Appendix

May 10, 2013

Derivation of Variational Spatial LDA

Variational Distribution

$$q(\phi, \pi, d, z | \lambda, \gamma, \eta, \varphi) = \prod_{k=1}^{K} q(\phi_k | \lambda) \prod_{m=1}^{M} q(\pi_m | \gamma) \prod_{n=1}^{N} q(d_n | \eta) q(z_n | \varphi)$$

$$(1)$$

where

 $\phi_k \in \mathbb{R}^V$ distribution over word of topic k

 $\pi_m \in \mathbb{R}^K$ distribution over topics of document m

 $d_n \in \mathbb{R}^M$ indicator vector where if $d_{nm} = 1$ means word n belongs to document m

 $z_n \in \mathbb{R}^K$ indicator vector where if $z_{nk} = 1$ means word n is assigned to topic k

Lower bound of log likelihood:

$$p(\boldsymbol{w}, \boldsymbol{c} | \alpha, \beta, \eta, \sigma) = \log \int_{\phi} \int_{\pi} \sum_{z} \sum_{z} p(\phi, \pi, d, z, \boldsymbol{w}, \boldsymbol{c} | \alpha, \beta, \eta, \sigma)$$
(2)

$$= \log \int_{\phi} \int_{\pi} \sum_{z} \sum_{z} q(\phi, \pi, d, z | \lambda, \gamma, \eta, \varphi) \frac{p(\phi, \pi, d, z, z, w | \alpha, \beta, \eta, \sigma)}{q(\phi, \pi, d, z | \lambda, \gamma, \eta, \varphi)} d\phi d\pi dd dz$$
 (3)

$$= \log E_q \frac{p(\phi, \pi, d, z, \boldsymbol{c}, \boldsymbol{w} | \alpha, \beta, \eta, \sigma)}{q(\phi, \pi, d, z | \lambda, \gamma, \eta, \varphi)}$$

$$\tag{4}$$

$$\geq E_q \log p(\phi, \pi, d, z, c, w | \alpha, \beta, \eta, \sigma) - E_q \log q(\phi, \pi, d, z | \lambda, \gamma, \eta, \varphi) = L$$
 (5)

where $E_q \equiv E_{q(\phi,\pi,d,z|\lambda,\gamma,\eta,\varphi)}$ Instead of maximizing p, we estimate it by maximizing the lower bound L

Consider the first term in L

$$E_q[\log p(\phi, \pi, d, z, \boldsymbol{c}, \boldsymbol{w} | \alpha, \beta, \eta, \sigma)] = E_q[\log p(\phi | \beta) p(\pi | \alpha) p(d | \eta) p(\boldsymbol{c} | c_d^{d}, \sigma) p(z | \pi_d) p(\boldsymbol{w} | \phi_d)$$
(6)

$$= E_q[\log p(\phi|\beta)] + E_q[p(\pi|\alpha)] + E_q[\log p(d|\eta)]$$
(7)

+
$$E_a[\log p(\boldsymbol{c}|\boldsymbol{c}_d^{\mathrm{d}}, \sigma)] + E_a[\log p(\boldsymbol{z}|\pi_d)] + E_a[\log p(\boldsymbol{w}|\phi_z)]$$
 (8)

(9)

$$E_q[\log p(\phi|\beta)] = E_q \log \prod_{k=1}^K p(\phi_k|\beta)$$
(10)

$$= \sum_{k=1}^{K} E_q[\log \Gamma(\sum_{v=1}^{V} \beta_v) - \sum_{v=1}^{V} \log \Gamma(\beta_v) + \sum_{v=1}^{V} (\beta_v - 1) \log(\phi_{kv})]$$
(11)

$$= \sum_{k=1}^{K} \log \Gamma(\sum_{v=1}^{V} \beta_{v}) - \sum_{v=1}^{V} \log \Gamma(\beta_{v}) + \sum_{v=1}^{V} (\beta_{v} - 1)(\Psi(\lambda_{kv}) - \Psi(\sum_{v'-1}^{V} \lambda_{kv'}))$$
(12)

$$E_q[\log p(\pi|\alpha)] = E_q[\log \prod_{m=1}^{M} p(\pi_m|\alpha)]$$
(13)

$$= \sum_{m=1}^{M} E_q \log(p(\pi_m | \alpha)) \tag{14}$$

$$= \sum_{m=1}^{M} E_q[\log(\Gamma(\sum_{k=1}^{K} \alpha_k)) - \sum_{k=1}^{K} \log(\Gamma(\alpha_k)) + \sum_{k=1}^{K} (\alpha_k - 1) \log(\pi_m k)]$$
(15)

$$= \sum_{m=1}^{M} \log(\Gamma(\sum_{k=1}^{K} \alpha_k)) - \sum_{k=1}^{K} \log(\Gamma(\alpha_k)) + \sum_{k=1}^{K} (\alpha_k - 1)(\Psi(\gamma_{mk}) - \Psi(\sum_{k'=1}^{K} \gamma_{mk'}))$$
(16)

