

Market microstructure model with Hawkes Process

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Abstract

Introduction

1 Univariate Model

The likelihood function under univariate Hawkes process model:

$$\begin{aligned} L(\Theta) &= \prod_{i=1}^n f(T_i) \prod_{i=1}^n (1 - F_{T_{i-1}}(T_i))(1 - F_{T_n}(T)) \\ &= \left[\prod_{i=1}^n \lambda(T_i) \right] \left[\prod_{i=1}^n \exp\left(-\int_{T_{i-1}}^{T_i} \lambda(s) ds\right) \right] \exp\left(-\int_{T_n}^T \lambda(t) dt\right) \\ &= \exp\left(-\int_0^T \lambda(t) dt\right) \prod_{i=1}^n \lambda(T_i) \end{aligned}$$

1.1 Exponential Decay Model

Suppose we are going to use exponential decay kernel and constant base intensity function:
 $\phi(t) = \alpha e^{-\beta t}$, $\lambda_0(t) = \mu \rightarrow \lambda(t) = \mu + \sum_{i:t>T_i} \alpha e^{-\beta(t-T_i)}$

$$L(\Theta) = \exp \left(- \int_0^T \mu + \sum_{i:t>T_i} \alpha e^{-\beta(t-T_i)} dt \right) \prod_{i=1}^n (\mu + \sum_{j=1}^{i-1} \alpha e^{-\beta(T_i-T_j)})$$

$$\ln L(\Theta) = \sum_{i=1}^n \ln \left(\mu + \alpha \sum_{j=1}^{i-1} e^{-\beta(T_i-T_j)} \right) - \mu T - \alpha \int_0^T \sum_{i:t>T_i} e^{-\beta(t-T_i)} dt$$

Differentiate with respect to μ :

$$\frac{\partial}{\partial \mu} \ln L(\Theta) = \sum_{i=1}^n \frac{1}{\mu + \alpha \sum_{j=1}^{i-1} e^{-\beta(T_i-T_j)}} - T = 0$$

Differentiate with respect to α :

$$\frac{\partial}{\partial \alpha} \ln L(\Theta) = \sum_{i=1}^n \frac{\sum_{j=1}^{i-1} e^{-\beta(T_i-T_j)}}{\mu + \alpha \sum_{j=1}^{i-1} e^{-\beta(T_i-T_j)}} - \int_0^T \sum_{i:t>T_i} e^{-\beta(t-T_i)} dt = 0$$

Differentiate with respect to β :

$$\frac{\partial}{\partial \beta} \ln L(\Theta) = \frac{-\alpha \sum_{j=1}^{i-1} (T_i - T_j) e^{-\beta(T_i-T_j)}}{\mu + \alpha \sum_{j=1}^{i-1} e^{-\beta(T_i-T_j)}} - \alpha \int_0^T \sum_{i:t>T_i} -(t - T_i) e^{-\beta(t-T_i)} dt = 0$$

Below are the three equations we need to solve to arrive at the values of μ , α and β .

$$\sum_{i=1}^n \frac{1}{\mu + \alpha \sum_{j=0}^i e^{-\beta(T_i-T_j)}} = T \quad (1)$$

$$\sum_{i=1}^n \frac{\sum_{j=1}^{i-1} e^{-\beta(T_i-T_j)}}{\mu + \alpha \sum_{j=1}^{i-1} e^{-\beta(T_i-T_j)}} = \int_0^T \sum_{i:t>T_i} e^{-\beta(t-T_i)} dt \quad (2)$$

$$\sum_{i=1}^n \frac{\sum_{j=1}^{i-1} (T_i - T_j) e^{-\beta(T_i-T_j)}}{\mu + \alpha \sum_{j=1}^{i-1} e^{-\beta(T_i-T_j)}} = \int_0^T \sum_{i:t>T_i} (t - T_i) e^{-\beta(t-T_i)} dt \quad (3)$$

