
29      end
30    end
31    update  $\alpha$  ;
32  end
33  if Likelihood converged then
34    return  $\gamma, \lambda, \phi$ ;
35  else
36    repeat E-M algorithm;
37  end

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