

Uncovering Causality from Multivariate Hawkes Integrated Cumulants

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Summary

We propose a new method to measure causal relationships between nodes based on their activity timestamps. We answer to the question: *How many events of type j does an event of type i cause?*

Introduction

A **temporal point process** is a collection of points $Z = \{\tau_1, \tau_2, \dots\}$ randomly located on \mathbb{R}^+ . Another representation is based on the counting process $N_t = \sum_{\tau \in Z} \mathbb{1}_{\tau \leq t}$.

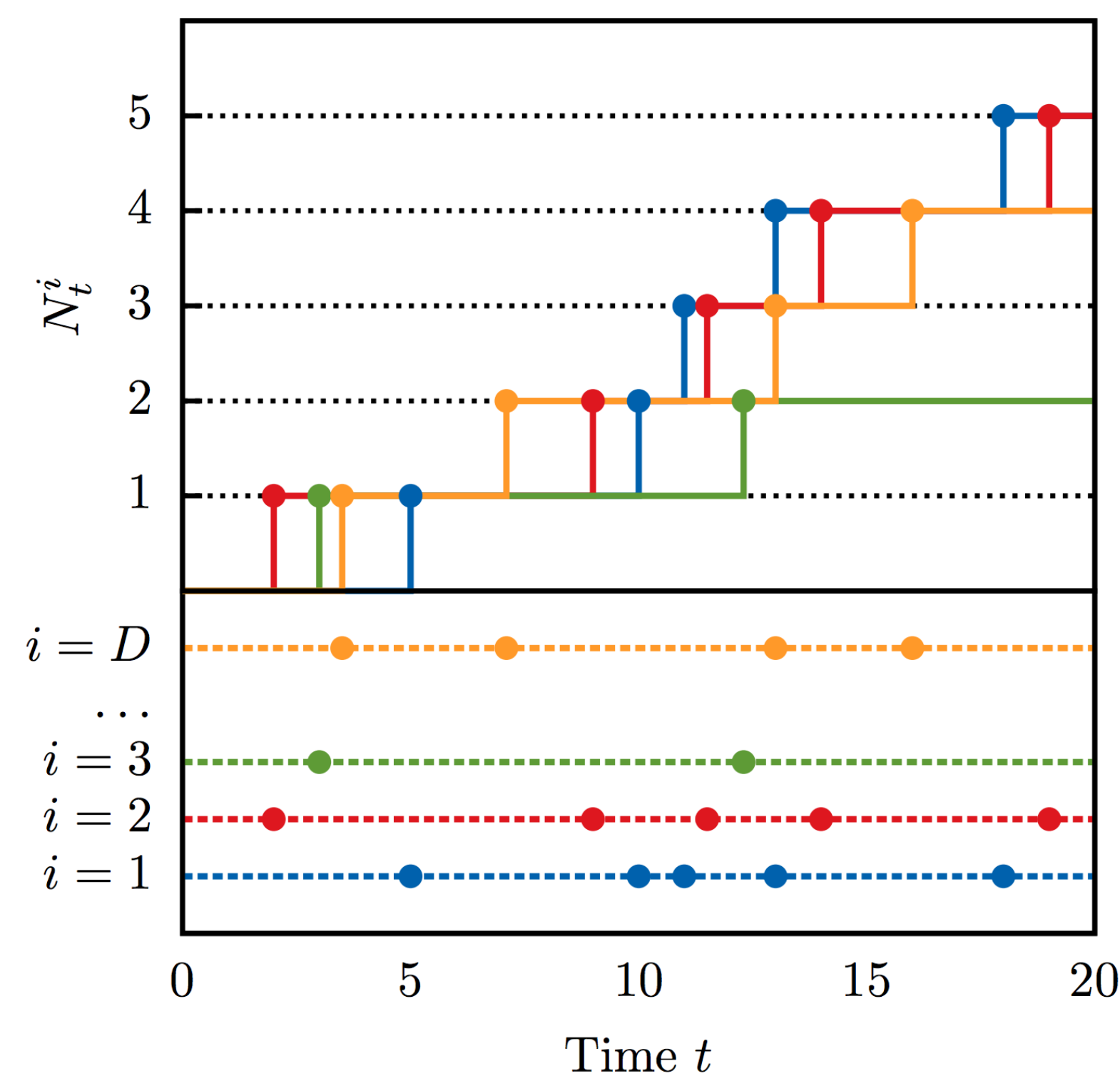


Figure 1: Example of multivariate point process

The vector of stochastic intensities $\lambda_t = [\lambda_t^1 \dots \lambda_t^D]^\top$ associated with the multivariate counting process N_t is defined as

$$\lambda_t^i = \lim_{dt \rightarrow 0} \frac{\mathbb{P}(N_{t+dt}^i - N_t^i = 1 | \mathcal{F}_t)}{dt}.$$