1.2 Univariate Power Law Kernel

Substituting $\phi(t) = \frac{\alpha}{(1+\beta t)^{1+\gamma}}$ in:

$$\begin{aligned}
L(\Theta) &= \exp\left(-\int_0^T \lambda(t \mid F_t) dt\right) \prod_{i=1}^n \lambda(T_i \mid F_{T_i}) \\
&= \exp\left(-\int_0^T \mu + \sum_{i:t>T_i} \frac{\alpha}{(1+\beta(t-T_i))^{1+\gamma}} dt\right) \prod_{i=1}^n \left(\mu + \sum_{j=0}^i \frac{\alpha}{(1+\beta(T_i-T_j))^{1+\gamma}}\right) \\
\ln L(\Theta) &= \sum_{i=1}^n \ln\left(\mu + \alpha \sum_{j=0}^i \frac{1}{(1+\beta(T_i-T_j))^{1+\gamma}}\right) - \mu T - \alpha \int_0^T \sum_{i:t>T_i} \frac{1}{(1+\beta(t-T_i))^{1+\gamma}} dt
\end{aligned}$$

Differentiate with respect to μ :

$$\frac{\partial}{\partial \mu} \ln L(\Theta) = \sum_{i=1}^n \frac{1}{\mu + \alpha \sum_{j=0}^i \frac{1}{(1+\beta(T_i-T_j))^{1+\gamma}}} - T = 0$$

Differentiate with respect to α :

$$\frac{\partial}{\partial \alpha} \ln L(\Theta) = \sum_{i=1}^n \frac{\sum_{j=0}^i \frac{1}{(1+\beta(T_i-T_j))^{1+\gamma}}}{\mu + \alpha \sum_{j=0}^i \frac{1}{(1+\beta(T_i-T_j))^{1+\gamma}}} - \int_0^T \sum_{i:t>T_i} \frac{1}{(1+\beta(t-T_i))^{1+\gamma}} dt = 0$$

Differentiate with respect to β :

$$\frac{\partial}{\partial \beta} \ln L(\Theta) = \sum_{i=1}^n \frac{\alpha \sum_{j=0}^i \frac{T_i-T_j}{(1+\beta(T_i-T_j))^{2+\gamma}}}{\mu + \alpha \sum_{j=0}^i \frac{1}{(1+\beta(T_i-T_j))^{1+\gamma}}} - \alpha \int_0^T \sum_{i:t>T_i} \frac{t-T_i}{(1+\beta(t-T_i))^{2+\gamma}} dt = 0$$

According to Bacry¹'s research, the empirical kernel is closed to the power law kernel when the parameter γ is closed to 0. Hence we can reduce the equations to these three below to solve for μ , α and β .

$$\sum_{i=1}^n \frac{1}{\mu + \alpha \sum_{j=0}^i \frac{1}{1+\beta(T_i-T_j)}} = T \tag{4}$$

$$\sum_{i=1}^n \frac{\sum_{j=0}^i \frac{1}{1+\beta(T_i-T_j)}}{\mu + \alpha \sum_{j=0}^i \frac{1}{1+\beta(T_i-T_j)}} = \int_0^T \sum_{i:t>T_i} \frac{1}{1+\beta(t-T_i)} dt \tag{5}$$

$$\sum_{i=1}^n \frac{\sum_{j=0}^i \frac{T_i-T_j}{(1+\beta(T_i-T_j))^2}}{\mu + \alpha \sum_{j=0}^i \frac{1}{1+\beta(T_i-T_j)}} = \int_0^T \sum_{i:t>T_i} \frac{t-T_i}{(1+\beta(t-T_i))^2} dt \tag{6}$$

