

# Package: hawkes (via r-universe)

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**Title** Hawkes process simulation and calibration toolkit

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**Description** The package allows to simulate Hawkes process both in univariate and multivariate settings. It gives functions to compute different moments of the number of jumps of the process on a given interval, such as mean, variance or autocorrelation of process jumps on time intervals separated by a lag.

**License** GPL (>= 2)

**Depends** R (>= 3.0.2)

**Imports** Rcpp (>= 0.11.1)

**LinkingTo** Rcpp, RcppArmadillo (>= 0.4.100.2.1)

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jumpAutocorrelation     *Autocorrelation of Hawkes process jumps on nonoverlapping time intervals with lag.*

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### Description

The function returns the theoretical autocorrelation of the number of jumps of a Hawkes process on nonoverlapping time intervals with lag

### Usage

```
jumpAutocorrelation(lambda0, alpha, beta, tau, lag)
```

### Arguments

lambda0	Vector of initial intensity, a scalar in the monovariate case.
alpha	Matrix of excitation, a scalar in the monovariate case. Excitation values are all positive.
beta	Vector of betas, a scalar in the monovariate case.
tau	Time interval length.
lag	Time lag.

### Details

Notice that in the scalar case, one must have  $\beta > \alpha$  for the process to be stable, and in the multivariate case, the matrix  $(\text{diag}(\beta) - \alpha)$  must have eigen values with strictly positive real parts for the process to be stable.

### Value

Returns a matrix containing the autocorrelation of the number of jumps of process components.

### References

Jose Da Fonseca and Riadh Zaatour Hawkes Process : Fast Calibration, Application to Trade Clustering and Diffusive Limit. *Journal of Futures Markets*, Volume 34, Issue 6, pages 497-606, June 2014.

Jose Da Fonseca and Riadh Zaatour Clustering and Mean Reversion in Hawkes Microstructure Models.

**Examples**

```
#One dimensional Hawkes process
lambda0<-0.02
alpha<-0.05
beta<-0.06
tau<-60#one minute
lag<-0#adjacent non overlapping intervals
h<-jumpAutocorrelation(lambda0,alpha,beta,tau,lag)

#Multivariate Hawkes process
lambda0<-c(0.02,0.02)
alpha<-matrix(c(0.05,0,0,0.05),byrow=TRUE,nrow=2)
beta<-c(0.06,0.06)
tau<-60#one minute
lag<-0#adjacent non overlapping intervals
h<-jumpAutocorrelation(lambda0,alpha,beta,tau,lag)
```

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jumpMean	<i>Mean of Hawkes process jumps.</i>
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**Description**

The function returns the theoretical mean of the number of jumps of a Hawkes process on a time interval of length tau.

**Usage**

```
jumpMean(lambda0, alpha, beta, tau)
```

**Arguments**

lambda0	Vector of initial intensity, a scalar in the monovariate case.
alpha	Matrix of excitation, a scalar in the monovariate case. Excitation values are all positive.
beta	Vector of betas, a scalar in the monovariate case.
tau	Time interval length.

**Details**

Notice that in the scalar case, one must have  $\beta > \alpha$  for the process to be stable, and in the multivariate case, the matrix  $(\text{diag}(\beta) - \alpha)$  must have eigen values with strictly positive real parts for the process to be stable.

**Value**

Returns a vector containing the mean number of jumps of every process component.

## References

Jose Da Fonseca and Riadh Zaatour Hawkes Process : Fast Calibration, Application to Trade Clustering and Diffusive Limit. *Journal of Futures Markets*, Volume 34, Issue 6, pages 497-606, June 2014.

Jose Da Fonseca and Riadh Zaatour Clustering and Mean Reversion in Hawkes Microstructure Models.

## Examples

```
#One dimensional Hawkes process
lambda0<-0.02
alpha<-0.05
beta<-0.06
tau<-60#one minute
h<-jumpMean(lambda0,alpha,beta,tau)

#Multivariate Hawkes process
lambda0<-c(0.02,0.02)
alpha<-matrix(c(0.05,0,0,0.05),byrow=TRUE,nrow=2)
beta<-c(0.06,0.06)
tau<-60#one minute
h<-jumpMean(lambda0,alpha,beta,tau)
```

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jumpVariance

*Variance of Hawkes process jumps.*

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## Description

The function returns the theoretical variance matrix of the number of jumps of a Hawkes process on a time interval of length tau.