Speed up things a little bit:

$$E_q[\log p(d|\eta)] = \sum_{n=1}^{N} E_q \log(p(d_n|\eta))$$
(17)

$$= \sum_{n=1}^{N} E_q \sum_{m=1}^{M} d_n m \log(\eta)$$
 (18)

$$= \sum_{n=1}^{N} \sum_{m=1}^{M} \eta \log(\eta)$$
 (19)

$$= (N+M)\eta \log(\eta) \tag{20}$$

$$E_q[\log p(\mathbf{c}|c_d^{\mathrm{d}}, \sigma)] = \sum_{n=1}^{N} E_q \log p(c_n|c_{d_n}^{\mathrm{d}})$$
(21)

$$= \sum_{n=1}^{N} \sum_{m=1}^{M} E_q[d_{mn} \log p(c_n | c_{d_{mn}}^{d})]$$
 (22)

$$= \sum_{n=1}^{N} \sum_{m=1}^{M} \eta(\delta_{g_{d_{mn}}^{d}} - \frac{(x_{d_{mn}}^{d} - x_{n})^{2} + (y_{d_{mn}}^{d} - y_{n})^{2}}{\sigma^{2}})$$
 (23)

(24)

$$E_q[\log p(z|\pi_d)] = \sum_{n=1}^{N} \sum_{m=1}^{M} E_q[d_{mm} \log(\pi_{mn})]$$
 (25)

$$= \sum_{n=1}^{N} \sum_{m=1}^{M} \eta(\Psi(\gamma_{mn}) - \Psi(\sum_{m'=1}^{M} \gamma_{m'n})$$
 (26)

$$E_q[\log p(\boldsymbol{w}|\phi_z)] = \sum_{n=1}^{N} \sum_{k=1}^{M} \sum_{k=1}^{K} \sum_{v=1}^{V} E_q[d_{mn}z_{kn}\boldsymbol{w_n^v}log(\phi_k v)]$$
(27)

$$= \sum_{n=1}^{N} \sum_{m=1}^{M} \sum_{k=1}^{K} \eta \phi_{kn} \boldsymbol{w}_{\boldsymbol{n}}^{\boldsymbol{v}} (\Psi(\lambda_{kv}) - \Psi(\sum_{k'}^{K} \lambda_{k'v})$$
(28)

(29)

Now, consider the second term in L, this is to use the log of the joint distribution under variational parameters in (1)

$$E_q \log q(\phi|\lambda) = \sum_{k}^{K} E_q \log q(\phi_k|\lambda)$$
(30)

$$= \sum_{k=1}^{K} \log \Gamma(\sum_{v=1}^{V} \lambda_{kv}) - \sum_{v=1}^{V} \log(\Gamma(\lambda_{kv})) + \sum_{v=1}^{V} (\lambda_{kv} - 1) E_q[\log(\phi_{kv})]$$
(31)

$$= \sum_{k}^{K} \log \Gamma(\sum_{v}^{V} \lambda_{kv}) - \sum_{v}^{V} \log(\Gamma(\lambda_{kv})) + \sum_{v}^{V} (\lambda_{kv} - 1)(\Psi(\lambda_{kv}) - \Psi(\sum_{v'}^{V} \lambda_{kv'})$$
(32)

(33)

$$E_q \log q(\pi|\gamma) = \sum_m^M E_q \log q(\pi_m|\gamma)$$
(34)

$$= \sum_{m}^{M} \log \Gamma(\sum_{k}^{K} \gamma_{km}) - \sum_{k}^{K} \log \Gamma(\gamma_{km}) + \sum_{k}^{K} (\gamma_{km}) E_{q}[\log(\pi_{km})]$$
(35)

$$= \sum_{m}^{M} \log \Gamma(\sum_{k}^{K} \gamma_{km}) - \sum_{k}^{K} \log \Gamma(\gamma_{km}) + \sum_{k}^{K} (\gamma_{km}) (\Psi(\gamma_{km}) - \Psi(\sum_{k'}^{K} \gamma_{km}))$$
(36)

(37)

$$E_q \log q(d_n|\eta) = \sum_{n=1}^{N} E_q \log p(d_n|\eta)$$
(38)

$$= \sum_{n}^{N} \sum_{m}^{M} E_q d_{mn} \log(\eta) \tag{39}$$

$$= \sum_{n=1}^{N} \sum_{m=1}^{M} E_q[d_{mn}\log(\eta)]$$

$$\tag{40}$$

$$= \sum_{n}^{N} \sum_{m}^{M} \eta \log(\eta)$$
 (41)

$$= (M+N)\eta\log(\eta) \tag{42}$$

(43)

$$E_q \log q(z_n|\eta) = \sum_{n=1}^{N} E_q \log q(z_n|\varphi)$$
(44)

$$= \sum_{n=1}^{N} \sum_{k=1}^{K} E_q z_{kn} \log(\varphi_{kn}) \tag{45}$$

$$= \sum_{n}^{N} \sum_{k}^{K} \varphi_{kn} (\Psi(\varphi_{kn}) - \Psi(\sum_{k'}^{K} \varphi_{k'n})$$

$$\tag{46}$$

(47)

