A Network Model for Dynamic Textual Communications with Application to Government Email Corpora

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1 Tie Generating Process

We assume the following generative process for each document d in a corpus D:

1. Choose the number of recipients

$$R_i^{(d)} \sim \text{zero-truncated Binomial}(A-1, \delta_i), \tag{1}$$

where $A-1$ comes from excluding the sender himself as a possible receiver (self-loop) and δ_i is the sender-specific probability of success. For example, we can use R function

```
library(actuar)
R_i = rztbinom(n = 1, size = 3, prob = 0.1)
```

2. (Data augmentation) For each sender $i \in \{1, ..., A\}$, create a list of receivers J_i by applying multivariate Wallenius' noncentral hypergeometric distribution (MWNCHypergeo) to every $j \in \mathcal{A}_{\backslash i}$

$$J_i^{(d)} \sim \text{MWNCHypergeo}\Big(\boldsymbol{m} = \mathbf{1}_{A-1}, N = R_i^{(d)}, \boldsymbol{\omega} = \{\lambda_{ij}^{(d)}\}_{j \in \mathcal{A}_{\backslash i}}\Big), \tag{2}$$

where $\mathbf{m}$ is the vector of availability (we have maximum 1 available for each actor except the sender), N is the total number of receivers to be sampled, and $\boldsymbol{\omega}$ is the weight for each actor to be sampled. Same as before, $\lambda_{ij}^{(d)}$ is evaluated at time $t_+^{(d-1)}$. Note that $_+$ denotes including the timepoint itself, meaning that λ_{ij} is obtained using the history of interactions until and including the timestamp $t^{(d-1)}$. For example, we can use R function

```
library(BiasedUrn)
J_i = rMFNCHypergeo(nran = 1, m = c(1,1,1,1), n = 2, odds = c(0.1, 0.2, 0.3, 0.4))
```

3. For every sender $i \in \mathcal{A}$, generate the time increments

$$\Delta T_{iJ_i}^{(d)} \sim \text{Exp}(\lambda_{iJ_i}^{(d)}), \tag{3}$$

where $\lambda_{iJ_i}^{(d)}(t) = \sum_{c=1}^{C} p_c^{(d)} \cdot \exp\Big\{\lambda_0^{(c)} + \frac{1}{|J_i|} \sum_{j \in J_i} \boldsymbol{b}^{(c)T} \boldsymbol{x}_t^{(c)}(i,j)\Big\} \cdot \prod_{j \in J_i} 1\{j \in \mathcal{A}_{\backslash i}\}$.

4. Set timestamp, sender, and receivers simultaneously (NOTE: $t^{(0)} = 0$):

$$\begin{aligned} t^{(d)} &= t^{(d-1)} + \min(\Delta T_{iJ_i}), \\ i^{(d)} &= i_{\min(\Delta T_{iJ_i})}, \\ J^{(d)} &= J_{i^{(d)}}. \end{aligned} \tag{4}$$

2 Inference

$$\begin{aligned}
&P(\mathcal{J}_{\text{a}}^{(d)}, \mathcal{T}_{\text{a}}^{(d)}, i_{\text{o}}^{(d)}, J_{\text{o}}^{(d)}, t_{\text{o}}^{(d)} | \mathcal{I}_{\text{o}}^{(<d)}, \mathcal{J}_{\text{o}}^{(<d)}, \mathcal{T}_{\text{o}}^{(<d)}, \mathcal{Z}, \mathcal{C}, \mathcal{B}, \boldsymbol{\delta}) \\
&= P\Big(\text{Number of recipients}\Big) \times P\Big(\text{Edge generation}\Big) \times P\Big(\text{Time generation}\Big) \times P\Big(\text{choose the observed}\Big) \\
&= \prod_{i\in\mathcal{A}} \Big(R_i^{(d)} \sim \text{ztbinom}(A-1, \delta_i)\Big) \times \prod_{i\in\mathcal{A}} \Big(J_i^{(d)} \sim \text{MWNCHypergeo}\Big(\mathbf{1}_{A-1}, R_i^{(d)}, \{\lambda_{ij}^{(d)}\}_{j\in\mathcal{A}_{\backslash i}}\Big)\Big) \\
&\quad\times \prod_{i\in\mathcal{A}} \Big(\Delta T_{iJ_i}^{(d)} \sim \text{Exp}(\lambda_{iJ_i}^{(d)})\Big) \times \prod_{i\in\mathcal{A}_{\backslash i_o^{(d)}}} P\Big(\Delta T_{iJ_i}^{(d)} > \Delta T_{i_o^{(d)} J_o^{(d)}}\Big) \\
&= \Big(\prod_{i\in\mathcal{A}} \binom{A-1}{R_i^{(d)}} \frac{\delta_i^{R_i^{(d)}} (1-\delta_i)^{A-1-R_i^{(d)}}}{1-(1-\delta_i)^{A-1}}\Big) \times \Big(\prod_{i\in\mathcal{A}} \text{dMWNCHypergeo}\Big(J_i^{(d)}; \mathbf{1}_{A-1}, R_i^{(d)}, \{\lambda_{ij}^{(d)}\}_{j\in\mathcal{A}_{\backslash i}}\Big)\Big) \\
&\quad\times \Big(\prod_{i\in\mathcal{A}} \lambda_{iJ_i}^{(d)} e^{-\Delta T_{iJ_i}^{(d)} \lambda_{iJ_i}^{(d)}}\Big) \times \Big(\prod_{i\in\mathcal{A}_{\backslash i_o^{(d)}}} e^{-\Delta T_{i_o^{(d)} J_o^{(d)}}^{(d)} \lambda_{iJ_i^{(d)}}^{(d)}}\Big),
\end{aligned} \tag{5}$$

