# The Multivariate Community Hawkes Model for Dependent Relational Events in Continuous-time Networks

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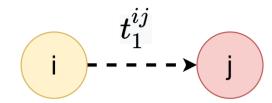


#### Motivation

- The Multivariate Community Hawkes (MULCH):
  - Flexible model for continuous-time networks that introduces dependence between node pairs in a controlled manner.

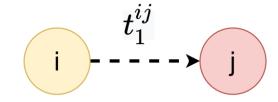
#### Continuous-time networks:

- Events between node pairs (i,j) at time  $t^{ij}$
- Ex: messages between users on social media.



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• Goal: Analyze interactions between nodes and simulate realistic networks.

#### Background: Multivariate Hawkes Process

- Multivariate Hawkes Process :
  - **Temporal point process** generates a sequence of timestamps  $\{t_1, t_2, ..., \}$ .
  - Self and Mutually-exciting temporal processes.
  - Event in one process excites future events in self and other processes.

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- Network Setup
  - node pairs (i, j) = processes
  - An Event between (i,j) (dashed)  $\rightarrow$  future events in both (i,j) and (j,i) (solid).

$$\{t_1^{ij}, t_2^{ij}, ..., \}$$

$$\{t_1^{ji}, t_2^{ji}, ..., \}$$

#### Background: Multivariate Hawkes Process

• Intensity function: instantaneous rate of events at time t

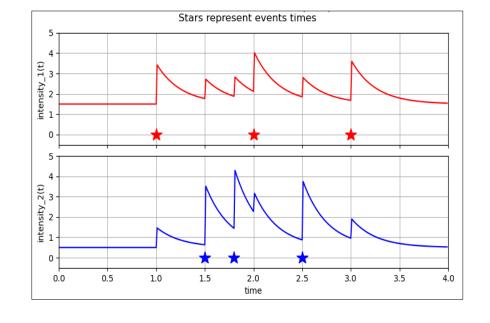
• Intensity of a node pair (i, j) – exponential kernel:

$$\lambda_{ij}(t) = \mu_{ij} + \sum_{(x,y)} \alpha^{xy \to ij} \sum_{s: t_s^{xy} < t} \beta e^{\beta(t - t_s^{xy})}$$

 $\mu_{ij}$  : baseline intensity

 $\alpha^{xy \to ij}$ : excitation from (x, y) to (i, j)

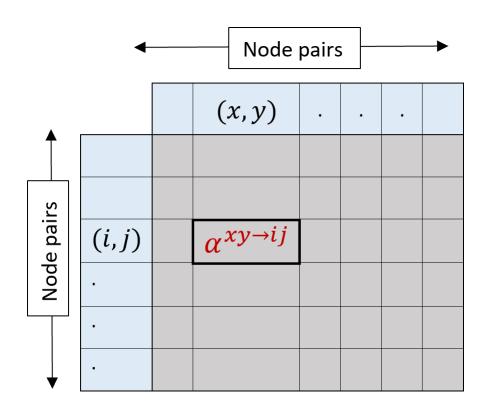
 $\beta$  : exponential decay rate



#### Problem!

$$\lambda_{ij}(t) = \mu_{ij} + \sum_{(x,y)} \alpha^{xy \to ij} \sum_{s: t_s^{xy} < t} \beta e^{\beta(t - t_s^{xy})}$$

- For a directed network with n nodes:
  - Number node pairs =  $n(n-1) \approx n^2$
  - Excitation matrix  $\alpha \approx n^2 \times n^2$
  - Intractable even for n = 100 nodes!



**Excitation matrix** 

- 1. Adapted from Stochastic Block Model (SBM), assume K blocks/communities:
  - Each node  $i \in block a$
  - Each node pair  $(i, j) \in \text{block pair } (a, b)$

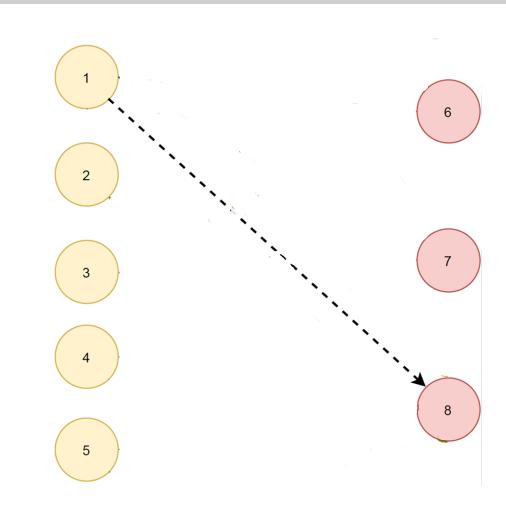


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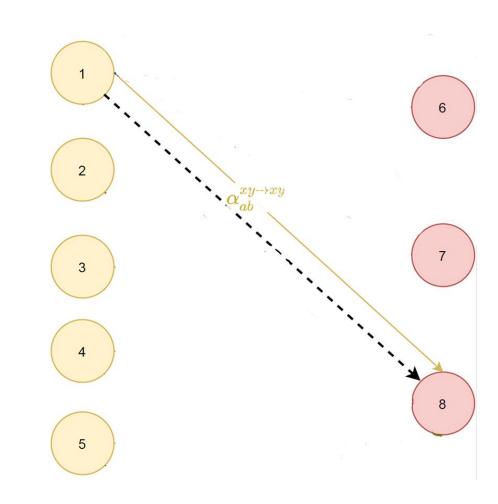


