

Modeling Dialogue Acts with Content Word Filtering and Speaker Preferences (Supplementary Material)

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1 Derivation for Gibbs Sampling

Sampling s For the current utterance u , refer to its parent utterance as p , its child utterances as $\mathbf{c} = \{c_1, \dots, c_m\}$, and its sentences as $\mathbf{e} = \{e_1, \dots, e_n\}$. The speaker of an utterance is denoted by a_u with its subscript being the utterance, e.g., a_u . For brevity, \mathbf{c}_i denotes $\{c_i, \dots, c_m\}$, and \mathbf{e}_i denotes $\{e_i, \dots, e_n\}$. \mathbf{a} denotes the author indices of all utterances. \mathbf{s} denotes the utterance-level DA indices of all utterances, and \mathbf{s}_c denotes the utterance-level DA indices of \mathbf{c} . \mathbf{z}^F denotes the sentence-level DA indices of all sentences in the data, and \mathbf{z}_e^F denotes the sentence-level DA indices of \mathbf{e} . A subscript starting with \setminus means exclusion, e.g., $\mathbf{s}_{\setminus u, c}$ means the utterance-level DA indices of all utterances but u and \mathbf{c} .

As we sample, N_{ij}^{SS} is the frequency of the utterances whose parent’s DA is i and whose own DA is j . N_{ij}^{AS} is the frequency of utterances whose speaker is i and DA is j . N_{ij}^{SF} is the frequency of sentences whose utterance has utterance-level DA i and whose own sentence-level DA is j . A superscript with “\utterances/sentences” (e.g., $N_{ij}^{SS\setminus u}$) means to exclude those utterances or sentences from counting.

$$P(s_u = j | \mathbf{a}, \mathbf{s}_{\setminus u}, \mathbf{z}^F, \nu, \gamma^S, \gamma^A, \alpha^F) \quad (1)$$

$$\begin{aligned} &\propto P(s_u = j, \mathbf{s}_{\mathbf{c}}, \mathbf{z}_{\mathbf{e}}^F | \mathbf{a}, \mathbf{s}_{\setminus u, \mathbf{c}}, \mathbf{z}_{\setminus \mathbf{e}}^F, \nu, \gamma^S, \gamma^A, \alpha^F) \\ &= P(s_u = j | \mathbf{a}, \mathbf{s}_{\setminus u, \mathbf{c}}, \mathbf{z}_{\setminus \mathbf{e}}^F, \nu, \gamma^S, \gamma^A) \end{aligned} \quad (2)$$

$$\prod_{i=1}^m P(s_{c_i} | \mathbf{a}, \mathbf{s}_{\setminus \mathbf{c}_i}, \mathbf{z}_{\setminus \mathbf{e}}^F, \nu, \gamma^S, \gamma^A) \quad (3)$$

$$\prod_{i=1}^n P(z_{e_i}^F | \mathbf{a}, \mathbf{z}_{\setminus \mathbf{e}_i}^F, \alpha^F). \quad (4)$$

(2) is approximated as

$$\nu \frac{N_{spj}^{SS \setminus u, \mathbf{c}} + \gamma^S}{\sum_k (N_{spk}^{SS \setminus u, \mathbf{c}} + \gamma^S)} + (1 - \nu) \frac{N_{auj}^{AS \setminus u} + \gamma^A}{\sum_k (N_{auk}^{AS \setminus u} + \gamma^A)}.$$

(3) is approximated as

$$\prod_{i=1}^m \left(\nu \frac{N_{js_{c_i}}^{SS \setminus \mathbf{c}_i} + \gamma^S}{\sum_k (N_{jk}^{SS \setminus \mathbf{c}_i} + \gamma^S)} + (1 - \nu) \frac{N_{a_{c_i} s_{c_i}}^{AS} + \gamma^A}{\sum_k (N_{a_{c_i} k}^{AS} + \gamma^A)} \right).$$

(4) is approximated as

$$\prod_{i=1}^n \frac{N_{jz_{e_i}^F}^{SF \setminus \mathbf{e}_i} + \alpha^F}{\sum_k (N_{jk}^{SF \setminus \mathbf{e}_i} + \alpha^F)}.$$

Sampling the sentence-level DA z_e^F of sentence e : Let ω denote all words in the data and ω_e the words in sentence e . \mathbf{w} and \mathbf{w}_e are the word indices of ω and ω_e , respectively, where $\mathbf{w}_{e>j} = \{w_{ej}, \dots, w_{e|\omega_e|}\}$. \mathbf{l} is the DA/content indicators of all words. \mathbf{z}^F is the sentence-level DAs of all sentences in the same utterance.