2 Bivariate Model

Likelihood function of the multivariate Hawkes process model with D variables:

$$\begin{aligned}
L(\Theta) &= \prod_{i=1}^D \left[\prod_{j=1}^{n_i} f(T_{i,j}) \prod_{j=1}^{n_i} (1 - F_{T_{i,j-1}}(T_{i,j})) (1 - F_{T_{i,n_i}}(T)) \right] \\
&= \prod_{i=1}^D \left[\prod_{j=1}^{n_i} \lambda_i(T_{i,j}) \left(\prod_{j=1}^{n_i} \exp\left(-\int_{T_{i,j-1}}^{T_{i,j}} \lambda_i(s) ds\right) \right) \exp\left(-\int_{T_{i,n_i}}^T \lambda_i(s) ds\right) \right] \\
&= \prod_{i=1}^D \left[\exp\left(-\int_0^T \lambda_i(s) ds\right) \prod_{j=1}^{n_i} \lambda_i(T_{i,j}) \right] \\
\ln L(\Theta) &= \sum_{i=1}^D \left[\sum_{j=1}^{n_i} \ln \lambda_i(T_{i,j}) - \int_0^T \lambda_i(s) ds \right]
\end{aligned}$$

Likelihood function of the bivariate Hawkes process model is one with $D = 2$.

2.1 Bivariate Exponential Kernel

$\phi^{1,1}(t) = \phi^{2,2}(t) = \phi^s(t)$, $\phi^{1,2}(t) = \phi^{2,1}(t) = \phi^c(t)$, $\lambda_{i,0}(t) = \mu_i$
 $\rightarrow \lambda_i(t) = \mu_i + \sum_{j=1}^2 \sum_{k:t>T_{j,k}} \phi^{i,j}(t)$ where $\phi^s(t) = \alpha^s e^{-\beta^s t}$, $\phi^c(t) = \alpha^c e^{-\beta^c t}$

Partial differentiations of ϕ are: $\frac{\partial \phi(t)}{\partial \alpha} = e^{-\beta t}$, $\frac{\partial \phi(t)}{\partial \beta} = \alpha e^{-\beta t} t = \phi(t)t$

$$\begin{aligned} \ln L(\Theta) &= \sum_{i=1}^2 \left[\sum_{j=1}^{n_i} \ln \lambda_i(T_{i,j}) - \int_0^T \lambda_i(s) ds \right] \\ &= \sum_{j=1}^{n_1} \ln \lambda_1(T_{1,j}) - \int_0^T \lambda_1(t) dt + \sum_{j=1}^{n_2} \ln \lambda_2(T_{2,j}) - \int_0^T \lambda_2(t) dt \\ \ln L(\Theta) &= \sum_{j=1}^{n_1} \ln \left(\mu_1 + \sum_{k=1}^{j-1} \phi^s(T_{1,j} - T_{1,k}) + \sum_{k:T_{1,j}>T_{2,k}} \phi^c(T_{1,j} - T_{2,k}) \right) \\ &\quad - \mu_1 T - \int_0^T \sum_{k:t>T_{1,k}} \phi^s(t - T_{1,k}) + \sum_{k:t>T_{2,k}} \phi^c(t - T_{2,k}) dt \\ &\quad + \sum_{j=1}^{n_2} \ln \left(\mu_2 + \sum_{k=1}^{j-1} \phi^s(T_{2,j} - T_{2,k}) + \sum_{k:T_{2,j}>T_{1,k}} \phi^c(T_{2,j} - T_{1,k}) \right) \\ &\quad - \mu_2 T - \int_0^T \sum_{k:t>T_{2,k}} \phi^s(t - T_{2,k}) + \sum_{k:t>T_{1,k}} \phi^c(t - T_{1,k}) dt \\ &\quad \text{where } \phi^s(t) = \alpha^s e^{-\beta^s t}, \phi^c(t) = \alpha^c e^{-\beta^c t} \end{aligned}$$