with the probability mass function of dMWNCHypergeo with $\boldsymbol{m} = \mathbf{1}_{A-1}$ given as

$$\text{dMWNCHypergeo}(\boldsymbol{x}; \boldsymbol{m} = \mathbf{1}_{A-1}, n = R_i^{(d)}, \boldsymbol{\omega} = \{\lambda_{ij}^{(d)}\}_{j\in\mathcal{A}_{\backslash i}}) = \int_0^1 \prod_{i=1}^{A-1} (1 - t^{\omega_i/d})^{x_i} dt,$$

where $d = \sum_{i=1}^{A-1} \omega_i(1-x_i)$, $\boldsymbol{x}$ is the $(A-1)$ length vector indicating the receivers, $\boldsymbol{\omega} = (\omega_1, ..., \omega_{A-1})$ is the weight or odds of each receiver to be chosen, and $n = \sum_{i=1}^{A-1} x_i$ is the total number of receivers chosen.

We can simplify this further by integreting out the latent time $\mathcal{T}_{\text{a}}^{(d)} = \{\Delta T_{iJ_i}^{(d)}\}_{i\in\mathcal{A}_{\backslash i_o^{(d)}}}$ in the last two terms:

$$\begin{aligned}
&\int_0^\infty \cdots \int_0^\infty \Big(\prod_{i\in\mathcal{A}_{\backslash i_o^{(d)}}} \lambda_{iJ_i}^{(d)} e^{-(\Delta T_{iJ_i}^{(d)} + \Delta T_{i_o^{(d)} J_o^{(d)}}^{(d)})\lambda_{iJ_i}^{(d)}}\Big) d\Delta T_{1J_1}^{(d)} \cdots d\Delta T_{AJ_A}^{(d)} \\
&= \prod_{i\in\mathcal{A}_{\backslash i_o^{(d)}}} e^{-\Delta T_{i_o^{(d)} J_o^{(d)}}^{(d)} \lambda_{iJ_i}^{(d)}} \Big(\int_0^\infty \lambda_{iJ_i}^{(d)} e^{-\Delta T_{iJ_i}^{(d)} \lambda_{iJ_i}^{(d)}} d\Delta T_{iJ_i}^{(d)}\Big) \\
&= \prod_{i\in\mathcal{A}_{\backslash i_o^{(d)}}} e^{-\Delta T_{i_o^{(d)} J_o^{(d)}}^{(d)} \lambda_{iJ_i}^{(d)}} \Big(\Big[-e^{-\Delta T_{iJ_i}^{(d)} \lambda_{iJ_i}^{(d)}}\Big]_{\Delta T_{iJ_i}^{(d)}=0}^{\infty}\Big) \\
&= e^{-\Delta T_{i_o^{(d)} J_o^{(d)}}^{(d)} \sum_{i\in\mathcal{A}_{\backslash i_o^{(d)}}} \lambda_{iJ_i}^{(d)}},
\end{aligned} \tag{6}$$

where $\Delta T_{i_o^{(d)} J_o^{(d)}}^{(d)}$ is the observed time difference between d^{th} and $(d-1)^{th}$ document (i.e. $t^{(d)} - t^{(d-1)}$). Therefore, we can simplify Equation (5) as below:

$$\begin{aligned}
&P(\mathcal{J}_{\text{a}}^{(d)}, i_{\text{o}}^{(d)}, J_{\text{o}}^{(d)}, t_{\text{o}}^{(d)} | \mathcal{I}_{\text{o}}^{(<d)}, \mathcal{J}_{\text{o}}^{(<d)}, \mathcal{T}_{\text{o}}^{(<d)}, \mathcal{Z}, \mathcal{C}, \mathcal{B}, \boldsymbol{\delta}) \\
&= \Big(\prod_{i\in\mathcal{A}} \binom{A-1}{R_i^{(d)}} \frac{\delta_i^{R_i^{(d)}} (1-\delta_i)^{A-1-R_i^{(d)}}}{1-(1-\delta_i)^{A-1}}\Big) \times \Big(\prod_{i\in\mathcal{A}} \int_0^1 \prod_{i=1}^{A-1} (1 - t^{\lambda_{ij}^{(d)} / \sum_{i=1}^{A-1} \lambda_{ij}^{(d)} (1-J_{ij}^{(d)})})^{J_{ij}^{(d)}} dt\Big) \\
&\quad\times \Big(\lambda_{i_o^{(d)} J_o^{(d)}}^{(d)}\Big) \times \Big(e^{-\Delta T_{i_o^{(d)} J_o^{(d)}}^{(d)} \sum_{i\in\mathcal{A}_{\backslash i_o^{(d)}}} \lambda_{iJ_i^{(d)}}^{(d)}}\Big),
\end{aligned} \tag{7}$$

where this joint distribution can be interpreted as 'probability of choosing the number of recipients from zero-truncated Binomial distribution × probability of choosing the latent receivers from Wallenius' noncenral hypergeometric distribution × probability of the observed time comes from Exponential distribution × probability of all latent time greater than the observed time, given that the latent time also come from Exponential distribution.'