Putting it all together:

$$L = E_q[\log p(\phi|\beta)] + E_q[p(\pi|\alpha)] + E_q[\log p(d|\eta)] + E_q[\log p(\mathbf{c}|\mathbf{c}_d^d, \sigma)] + E_q[\log p(\mathbf{z}|\pi_d)] + E_q[\log p(\mathbf{w}|\phi_z)]$$
(48)

$$+ E_q \log q(\phi|\lambda) + E_q \log q(\pi|\gamma) + E_q \log q(d_n|\eta) + E_q \log q(z_n|\eta)$$

$$\tag{49}$$

$$= \sum_{k=1}^{K} \log \Gamma(\sum_{v=1}^{V} \beta_v) - \sum_{v=1}^{V} \log \Gamma(\beta_v) + \sum_{v=1}^{V} (\beta_v - 1)(\Psi(\lambda_{kv}) - \Psi(\sum_{v'=1}^{V} \lambda_{kv'}))$$
(50)

$$+ \sum_{m=1}^{M} \log(\Gamma(\sum_{k=1}^{K} \alpha_{k})) - \sum_{k=1}^{K} \log(\Gamma(\alpha_{k})) + \sum_{k=1}^{K} (\alpha_{k} - 1)(\Psi(\gamma_{mk}) - \Psi(\sum_{k'=1}^{K} \gamma_{mk'}))$$
(51)

$$+ (N+M)\eta \log(\eta) \tag{52}$$

$$+ \sum_{n=1}^{N} \sum_{m=1}^{M} \eta \left(\delta_{g_{d_{mn}}^{d}} - \frac{(x_{d_{mn}}^{d} - x_{n})^{2} + (y_{d_{mn}}^{d} - y_{n})^{2}}{\sigma^{2}} \right)$$
 (53)

$$+ \sum_{n=1}^{N} \sum_{m=1}^{M} \eta(\Psi(\gamma_{mn}) - \Psi(\sum_{m'=1}^{M} \gamma_{m'n})$$
 (54)

$$+ \sum_{n}^{N} \sum_{m}^{M} \sum_{k}^{K} \eta \phi_{kn} \boldsymbol{w}_{n}^{v} (\Psi(\lambda_{kv}) - \Psi(\sum_{k'}^{K} \lambda_{k'v})$$

$$(55)$$

$$+ \sum_{k}^{K} \log \Gamma(\sum_{v}^{V} \lambda_{kv}) - \sum_{v}^{V} \log(\Gamma(\lambda_{kv})) + \sum_{v}^{V} (\lambda_{kv} - 1)(\Psi(\lambda_{kv}) - \Psi(\sum_{v'}^{V} \lambda_{kv'})$$

$$(56)$$

$$+ \sum_{m}^{M} \log \Gamma(\sum_{k}^{K} \gamma_{km}) - \sum_{k}^{K} \log \Gamma(\gamma_{km}) + \sum_{k}^{K} (\gamma_{km}) (\Psi(\gamma_{km}) - \Psi(\sum_{k'}^{K} \gamma_{km}))$$

$$(57)$$

$$+ (M+N)\eta \log(\eta) \tag{58}$$

$$+ \sum_{n=1}^{N} \sum_{k=1}^{K} \varphi_{kn} (\Psi(\varphi_{kn}) - \Psi(\sum_{k'}^{K} \varphi_{k'n})$$

$$(59)$$

(60)

Maximize with respect to each term:

$$L_{\lambda} = \sum_{v=1}^{V} (\beta_v - 1)(\Psi(\lambda_{kv}) - \Psi(\sum_{v'=1}^{V} \lambda_{kv'}))$$
(61)

$$+ \sum_{n}^{N} \sum_{m}^{M} \sum_{k}^{K} \eta \phi_{kn} \boldsymbol{w}_{\boldsymbol{n}}^{\boldsymbol{v}} (\Psi(\lambda_{kv}) - \Psi(\sum_{k'}^{K} \lambda_{k'v})$$

$$(62)$$

$$+ \sum_{k}^{K} \log \Gamma(\sum_{v}^{V} \lambda_{kv}) - \sum_{v}^{V} \log(\Gamma(\lambda_{kv})) + \sum_{v}^{V} (\lambda_{kv} - 1)(\Psi(\lambda_{kv}) - \Psi(\sum_{v'}^{V} \lambda_{kv'})$$

$$(63)$$