The Hyperedge Event Model

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Collaborators







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Motivations: Hyperedge Event Model

- ► Hyperedge: directed edges from one sender to multiple receivers or from multiple senders to one receiver
- ▶ Event: timestamped events in the continous-time scale
- ▶ Model: statistical framework to jointly understand
 - "who interacts with whom, and when?"

Generative Process: "Who Interacts with Whom"

For event $e = 1, \dots, E$, between $i \in \{1, \dots, A\}$ & $j \in \{1, \dots, A\}$,

▶ Receiver intensity for every sender-receiver pair $(i,j)_{i\neq j}$

$$\lambda_{iej} = \boldsymbol{b}^{\mathsf{T}} \boldsymbol{x}_{iej},$$

where x_{iej} is a set of receiver selection features or covariates and b is the corresponding P-dimensional coefficient.

▶ Every sender i selects candidate receivers from non-empty multivariate Bernoulli distribution $u_{ie} \sim \mathsf{MB}_G(\lambda_{ie1}, \dots, \lambda_{ieA})$

$$P(\boldsymbol{u}_{ie}|\boldsymbol{b}, \boldsymbol{x}_{iej}) \propto \exp\Big(\log(I(||\boldsymbol{u}_{ie}||_1 > 0)) + \sum_{j \neq i} \lambda_{iej} u_{iej}\Big)$$

¹Fellows and Handcock (2017); Dai et al. (2013)



Generative Process: "and When"

▶ Timing rate for each sender i

$$\mu_{ie} = g^{-1}(\boldsymbol{\eta}^T \boldsymbol{y}_{ie}),$$

where \mathbf{y}_{ie} is a set of timing features or covariates and η is the corresponding Q-dimensional coefficient.

▶ Generalized linear model (GLM) for time increment τ_{ie} so that

$$E(\tau_{ie}) = \mu_{ie}$$
 and $V(\tau_{ie}) = V(\mu_{ie})$,

with a choice of distribution from exponential family.

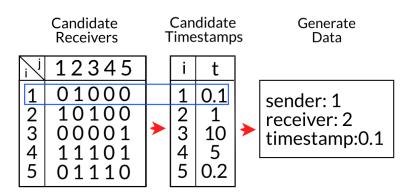
► Select the sender-receiver-set with the smallest time increment²

$$egin{aligned} \mathbf{s}_e &= \operatorname{argmin}_i(au_{ie}), \ \mathbf{r}_e &= \mathbf{u}_{s_e e}, \ t_e &= t_{e-1} + au_{s_e e}. \end{aligned}$$



²Snijders (1996)

Generative Process: Sender, Receivers, and Timestamps³



Bayesian inference: invert the generative process to obtain the posterior distribution over the latent variables—i.e., (u, b, η) .



³Assuming $t_{e-1} = 0$ for simplicity.

Application: Montgomery County Government Email Data⁴

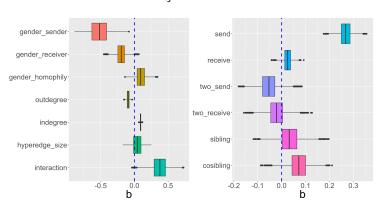
- Email corpora covering inboxes and outboxes of Montgomery county government managers in North Carolina
- ► Contains E = 680 emails, sent and received by A = 18 department managers over 3 months (March–May) in 2012.

"To what extent are nodal, dyadic or triadic network effects relevant to predicting future emails?"



Results: Exploratory Analysis on **b**

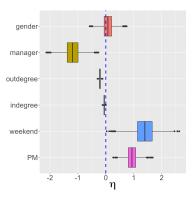
$$\operatorname{logit}(\lambda_{iej}) = \operatorname{log}\left(\frac{\lambda_{iej}}{1 - \lambda_{iei}}\right) = b_1 + b_2 x_{iej2} \ldots + b_{14} x_{iej14},$$



- Log odds is two times less if the sender is a woman.
- ▶ If i sent n number of emails to j last week, then i is $e^{0.27n} \approx (1.32)^n$ times more likely to send an email to j.

Results: Exploratory Analysis on η

$$\log(\tau_{ie}) \sim N(\mu_{ie}, \sigma_{\tau}^2)$$
, with $\mu_{ie} = \eta_1 + \eta_2 y_{ie2} \ldots + \eta_7 y_{ie7}$.

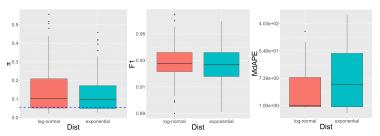


- ▶ If an email was sent during weekend or PM, then time to next email takes $e^{1.55} \approx 4.72$ and $e^{0.98} \approx 2.67$ hours longer.
- ▶ manager, outdegree, and indegree shorten time to next email.

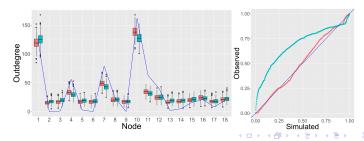


Comparison: Lognormal vs. Exponential

Out-of-sample predictions: sender, receiver, and timestamp



► Posterior predictive checks (PPC)



Conclusion and Discussion

- Account for hyperedges without duplications
- ► Flexible choice of continuous-time distribution via GLM
- Reverse the process for multiple senders to one receiver (e.g., international sanctions and co-sponsorship of bills)
- Sources: http://arxiv.org/abs/1807.08225
 https://github.com/desmarais-lab/MulticastNetwork