# **Cardinality-Regularized Hawkes-Granger Model** Tsuyoshi ("Ide-san") Idé, Georgios Kollias, Dzung T. Phan, Naoki Abe / IBM T.J. Watson Research Center



- $\phi_d(t)$ : decay function of the type-d
- $\circ A_{k,l}$ : Impact matrix (btwn. type k and l)

Relationship with Granger causality  $\circ$  If  $A_{k,l} = 0$ , then type l is Granger-noncause of type k

# **Challenges in maximum likelihood**

Log likelihood

$$= \sum_{n=1}^{N} \left\{ \ln \lambda_{d_n}(t_n \mid \mathcal{H}_{n-1}) - \int_{t_{n-1}}^{t_n} du \ \lambda_{d_n}(u \mid \mathcal{H}_{n-1}) \right\}$$

- Native maxim likelihood is numerically very tricky
- EM-like algorithm ("MM") is efficient but does not lead to a <u>sparse</u> solution

# What about existing "sparse" Hawkes models?

**Theorem 1**: The L1-regularized Hawkes model does not provide any sparse solution for **A**.

Simply because the Jensen bound of log-likelihood depends on  $+\ln A_{k,l}$ • Prohibits  $A_{k,l} = 0$  for all (k, l).

#### **Research question**

How do we get a <u>sparse</u> solution in the MM framework of the Hawkes process? k=1



**Applications** ( $\rightarrow$  paper) Event grouping and deduplication in AlOps Failure propagation analysis in power grid



### **LO-regularized Hawkes process**

Log likelihood for A under LO (and L2) penalty takes the form of

$$\sum_{l=1}^{D} \left( Q_{k,l} \ln A_{k,l} - H_{k,l} A_{k,l} - \frac{1}{2} \nu_A A_{k,l}^2 \right) - \tau \|\mathbf{A}\|_0$$

#### Main contribution

• Found a semi-analytic solution under LO penalty with the notion of " $\epsilon$ -sparsity" • Applied it to simultaneous instanceand type-level sparse causal diagnosis

# **Instance- and type-level Granger** causal diagnosis: How?

Type-level causal analysis:

 $\circ$  Look at  $\{A_{k,l}\}$  – If 0, no causal relation btwn k,l.

Instance-level causal analysis:

 $\circ$  Look at the variational distribution  $\{q_{n,i}\}$  in the MM algorithm – degree of how strongly instance *n* was triggered by instance *i* 