## Usage

```
jumpVariance(lambda0, alpha, beta, tau)
```

## Arguments

lambda0	Vector of initial intensity, a scalar in the monovariate case.
alpha	Matrix of excitation, a scalar in the monovariate case. Excitation values are all positive.
beta	Vector of betas, a scalar in the monovariate case.
tau	Time interval length.

## Details

Notice that in the scalar case, one must have  $\beta > \alpha$  for the process to be stable, and in the multivariate case, the matrix  $(\text{diag}(\beta) - \alpha)$  must have eigen values with strictly positive real parts for the process to be stable.

**Value**

Returns a matrix containing the variance of the number of jumps of every process component.

**References**

Jose Da Fonseca and Riadh Zaatour Hawkes Process : Fast Calibration, Application to Trade Clustering and Diffusive Limit. *Journal of Futures Markets*, Volume 34, Issue 6, pages 497-606, June 2014.

Jose Da Fonseca and Riadh Zaatour Clustering and Mean Reversion in Hawkes Microstructure Models.

**Examples**

```
#One dimensional Hawkes process
lambda0<-0.02
alpha<-0.05
beta<-0.06
tau<-60#one minute
h<-jumpVariance(lambda0,alpha,beta,tau)

#Multivariate Hawkes process
lambda0<-c(0.02,0.02)
alpha<-matrix(c(0.05,0,0,0.05),byrow=TRUE,nrow=2)
beta<-c(0.06,0.06)
tau<-60#one minute
h<-jumpVariance(lambda0,alpha,beta,tau)
```

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likelihoodHawkes

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*Compute the likelihood function of a hawkes process*


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**Description**

Compute the likelihood function of a hawkes process for the given parameter and given the jump times vector (or list of vectors in the multivariate case), and until a time horizon.

**Usage**

```
likelihoodHawkes(lambda0, alpha, beta, history)
```

**Arguments**

lambda0	Vector of initial intensity, a scalar in the monovariate case.
alpha	Matrix of excitation, a scalar in the monovariate case. Excitation values are all positive.
beta	Vector of betas, a scalar in the monovariate case.
history	Jump times vector (or list of vectors in the multivariate case).

**Value**

Returns the opposite of the likelihood.

**References**

Y. Ogata. (1981) On Lewis simulation method for point processes. *IEEE Transactions on Information Theory*, **31**

**Examples**

```
#One dimensional Hawkes process
lambda0<-0.2
alpha<-0.5
beta<-0.7
history<-simulateHawkes(lambda0,alpha,beta,3600)
l<-likelihoodHawkes(lambda0,alpha,beta,history[[1]])

#Multivariate Hawkes process
lambda0<-c(0.2,0.2)
alpha<-matrix(c(0.5,0,0,0.5),byrow=TRUE,nrow=2)
beta<-c(0.7,0.7)
history<-simulateHawkes(lambda0,alpha,beta,3600)
l<-likelihoodHawkes(lambda0,alpha,beta,history)
```

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simulateHawkes	<i>Hawkes process simulation Function</i>
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**Description**

The function simulates a Hawkes process for the given parameter, and until a time horizon.

**Usage**

```
simulateHawkes(lambda0, alpha, beta, horizon)
```

**Arguments**

lambda0	Vector of initial intensity, a scalar in the monovariate case.
alpha	Matrix of excitation, a scalar in the monovariate case. Excitation values are all positive.
beta	Vector of betas, a scalar in the monovariate case.
horizon	Time horizon until which the simulation is to be conducted.

**Details**

Notice that in the scalar case, one must have  $\beta > \alpha$  for the process to be stable, and in the multivariate case, the matrix  $(\text{diag}(\beta) - \alpha)$  must have eigen values with strictly positive real parts for the process to be stable.

**Value**

Returns a vector of jump times in the univariate case, and a list of such vectors for every component in the multivariate case.

**References**

Y. Ogata. (1981) On Lewis simulation method for point processes. *IEEE Transactions on Information Theory*, **31**

**Examples**

```
#One dimensional Hawkes process
lambda0<-0.2
alpha<-0.5
beta<-0.7
horizon<-3600#one hour
h<-simulateHawkes(lambda0,alpha,beta,horizon)

#Multivariate Hawkes process
lambda0<-c(0.2,0.2)
alpha<-matrix(c(0.5,0,0,0.5),byrow=TRUE,nrow=2)
beta<-c(0.7,0.7)
horizon<-3600#one hour
h<-simulateHawkes(lambda0,alpha,beta,horizon)
```

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