2. A node pair  $(i, j) \in \text{block pair } (a, b) \text{ can only be excited by node pairs in block pair } (a, b) & \text{block pair } (b, a)$ 

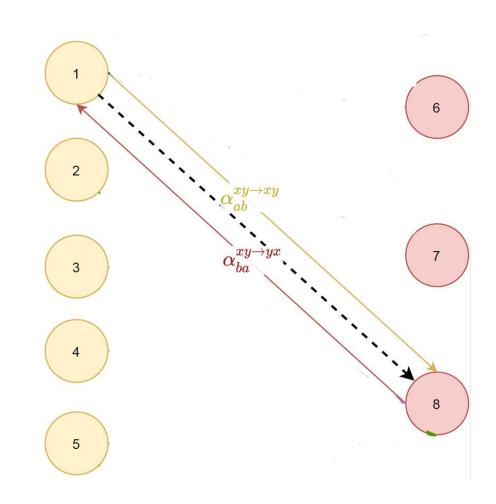
- 3. Inspired by research in social sciences, we model **6 types of excitation**:
- Example:
  - Nodes 1-5 in block a
  - Nodes 6-8 in block b
  - Message between (1,8)



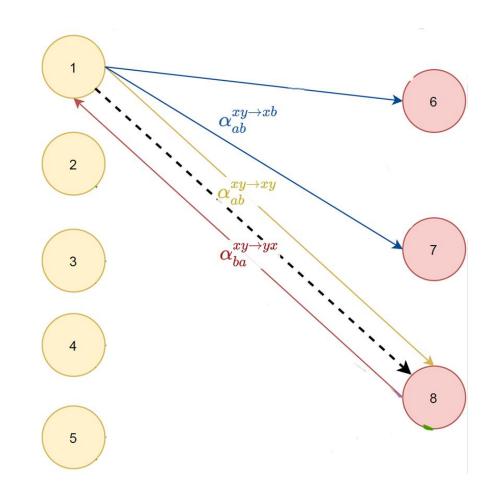
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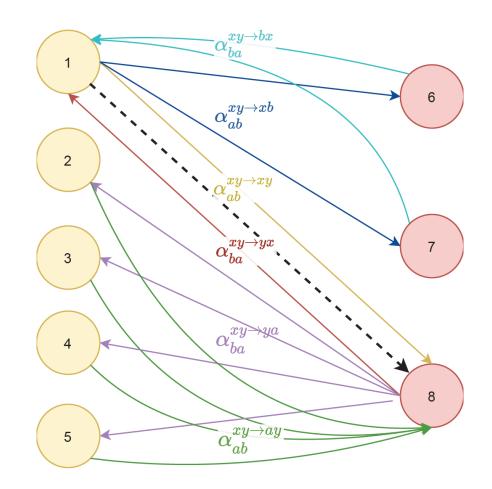
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  - ➤ 1 to 6 and 7 (Turn continuation)



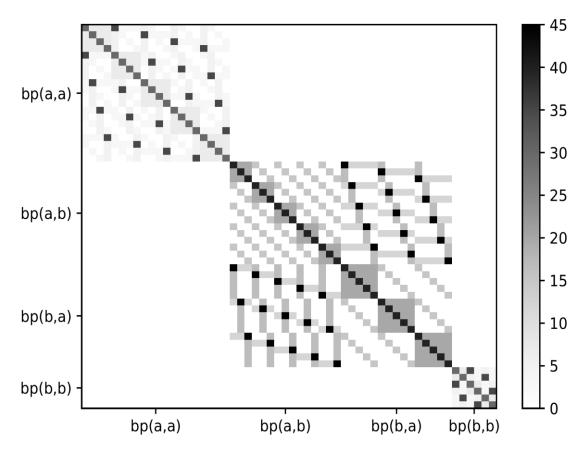
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  - ➤ 1 to 6 and 7 (Turn continuation)
  - ➤ Other three (Generalized reciprocity, Allied continuation, Allied reciprocity)



## What have we gained?

• Estimation: spectral clustering with a likelihood refinement procedure

- 1. Community/block detection.
- 2. Excitation matrix has a diagonal structure (sparser).
- 3. Meaningful estimated parameters.