$$\begin{aligned} \frac{\partial}{\partial \mu_1} \ln L(\Theta) &= \sum_{j=1}^{n_1} \left(\mu_1 + \sum_{k=1}^{j-1} \phi^s(T_{1,j} - T_{1,k}) + \sum_{k:T_{1,j}>T_{2,k}} \phi^c(T_{1,j} - T_{2,k}) \right)^{-1} - T = 0 \\ \rightarrow T &= \sum_{j=1}^{n_1} \left(\mu_1 + \sum_{k=1}^{j-1} \phi^s(T_{1,j} - T_{1,k}) + \sum_{k:T_{1,j}>T_{2,k}} \phi^c(T_{1,j} - T_{2,k}) \right)^{-1} \\ \rightarrow T &= \sum_{j=1}^{n_1} \left(\mu_1 + \sum_{k=1}^{j-1} \alpha^s e^{-\beta^s (T_{1,j} - T_{1,k})} + \sum_{k:T_{1,j}>T_{2,k}} \alpha^c e^{-\beta^c (T_{1,j} - T_{2,k})} \right)^{-1} \end{aligned}$$

$$\begin{aligned}
\frac{\partial}{\partial \mu_2} \ln L(\Theta) &= \sum_{j=1}^{n_2} \left(\mu_2 + \sum_{k=1}^{j-1} \phi^s(T_{2,j} - T_{2,k}) + \sum_{k:T_{2,j} > T_{1,k}} \phi^c(T_{2,j} - T_{1,k}) \right)^{-1} - T = 0 \\
\rightarrow T &= \sum_{j=1}^{n_2} \left(\mu_2 + \sum_{k=1}^{j-1} \phi^s(T_{2,j} - T_{2,k}) + \sum_{k:T_{2,j} > T_{1,k}} \phi^c(T_{2,j} - T_{1,k}) \right)^{-1} \\
\rightarrow T &= \sum_{j=1}^{n_2} \left(\mu_2 + \sum_{k=1}^{j-1} \alpha^s e^{-\beta^s(T_{2,j} - T_{2,k})} + \sum_{k:T_{2,j} > T_{1,k}} \alpha^c e^{-\beta^c(T_{2,j} - T_{1,k})} \right)^{-1}
\end{aligned}$$

$$\begin{aligned}
\frac{\partial}{\partial \alpha^s} \ln L(\Theta) &= \sum_{j=1}^{n_1} \frac{\sum_{k=1}^{j-1} \frac{\partial \phi^s(T_{1,j} - T_{1,k})}{\partial \alpha^s}}{\mu_1 + \sum_{k=1}^{j-1} \phi^s(T_{1,j} - T_{1,k}) + \sum_{k:T_{1,j} > T_{2,k}} \phi^c(T_{1,j} - T_{2,k})} \\
&\quad - \int_0^T \sum_{k:t > T_{1,k}} \frac{\partial \phi^s(t - T_{1,k})}{\partial \alpha^s} dt - \int_0^T \sum_{k:t > T_{2,k}} \frac{\partial \phi^s(t - T_{2,k})}{\partial \alpha^s} dt \\
&\quad + \sum_{j=1}^{n_2} \frac{\sum_{k=1}^{j-1} \frac{\partial \phi^s(T_{2,j} - T_{2,k})}{\partial \alpha^s}}{\mu_2 + \sum_{k=1}^{j-1} \phi^s(T_{2,j} - T_{2,k}) + \sum_{k:T_{2,j} > T_{1,k}} \phi^c(T_{2,j} - T_{1,k})} = 0 \\
&\rightarrow \sum_{j=1}^{n_1} \frac{\sum_{k=1}^{j-1} e^{-\beta^s(T_{1,j} - T_{1,k})}}{\mu_1 + \sum_{k=1}^{j-1} \phi^s(T_{1,j} - T_{1,k}) + \sum_{k:T_{1,j} > T_{2,k}} \phi^c(T_{1,j} - T_{2,k})} \\
&\quad + \sum_{j=1}^{n_2} \frac{\sum_{k=1}^{j-1} e^{-\beta^s(T_{2,j} - T_{2,k})}}{\mu_2 + \sum_{k=1}^{j-1} \phi^s(T_{2,j} - T_{2,k}) + \sum_{k:T_{2,j} > T_{1,k}} \phi^c(T_{2,j} - T_{1,k})} \\
&= \int_0^T \sum_{k:t > T_{1,k}} e^{-\beta^s(t - T_{1,k})} dt + \int_0^T \sum_{k:t > T_{2,k}} e^{-\beta^s(t - T_{2,k})} dt
\end{aligned}$$

