

# Lin\_Masters

*Frances Lin*

*June 2021*

Major Professor: James Molyneux

Committee Members: Lisa Madsen & Charlotte Wickham

## Abstract

## 1 Introduction

### Motivation

### Applications

Applications of Hawkes Processes can be found in a wide variety of fields such as seismology, criminology, insurance, finance, social media, and neuroscience.

In seismology, an event can be an earthquake occurrence that causes aftershocks. In criminology, an event can be a gang rivalry that triggers retaliations following the gang crime. In insurance, an event can be a standard claim that increases claims. In finance, an event can be a transaction that influences future prices or volumes of transactions or a news that leads to movements in stock prices or trading behaviors. In social media, an event can be a tweet about an event on Twitter that follows a cascade of retweets from other users on the same social networking platform. In neuroscience, an event can be firing of a neuron that triggers spikes (or action potentials) of other neurons.

### Objectives

The objectives of this project is to . . . , which include

1. defining and discussing properties of counting process, (homogeneous) Poisson process, nonhomogeneous Poisson process, Hawkes process, and spatio-temporal Hawkes process,
2. discussing the algorithms and simulating all of the above processes,
3. extending (temporal) Hawkes process to spatio-temporal Hawkes process, and
4. discussing future work of Hawkes (self-exciting) process.

## 2 Definitions and Properties

### 2.1 Counting Process

**Definition 2.1.1** (Point Process) Let  $\{T_i, i \in N\}$  be a sequence of non-negative random variables such that  $T_i < T_{i+1} \forall i \in N$ , a point process on  $R^+$  is defined as

$$\{T_i, i \in N\}$$

**Definition 2.1.2** (Stochastic Process) A stochastic process is a family of random variables indexed by time  $t$  and is defined as

$$\{X(t), t \in T\}$$

**Definition 2.1.3** (Counting Process) Let  $N(t)$  be the total number of events up to some time  $t$ , a stochastic process is said to be a counting process and is defined as

$$\{N(t), t \geq 0\}$$

**Definition 2.1.4** (Counting Process) Let  $\{T_i, i \in N\}$  be a point process, a counting process associated with  $\{T_i, i \in N\}$  is defined as

$$N(t) = \sum_{i \in N} I_{\{T_i \leq t\}}$$

**Corollary 2.1.1** A counting process satisfies that

1.  $N(t) \geq 0$
2.  $N(t)$  is an integer
3. If  $s \leq t$ , then  $N(s) \leq N(t)$
4. If  $s < t$ , then  $N(t) - N(s)$  is the number of events occur in the interval  $(s, t]$

**Proposition 2.1.1** A counting process has the following properties

1. (Independence) If... , a counting process is said to have independent increments.
2. (Stationarity) If... , a counting process is said to have stationary increments.
3. (Homogeneity)

## 2.2 Poisson Process

**Definition 2.2.1** (Poisson Process) If the following conditions hold, a counting process  $\{N(t), t \geq 0\}$  is said to be a Poisson Process with constant rate (or intensity)  $\lambda > 0$

1.  $N(0) = 0$
2.  $N(t)$  has independent increments
3.  $P(N(t+h) - N(t) = 1) = \lambda h + o(h)$
4.  $P(N(t+h) - N(t) \geq 2) = o(h)$

In other words, 1. the process starts at  $t = 0$ , 2. the increments are independent, 3.  $\lambda$  is the rate (or intensity), and 4. no 2 or more events can occur at the same location.

**Proposition 2.2.1** Poisson process has the following properties

1.  $N(t)$ , the number of events in any interval  $t$ ,  $\sim \text{Pos}(\lambda t)$ . That is, for all  $s, t \geq 0$

$$P(N(t+s) - N(s) = n) = \frac{(\lambda t)^n e^{-\lambda t}}{n!}$$

$n = 0, 1, \dots$

2.  $W$ , the interarrival times,  $\overset{iid}{\sim} \exp(\frac{1}{\lambda})$ . That is, for rate  $\lambda > 0$ , the interarrival time  $W_i$   $i = 1, 2, \dots$

$$P(W_1 > t) = P(N(t) = 0) = e^{-\lambda t}$$

This is because  $p(1^{st}$  arrival arrives after time  $t$ ) is the same as  $p(\text{no arrival in the interval } [0, t])$ ,

$$F(t) = P(W_1 \leq t) = 1 - e^{-\lambda t}$$

and

$$f(t) = \lambda e^{-\lambda t}.$$

Similarly,

$$P(W_2 > t | W_1 = s) = P(N(t+s) - N(s) = 0) = P(N(t) = 0) = e^{-\lambda t}.$$

## 2.3 Nonhomogeneous Poisson Process

Poisson process has stationary increments since the distribution of the number of events  $N(t)$  that occur in any interval of time  $t$  depends only on the length of the time interval but not the location of the interval. In contrast, nonhomogeneous Poisson process does not have stationary increments since the distribution of  $N(t)$  can change when shifted in  $t$ .