N_t is called a **Hawkes point process** if the stochastic intensities can be written as

$$\lambda_t^i = \mu^i + \sum_{j=1}^D \sum_{\tau_j \in Z^j} \phi^{ij}(t - \tau_j),$$

where $\mu^i \in \mathbb{R}^+$ is an exogenous intensity and ϕ^{ij} are positive, integrable and causal (with support in \mathbb{R}^+) functions called *kernels* encoding the impact of an action by j on the activity of i .

Hawkes Causality

Instead of trying to estimate the kernels ϕ^{ij} , we focus on the direct estimation of their *integrals*:

$$g^{ij} = \int_0^{+\infty} \phi^{ij}(u) du. \quad (1)$$

The matrix \mathbf{G} can be seen as a weighted adjacency matrix. Using Hawkes process representation as a Poisson cluster process [1], we introduce the counting function $N_t^{i \leftarrow j}$ that counts the number of events of i whose direct ancestor is an event of j . We know that

$$\mathbb{E}[dN_t^{i \leftarrow j}] = g^{ij} \mathbb{E}[dN_t^j] = g^{ij} \Lambda^j dt. \quad (2)$$

The integral g^{ij} is **the average number of events of i triggered by one event of type j** , as described at Eq. (2). This is also linked to Granger causality (see the paper for more details).

Integrated Cumulants

A general formula for the cumulants of Hawkes processes is provided in [2] (cumulants are centered moments, up to the third order). The integrals of these cumulants are related to $\mathbf{R} = (\mathbf{I} - \mathbf{G})^{-1}$. For the purpose of our method, we only need to consider cumulants up to the third order.

$$\Lambda^i = \sum_{m=1}^D R^{im} \mu^m \quad (3)$$

$$C^{ij} = \sum_{m=1}^D \Lambda^m R^{im} R^{jm} \quad (4)$$

$$S^{ijk} = \sum_{m=1}^D (R^{im} R^{jm} C^{km} + R^{im} C^{jm} R^{km} + C^{im} R^{jm} R^{km} - 2\Lambda^m R^{im} R^{jm} R^{km}). \quad (5)$$

We designed asymptotically unbiased estimators of the integrated cumulants above, see the paper for details.

Nonparametric Estimation

To recover \mathbf{G} , we minimize the following criterion using SGD:

$$\mathcal{L}(\mathbf{R}) = (1 - \kappa) \|\mathbf{S}_c(\mathbf{R}) - \widehat{\mathbf{S}}_c\|_2^2 + \kappa \|\mathbf{C}(\mathbf{R}) - \widehat{\mathbf{C}}\|_2^2, \quad (6)$$

Our method is called **NPHC** (for Nonparametric Hawkes Cumulants).

Consistency of NPHC

Suppose that (N_t) is observed on \mathbb{R}^+ and assume that

- $g_0(\mathbf{R}) = 0$ if and only if $\mathbf{R} = \mathbf{R}_0$;
- $\mathbf{R} \in \Theta$, which is a compact set;
- the spectral radius of the kernel norm matrix satisfies $\|\mathbf{G}_0\| < 1$;
- $H_T \rightarrow \infty$ and $H_T^2/T \rightarrow 0$.

Then

$$\widehat{\mathbf{G}}_T = \mathbf{I} - \left(\arg \min_{\mathbf{R} \in \Theta} \mathcal{L}_T(\mathbf{R}) \right)^{-1} \xrightarrow{\mathbb{P}} \mathbf{G}_0.$$

Results

We consider the following 8-dimensional point process, already introduced in [3], which models the dynamic of an order book:

$$N_t = (P_t^+, P_t^-, T_t^a, T_t^b, L_t^a, L_t^b, C_t^a, C_t^b)$$

Each dimension counts the number of events before t :

- P^+ (P^-): upwards (downward) mid-price move triggered by any order.
- T^a (T^b): market order at the ask (bid) that does not move the price.
- L^a (L^b): limit order at the ask (bid) that does not move the price.
- C^a (C^b): cancel order at the ask (bid) that does not move the price.