$$\begin{aligned}
\frac{\partial}{\partial \alpha^c} \ln L(\Theta) &= \sum_{j=1}^{n_1} \frac{\sum_{k:T_{1,j} > T_{2,k}} \frac{\partial \phi^c(T_{1,j} - T_{2,k})}{\partial \alpha^c}}{\mu_1 + \sum_{k=1}^{j-1} \phi^s(T_{1,j} - T_{1,k}) + \sum_{k:T_{1,j} > T_{2,k}} \phi^c(T_{1,j} - T_{2,k})} \\
&\quad - \int_0^T \sum_{k:t > T_{2,k}} \frac{\partial \phi^c(t - T_{2,k})}{\partial \alpha^c} dt - \int_0^T \sum_{k:t > T_{1,k}} \frac{\partial \phi^c(t - T_{1,k})}{\partial \alpha^c} dt \\
&\quad + \sum_{j=1}^{n_2} \frac{\sum_{k:T_{2,j} > T_{1,k}} \frac{\partial \phi^c(T_{2,j} - T_{1,k})}{\partial \alpha^c}}{\mu_2 + \sum_{k=1}^{j-1} \phi^s(T_{2,j} - T_{2,k}) + \sum_{k:T_{2,j} > T_{1,k}} \phi^c(T_{2,j} - T_{1,k})} = 0 \\
&\rightarrow \sum_{j=1}^{n_1} \frac{\sum_{k:T_{1,j} > T_{2,k}} e^{-\beta^c(T_{1,j} - T_{2,k})}}{\mu_1 + \sum_{k=1}^{j-1} \phi^s(T_{1,j} - T_{1,k}) + \sum_{k:T_{1,j} > T_{2,k}} \phi^c(T_{1,j} - T_{2,k})} \\
&\quad + \sum_{j=1}^{n_2} \frac{\sum_{k:T_{2,j} > T_{1,k}} e^{-\beta^c(T_{2,j} - T_{1,k})}}{\mu_2 + \sum_{k=1}^{j-1} \phi^s(T_{2,j} - T_{2,k}) + \sum_{k:T_{2,j} > T_{1,k}} \phi^c(T_{2,j} - T_{1,k})} \\
&= \int_0^T \sum_{k:t > T_{2,k}} e^{-\beta^c(t - T_{2,k})} dt + \int_0^T \sum_{k:t > T_{1,k}} e^{-\beta^c(t - T_{1,k})} dt
\end{aligned}$$

