

# Package ‘seismic’

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**Type** Package

**Title** Predict Information Cascade by Self-Exciting Point Process

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**Description**

An implementation of self-exciting point process model for information cascades, which occurs when many people engage in the same acts after observing the actions of others (e.g. post resharing on Facebook or Twitter). It provides functions to estimate the infectiousness of an information cascade and predict its popularity given the observed history. See <<http://snap.stanford.edu/seismic/>> for more information and datasets.

**URL** <http://snap.stanford.edu/seismic/>

**License** GPL-3

**NeedsCompilation** no

**RoxygenNote** 7.1.2

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`get.infectiousness`      *Estimate the infectiousness of an information cascade*

## Description

Estimate the infectiousness of an information cascade

## Usage

```
get.infectiousness(
  share.time,
  degree,
  p.time,
  max.window = 2 * 60 * 60,
  min.window = 300,
  min.count = 5
)
```

## Arguments

<code>share.time</code>	observed resharing times, sorted, <code>share.time[1] = 0</code>
<code>degree</code>	observed node degrees
<code>p.time</code>	equally spaced vector of time to estimate the infectiousness, <code>p.time[1]=0</code>
<code>max.window</code>	maximum span of the locally weight kernel
<code>min.window</code>	minimum span of the locally weight kernel
<code>min.count</code>	the minimum number of reshарings included in the window

## Details

Use a triangular kernel with shape changing over time. At time `p.time`, use a triangluer kernel with slope =  $\min(\max(1/(p.time/2), 1/min.window), \text{max.window})$ .

## Value

a list of three vectors:

- `infectiousness`. the estimated infectiousness
- `p.up`. the upper 95 percent approximate confidence interval
- `p.low`. the lower 95 percent approximate confidence interval

## Examples

```
data(tweet)
pred.time <- seq(0, 6 * 60 * 60, by = 60)
infectiousness <- get.infectiousness(tweet[, 1], tweet[, 2], pred.time)
plot(pred.time, infectiousness$infectiousness)
```

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pred.cascade	<i>Predict the popularity of information cascade</i>
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## Description

Predict the popularity of information cascade

## Usage

```
pred.cascade(  
  p.time,  
  infectiousness,  
  share.time,  
  degree,  
  n.star = 100,  
  features.return = FALSE  
)
```

## Arguments

p.time	equally spaced vector of time to estimate the infectiousness, p.time[1]=0
infectiousness	a vector of estimated infectiousness, returned by <a href="#">get.infectiousness</a>
share.time	observed resharing times, sorted, share.time[1] =0
degree	observed node degrees
n.star	the average node degree in the social network
features.return	if TRUE, returns a matrix of features to be used to further calibrate the prediction

## Value

a vector of predicted populatir at each time in p.time.

## Examples

```
data(tweet)  
pred.time <- seq(0, 6 * 60 * 60, by = 60)  
infectiousness <- get.infectiousness(tweet[, 1], tweet[, 2], pred.time)  
pred <- pred.cascade(pred.time, infectiousness$infectiousness, tweet[, 1], tweet[, 2], n.star = 100)  
plot(pred.time, pred)
```

seismic

*Predicting information cascade by self-exciting point process model***Description**

This package implements a self-exciting point process model for information cascades. An information cascade occurs when many people engage in the same acts after observing the actions of others. Typical examples are post/photo resharing on Facebook and retweets on Twitter. The package provides functions to estimate the infectiousness of an information cascade and predict its popularity given the observed history. For more information, see <http://snap.stanford.edu/seismic/>.

**References**

SEISMIC: A Self-Exciting Point Process Model for Predicting Tweet Popularity by Q. Zhao, M. Erdogdu, H. He, A. Rajaraman, J. Leskovec, ACM SIGKDD Conference on Knowledge Discovery and Data Mining (KDD), 2015.

tweet

*An example information cascade***Description**

A dataset containing all the (relative) resharing time and node degree of a tweet. The original Twitter ID is 127001313513967616.

**Format**

A data frame with 15563 rows and 2 columns

**Details**

- relative\_time\_second. resharing time in seconds
- number\_of\_followers. number of followers

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