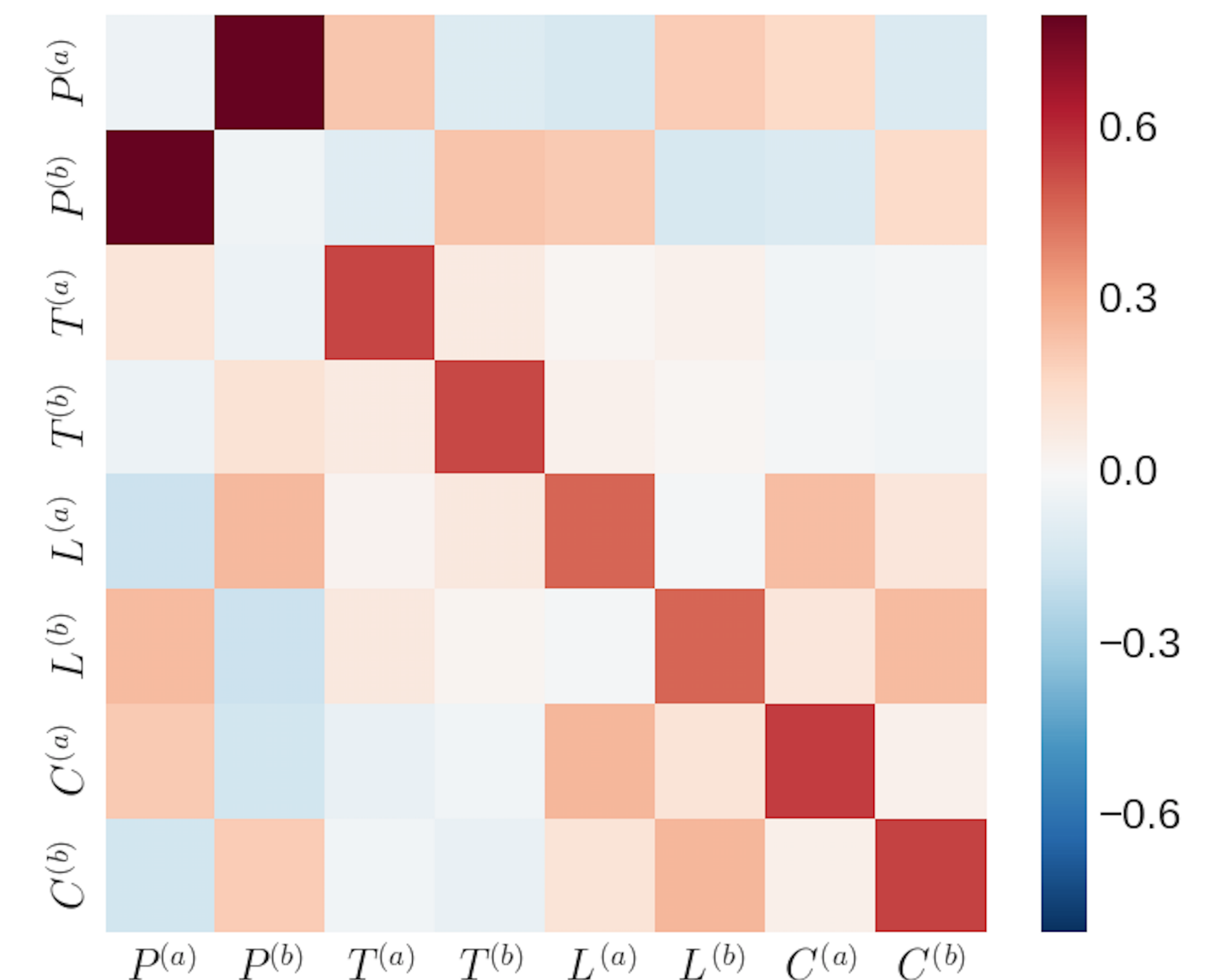


Figure 2: Estimated $\widehat{\mathbf{G}}$ on DAX order book data.

References

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