

UNCOVERING PRODUCT CANNIBALISATION

USING MULTIVARIATE HAWKES PROCESSES

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RESEARCH GOAL




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We aim to quantify **product cannibalisation** for existing products in an apparel wholesale data set.

Product cannibalisation refers to the decrease in sales of one product due to (the introduction of) a closely related product. [Copulsky, 1976]

The current focus is on **inference** to detect and understand product cannibalisation for products that already have a sales history.

Very important from a business perspective!

-  Find a model
-  Implement that model
-  Fit that model on real data

FIND A MODEL

POINT PROCESSES

Data: event times (plus additional covariates)

Let $N(t)$ be the number of observed events from 0 to t .

Homogeneous (λ constant) Poisson point process:

$$\mathbb{P}[N(t) = N] = \frac{(\lambda t)^N}{N!} e^{-\lambda t} \quad (1)$$

$$\mathbb{E}[N(t)] = \lambda t \quad (2)$$

Inhomogeneous ($\lambda(t)$ variable) Poisson point process:

$$p(t_1 \dots t_N) = \prod_{i=1}^N \lambda(t_i) e^{-\int_0^\infty \lambda(z) dz} \quad (3)$$

$$\mathbb{E}[N(t)] = \int_0^t \lambda(z) dz \quad (4)$$

[Daley and Vere-Jones, 2003]

UNIVARIATE HAWKES PROCESS

Hawkes processes [Hawkes, 1971] are a class of point processes that are used to model event data when the events can occur in clusters or bursts. They are defined on the interval $[0, T]$ with **conditional intensity function**:

$$\lambda(t) = \mu(t) + \sum_{i:t > t_i} K g(t - t_i) \quad (5)$$

Here, $\mu(t)$ can capture seasonality and underlying trends and we use $g(t - t_i) = \beta e^{-\beta(t-t_i)}$ for the self-excitement.

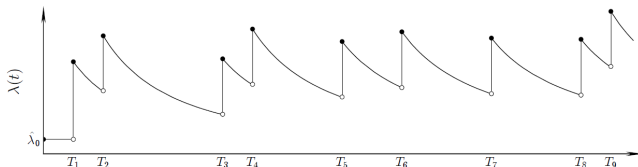


Figure 1: Intensity function with self exciting kernel [Rizoiu et al., 2017]

MULTIVARIATE HAWKES PROCESS

Assume that there are M dimensions with data

$Y_1 = (t_{11} \dots t_{1N_1}) \dots Y_M = (t_{M1} \dots t_{MN_M})$. At time t the intensity in dimension i is defined as the sum of the background rate $\mu_i(t)$ and contributions from all dimensions:

$$\lambda_i(t) = \left[\mu_i(t) + \sum_{j=1}^M \sum_{l:t > t_{jl}} K_{ji} g_{ji}(t - t_{jl}) \right]_+ \quad (6)$$

We assume the following form for the influence for all i, j :

$g_{ij}(x) > 0$ for $x > 0$ and $\int_0^\infty g_{ij}(x) dx = 1$. Here, each $K_{ij} < 1$, and we write them as matrix $\mathbf{K} = \{K_{ij}\}$ where $i, j = 1 \dots M$.

We use a **multivariate** Hawkes Process where each dimension represents one product. This allows us to estimate the ‘influence’ K_{ij} from an event (sale) of one product i onto each product $j = 1 \dots M$.

A positive influence $K_{ij} > 0$ is called **excitation**, a negative influence $K_{ij} < 0$ is referred to as **inhibition**. The latter is interpreted as **product cannibalisation**.

IMPLEMENT THAT MODEL

"UNDER THE HOOD"

To implement this model we needed to overcome a few challenges:

- Ensuring a non-negative intensity \rightarrow link function
- Integrating the intensity \rightarrow numerical approximation
- Checking for stability \rightarrow new criterion

All methodological details can be found in the mathematical draft on ArXiv [Deutsch and Ross, 2022].

 **FIT THAT MODEL ON REAL DATA**

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POSTERIOR

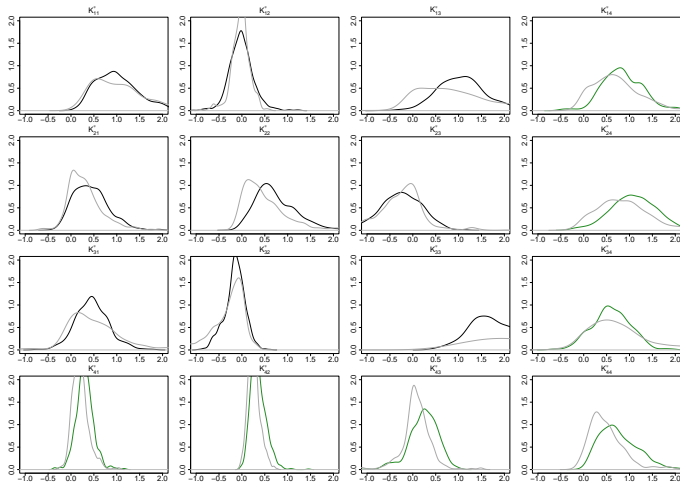


Figure 2: Posterior density estimates of K^* (excitation/inhibition parameter), using independent Normal (black) and horseshoe (grey) priors.

SUMMARY

Methodological Advances




to make the implementation of a multivariate Hawkes Process with inhibition easier.

Formalisation of Product Cannibalisation

as a mathematical concept that can be estimated, monitored, and predicted.

[Deutsch and Ross, 2022]
arxiv.org/abs/2201.05009



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