As we sample, N_{ij}^{FW} is the frequency of the words whose l is “DA” and word index is j . As before, a superscript with “\ words” means to exclude those words from counting.

$$\begin{aligned}
P(z_e^F = k | \mathbf{s}, \mathbf{z}_{\setminus e}^F, \mathbf{w}, \mathbf{l}, \mathbf{b}, \alpha^F, \beta) \\
\propto P(z_e^F = k, \mathbf{w}_e | \mathbf{s}, \mathbf{z}_{\setminus e}^F, \mathbf{w}_{\setminus e}, \mathbf{l}, \mathbf{b}, \alpha^F, \beta) \\
\propto P(z_e^F = k | \mathbf{s}, \mathbf{z}_{\setminus e}^F, \alpha^F)
\end{aligned} \tag{5}$$

$$\prod_{\substack{j=1 \\ l_j = \text{DA}}}^{|\omega_e|} P(w_{ej} | \mathbf{z}^F, \mathbf{l}, \mathbf{w} \setminus \mathbf{w}_{e>j}, \beta). \tag{6}$$

(5) is approximated as

$$\frac{N_{s_u k}^{SF \setminus e} + \alpha^F}{\sum_{k'} N_{s_u k'}^{SF \setminus e} + \alpha^F}.$$

(6) is approximated as

$$\prod_{\substack{j=1 \\ l_j = \text{DA}}}^{|\omega_e|} \frac{N_{k w_{ej}}^{FW \setminus \omega_{e>j}} + \beta}{\sum_{w'} N_{k w'}^{FW \setminus \omega_{e>j}} + \beta}.$$

Sampling the background topic z_d^B of discussion d : ω_d is all words in d and \mathbf{w}_d is their word indices, where $\mathbf{w}_{d>j} = \{w_{dj}, \dots, w_{d|\omega_d|}\}$. Now \mathbf{l} denotes all indicators in d .

As we sample, N_j^B is the frequency of the discussions whose background topic is j . As before, a superscript with “ $\setminus d$ ” means to exclude d from counting.

$$\begin{aligned}
P(z_d^B = k | z_{\setminus d}^B, \mathbf{w}, \mathbf{l}, \alpha^B, \beta) \\
\propto P(z_d^B = k, \mathbf{w}_d | z_{\setminus d}^B, \mathbf{w}_{\setminus d}, \mathbf{l}, \alpha^B, \beta) \\
\propto P(z_d^B = k | z_{\setminus d}^B, \alpha^B)
\end{aligned} \tag{7}$$

$$\prod_{\substack{j=1 \\ l_j = \text{content}}}^{|\omega_d|} P(w_{dj} | \mathbf{z}^B, \mathbf{l}, \mathbf{w} \setminus \mathbf{w}_{d>j}, \beta). \tag{8}$$

(7) is approximated as

$$\frac{N_k^{B \setminus d} + \alpha^B}{\sum_{k'} N_k^{B \setminus d} + \alpha^B}.$$

(8) is approximated as

$$\prod_{\substack{j=1 \\ l_j=\text{content}}}^{|\omega_d|} \frac{N_{kw_{dj}}^{BW \setminus \omega_{d>j}} + \beta}{\sum_{w'} N_{kw'}^{BW \setminus \omega_{d>j}} + \beta}.$$

Sampling l of word ω : Let ω be in sentence e and have word index w .

$$\begin{aligned} P(l = \text{DA} | \mathbf{l}_{\setminus \omega}, \mathbf{w}, \mathbf{z}^F, \eta, \beta) &\propto P(l = \text{DA}, w | \mathbf{l}_{\setminus \omega}, \mathbf{w}_{\setminus \omega}, \mathbf{z}^F, \eta, \beta) \\ &= P(l = \text{DA} | \eta) P(w | \mathbf{w}_{\setminus \omega}, \mathbf{z}^F, \mathbf{1}, \beta) \\ &\approx \eta \frac{N_{z_e^F w}^{FW \setminus \omega} + \beta}{\sum_{w'} N_{z_e^F w'}^{FW \setminus \omega} + \beta}. \end{aligned}$$

Similarly,

$$\begin{aligned} P(l = \text{content} | \mathbf{l}_{\setminus \omega}, \mathbf{w}, \mathbf{z}^B, \eta, \beta) &\propto P(l = \text{content}, w | \mathbf{l}_{\setminus \omega}, \mathbf{w}_{\setminus \omega}, \mathbf{z}^B, \eta, \beta) \\ &= P(l = \text{content} | \eta) P(w | \mathbf{w}_{\setminus \omega}, \mathbf{z}^B, \mathbf{1}, \beta) \\ &\approx (1 - \eta) \frac{N_{z_d^B w}^{BW \setminus \omega} + \beta}{\sum_{w'} N_{z_d^B w'}^{BW \setminus \omega} + \beta}. \end{aligned}$$