$$\begin{aligned}
\frac{\partial}{\partial \beta^s} \ln L(\Theta) &= \sum_{j=1}^{n_1} \frac{\sum_{k=1}^{j-1} \frac{\partial \phi^s(T_{1,j} - T_{1,k})}{\partial \beta^s}}{\mu_1 + \sum_{k=1}^{j-1} \phi^s(T_{1,j} - T_{1,k}) + \sum_{k:T_{1,j} > T_{2,k}} \phi^c(T_{1,j} - T_{2,k})} \\
&\quad - \int_0^T \sum_{k:t > T_{1,k}} \frac{\partial \phi^s(t - T_{1,k})}{\partial \beta^s} dt - \int_0^T \sum_{k:t > T_{2,k}} \frac{\partial \phi^s(t - T_{2,k})}{\partial \beta^s} dt \\
&\quad + \sum_{j=1}^{n_2} \frac{\sum_{k=1}^{j-1} \frac{\partial \phi^s(T_{2,j} - T_{2,k})}{\partial \beta^s}}{\mu_2 + \sum_{k=1}^{j-1} \phi^s(T_{2,j} - T_{2,k}) + \sum_{k:T_{2,j} > T_{1,k}} \phi^c(T_{2,j} - T_{1,k})} = 0 \\
&\rightarrow \sum_{j=1}^{n_1} \frac{\sum_{k=1}^{j-1} (T_{1,j} - T_{1,k}) \phi^s(T_{1,j} - T_{1,k})}{\mu_1 + \sum_{k=1}^{j-1} \phi^s(T_{1,j} - T_{1,k}) + \sum_{k:T_{1,j} > T_{2,k}} \phi^c(T_{1,j} - T_{2,k})} \\
&\quad + \sum_{j=1}^{n_2} \frac{\sum_{k=1}^{j-1} (T_{2,j} - T_{2,k}) \phi^s(T_{2,j} - T_{2,k})}{\mu_2 + \sum_{k=1}^{j-1} \phi^s(T_{2,j} - T_{2,k}) + \sum_{k:T_{2,j} > T_{1,k}} \phi^c(T_{2,j} - T_{1,k})} \\
&= \int_0^T \sum_{k:t > T_{1,k}} (t - T_{1,k}) \phi^s(t - T_{1,k}) dt + \int_0^T \sum_{k:t > T_{2,k}} (t - T_{2,k}) \phi^s(t - T_{2,k}) dt
\end{aligned}$$

$$\begin{aligned}
\frac{\partial}{\partial \beta^c} \ln L(\Theta) &= \sum_{j=1}^{n_1} \frac{\sum_{k:T_{1,j} > T_{2,k}} \frac{\partial \phi^c(T_{1,j} - T_{2,k})}{\partial \beta^c}}{\mu_1 + \sum_{k=1}^{j-1} \phi^s(T_{1,j} - T_{1,k}) + \sum_{k:T_{1,j} > T_{2,k}} \phi^c(T_{1,j} - T_{2,k})} \\
&\quad - \int_0^T \sum_{k:t > T_{2,k}} \frac{\partial \phi^c(t - T_{2,k})}{\partial \beta^c} dt - \int_0^T \sum_{k:t > T_{1,k}} \frac{\partial \phi^c(t - T_{1,k})}{\partial \beta^c} dt \\
&\quad + \sum_{j=1}^{n_2} \frac{\sum_{k:T_{2,j} > T_{1,k}} \frac{\partial \phi^c(T_{2,j} - T_{1,k})}{\partial \beta^c}}{\mu_2 + \sum_{k=1}^{j-1} \phi^s(T_{2,j} - T_{2,k}) + \sum_{k:T_{2,j} > T_{1,k}} \phi^c(T_{2,j} - T_{1,k})} = 0 \\
&\rightarrow \sum_{j=1}^{n_1} \frac{\sum_{k:T_{1,j} > T_{2,k}} (T_{1,j} - T_{2,k}) \phi^c(T_{1,j} - T_{2,k})}{\mu_1 + \sum_{k=1}^{j-1} \phi^s(T_{1,j} - T_{1,k}) + \sum_{k:T_{1,j} > T_{2,k}} \phi^c(T_{1,j} - T_{2,k})} \\
&\quad + \sum_{j=1}^{n_2} \frac{\sum_{k:T_{2,j} > T_{1,k}} (T_{2,j} - T_{1,k}) \phi^c(T_{2,j} - T_{1,k})}{\mu_2 + \sum_{k=1}^{j-1} \phi^s(T_{2,j} - T_{2,k}) + \sum_{k:T_{2,j} > T_{1,k}} \phi^c(T_{2,j} - T_{1,k})} \\
&= \int_0^T \sum_{k:t > T_{2,k}} (t - T_{2,k}) \phi^c(t - T_{2,k}) dt + \int_0^T \sum_{k:t > T_{1,k}} (t - T_{1,k}) \phi^c(t - T_{1,k}) dt
\end{aligned}$$

References

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