**Definition 2.3.1** (Nonhomogeneous Poisson Process) If the following conditions hold, a counting process  $\{N(t), t \geq 0\}$  is said to be a nonhomogeneous Poisson Process with intensity function of time  $\lambda(t), t > 0$

1.  $N(0) = 0$
2.  $N(t)$  has independent increments
3.  $P(N(t+h) - N(t) = 1) = \lambda(t)h + o(h)$
4.  $P(N(t+h) - N(t) \geq 2) = o(h)$

**Proposition 2.2.1** Nonhomogeneous Poisson process has the following properties

1.  $N(t)$ , the number of events in any interval  $t$ ,  $\sim \text{Pos}(\int_0^t \lambda(s) ds)$ . That is, for all  $s, t \geq 0$

$$P(N(t) = n) = \frac{(\int_0^t \lambda(s) ds)^n e^{-\int_0^t \lambda(s) ds}}{n!}$$

or

$$P(N(b) - N(a) = n) = \frac{(\int_a^b \lambda(s) ds)^n e^{-\int_a^b \lambda(s) ds}}{n!}$$

$n = 0, 1, \dots$

## 2.4 Hawkes Process

In previous cases, the events either arrive independently, arrive at a constant rate  $\lambda$  (e.g. Poisson process), or depend on an intensity function  $\lambda(t)$  (e.g. nonhomogeneous Poisson Process). In the case of Hawkes process, the arrival rate of the events depends not only on time  $t$  but also past events  $\mathcal{H}_t^N$ . In other words, the arrival of an event also increases the occurrence of future events.

**Definition 2.4.1** (Hawkes Process) A counting process  $\{N(t), t \geq 0\}$  associated with past events  $\{\mathcal{H}_t^N, t > 0\}$  is said to be a Hawkes process with conditional intensity function  $\lambda(t|\mathcal{H}_t^N), t > 0$  and takes the form

$$\lambda(t|\mathcal{H}_t^N) = \lambda_0(t) + \sum_{i:T_i < t} \phi(t - T_i)$$

where

- $\lambda_0(t)$  is the base intensity function (or  $\mu$  the constant background rate)
- $T_i < t$  are the events time occur before current time  $t$
- $\phi(\cdot)$  is the kernel function (or  $g(\cdot)$  the triggering function) through which intensity function depends on past events
- $\mathcal{H}_t^N$  is called the natural filtration (or simply  $\mathcal{H}_t$  past history) which represents the internal history of  $N$  up to time  $t$

*Corollary 2.4.1* Hawkes process satisfies that

1.  $N(t) = 0$
2.  $\lambda(t|\mathcal{H}_t^N) = \lambda_0(t) + \int_{-\infty}^t \phi(t - T_i) dN(s) = \lambda_0(t) + \sum_{i:T_i < t} \phi(t - T_i)$
3.  $P(N(t+h) - N(t) = 1 | \mathcal{H}_t^N) = \lambda(t)h + o(h)$
4.  $P(N(t+h) - N(t) \geq 2 | \mathcal{H}_t^N) = o(h)$

**2.4.1** Choices of  $\phi(\cdot)$  include, for example, exponentially decaying function and power-law kernel, and they take the form of

$$\phi(x) = \alpha e^{-\beta x}$$

$$\phi(x) = \frac{\alpha}{(x + \beta)^{\eta+1}}$$

**2.4.2** Two types of Hawkes processes include

1. Intensity-based Hawkes Process
2. Cluster-based Hawkes Process

## 2.5 Hawkes Process (Spatio-Temporal Form)

Spatio-temporal self-exciting point processes is an extension of temporal Hawkes point processes. Recall that temporal Hawkes point processes take the form of

$$\lambda(t|\mathcal{H}_t) = \mu + \sum_{i:T_i < t} g(t - t_i)$$

Spatio-temporal self-exciting point processes take the form of

$$\lambda(t|\mathcal{H}_t) = \mu(s) + \sum_{i:T_i < t} g(s - s_i, t - t_i)$$

where

- $s_i, i = 1, 2, \dots$  are the sequence of locations of events
- $t_i, i = 1, 2, \dots$  are the times of the events

For this project, we focus on

## 3 Algorithms

### 3.1 Counting Process

### 3.2 Poisson Process

### 3.3 Nonhomogeneous Poisson Process

### 3.4 Hawkes Process

## 4 Conclusions and Discussion

## Acknowledgments

## Reference

- Chen, Y. (2016). Thinning algorithms for simulating point processes. Florida State University, Tallahassee, FL.
- Obral, K. (2016). Simulation, estimation and applications of hawkes processes. (Master's thesis, University of Minnesota, Twin Cities, United States).
- Rizoiu, M. A., Lee, Y., Mishra, S., & Xie, L. (2017). A tutorial on hawkes processes for events in social media. arXiv preprint arXiv:1708.06401.