## Results: Predictive Accuracy

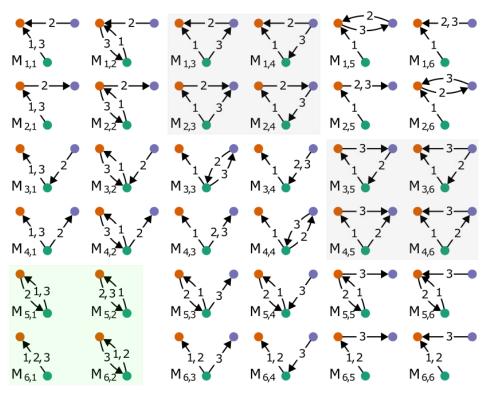
- Split dataset into 0.8 train: 0.2 test
- Predictive log-likelihood  $\mathcal L$  score:

$$test \mathcal{L} = \frac{(full \mathcal{L} - train \mathcal{L})}{number of test events}$$

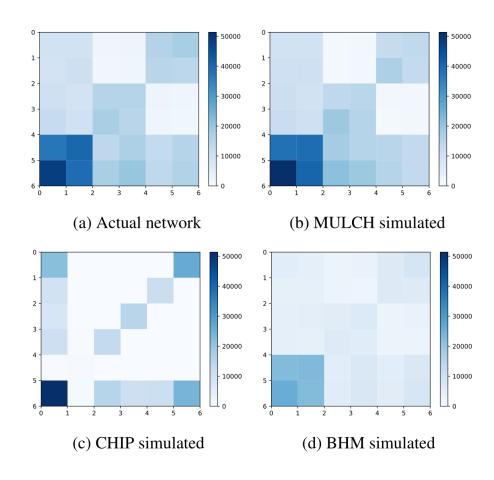
Model	Reality	Enron	MID	Facebook
MULCH	-3.82	-5.13	-3.53	-6.82
внм	-5.37	-7.49	-5.33	-14.4
CHIP	-4.83	-5.61	-3.67	-9.46
REM	-6.11	-6.84		
ADM4	-8.52			
DLS	-5.65	-7.57	-4.52	

## Results: Generative Accuracy

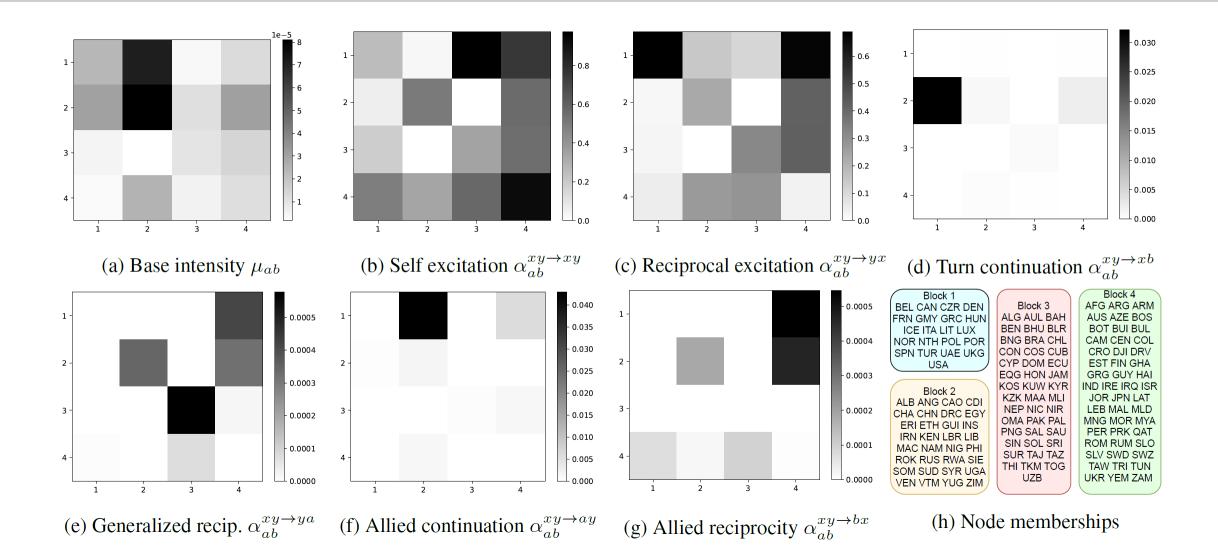
• Compare counts of temporal motifs on Reality Mining



**Photo credit**: Paranjape, A., Benson, A. R., and Leskovec, J. Motifs in temporal networks. In Proceedings of the 10th ACM International Conference on Web Search and Data Mining, pp. 601–610, 2017.



## Case Study: Militarized Interstate Disputes dataset



# Thank you!



• Code: <a href="https://github.com/IdeasLabUT/Multivariate-Community-Hawkes">https://github.com/IdeasLabUT/Multivariate-Community-Hawkes</a>









