

Large variance and fat tail of damage by natural disaster

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Welcome to ASCE-ICVRAM-ISUMA 2014





Environmental Change Institute



VIRTUAL ENGINEERING CENTRE

ASCE-ICVRAM-ISUMA provides a multidisciplinary forum for the exchange of knowledge and expertise, in the Quantification, Mitigation and Management of Risk and Uncertainty, and in Decision Making.

The event is aimed at specialised and synergetic developments in both theory and practice.

This joint conference is being organised by the Institute for Risk & Uncertainty and in collaboration with the ASCE Council on Disaster Risk Management, the Environmental Change Institute (University of Oxford) and the Virtual Engineering Centre.

The proceedings of the conference are now available: <u>ASCE library</u>.



ICVRAM2014 Conference timeline:

30th November 2013: ASCE Full paper submission

15th February 2014: Deposit and/or early bird registration deadline

1st of May 2014: Full paper submission (not included in the ASCE proceedings)

1st June 2014: Scientific programme

13th-16th July 2014: Conference

View photos of the ASCE-ICVRAM-ISUMA 2014



Extreme events occur!

Statistics of damage

Comparison of United States Natural Disaster Fatalities



Nishenko & Barton, USGS (1995)

Statistics of damage



Figure 4. Empirical damage distributions by tornadoes in the United States from 1970 to 2011 and estimated power-law exponents (a) for the numbers of death and injured and (b) for property and crop damages estimated in US dollars.

Extreme events occur!

damage = natural disaster × population/property × vulnerability

Stromberg, J. Econ. Perspect. (2007)

Assumption 1a

Population/property is power-law distributed.



Fig. 1. Power-law exponents and goodness-of-fit tests for wealth data. Note: Vertical bars show 95% confidence intervals.



Assumption 1b

Population/property is spatially correlated.



Figure 1. (a) Random, (b) correlated, and (c) concentrated configurations of values on a two-dimensional lattice of size 100×100 , where the probability density function of value v follows a power-law as $P(v) \sim v^{-\alpha}$ with $\alpha = 2.5$. The height at each site represents a logarithm of the value.

Assumption 2

• Natural disaster is power-law distributed.



Clauset et al., SIAM Rev. (2009)

Assumption 3

• Vulnerability is constant.

Question

- Population/property: power-law exponent a, spatial correlation c
- Natural disaster: power-law with exponent
- Damage: power-law with exponent y

 $\gamma(\alpha, \beta, c)?$

Model

- Each site on 2D lattice: a value v from P(v)~v^{-a}
- Spatial correlation c adjusted by simulated annealing: 0 ≤ c ≤ 1
- A tornado at a random site moves in a random direction on a line with length I from P(I)~I^{-β}
- A damage is the sum of values on the line:

$$D(i_0, j_0, l) = A \sum_{i=i_0}^{i_0+l-1} v_{i,j_0}$$

Random config.

• Totally uncorrelated case (c=0):

 $D = \sum_{n=1}^{l} v_n \qquad \begin{array}{c} P(v) \sim v^{-\alpha} \\ P(l) \sim l^{-\beta} \end{array} \quad \text{Jo et al., PRE (2013)} \end{array}$



$$P(D) \sim D^{-\alpha} + D^{-\beta} + D^{-(\alpha-1)(\beta-1)-1}$$

~ $D^{-\gamma}$,

 $\gamma = \min\{(\alpha - 1)(\beta - 1) + 1, \alpha, \beta\}$

Cf. Contextual bursts

 $P(l) \sim l^{-\alpha}$ collective real inter-event time l В В C A Α A AB С В naucontextual ordinal inter-event time contextual real inter-event time $P(n) \sim n^{-\beta}$ $P(\tau) \sim \tau^{-\alpha'}$

$$\tau = \sum_{i=1}^{n} l_i \quad \Longrightarrow \quad \alpha' = \min\{(\alpha - 1)(\beta - 1) + 1, \alpha, \beta\}$$

$$\tau = \sum_{i=1}^{n} l_i \qquad P(\tau) = \sum_{n=1}^{\infty} P(n) F_n(\tau)$$
$$F_n(\tau) = \prod_{i=1}^{n} \int_{l_0}^{\infty} dl_i P(l_i) \delta\left(\tau - \sum_{i=1}^{n} l_i\right)$$

 l_i au

→ "partition function" of mass transport models

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Majumdar et al., PRL (2005)
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$$n = #$$
 sites (system size)
 $l_i = #$ particles at site *i*
 $T = total # particles$

Evans & Hanney, JPA (2005)

Random config.





Concentrated config.

• Totally correlated case (c=1):

$$v(r) \simeq v_0 r^{-\mu}, \ \mu = \frac{2}{\alpha - 1}$$



$$D(r_0, \theta_0, l, \phi) \simeq w v_0 \int_0^l r(t)^{-\mu} dt,$$

$$r(t) = \sqrt{(r_0 \cos \theta_0 + t \cos \phi)^2 + (r_0 \sin \theta_0 + t \sin \phi)^2},$$

$$\gamma = \min\{\alpha,\beta\}$$

Concentrated config.





Correlated config.



$$D(i_0, j_0, l) = A \sum_{i=i_0}^{i_0+l-1} v_{i,j_0}$$

- If $A(v) \propto v^\eta$, the effective v is $\, v' \propto v^{1+\eta}$
- The positive η : cascading due to the concentration
- The negative η : robust design of buildings etc.

Remarks

- The risk analysis based on the thin-tailed distributions must be improved.
- More realistic simulation and theory?

Social Connectome

This is a satellite meeting of <u>NetSci2016</u>.

Date: May 30, 2016 (Monday) Venue: Dongkang D, The K-Hotel, Seoul, Korea

Social Connectome: The anatomy of social networks and its modeling

Understanding the structure and dynamics of social networks is crucial for investigating social j interactions between human individuals. Recently, the access to large-scale datasets on human societal level, or Big Data, has helped to reveal a number of common features or stylized facts f social networks. However, we still do not have a complete map of social interactions due to sam addition, many large-scale datasets are obtained from only one communication channel, while many different channels. Moreover, the fact that social networks are continuously evolving mak channel picture rather difficult. Hence, we propose the notion of Social Connectome that explor social interactions at a societal level as precisely as possible, given privacy and other constraint the Connectome in the neuroscience aiming to make a complete map of neural connections in a only for revealing and understanding various complex social phenomena, but also for possible policy-making. For our satellite meeting, we invite the world-class researchers in the field to di social network analysis and modeling as follows:

- social networks as a temporal network: empirical analyses and modeling
- multi-layer nature of social networks and overlapping communities
- sampling issue: how sampling only one channel may lead to a bias in network quantities
- applications to engineering and policy-making.

Organizers

Hang-Hyun Jo (POSTECH, Chair) Woo-Sung Jung (POSTECH) Nobuyasu Ito (University of Tokyo & RIKEN) Jari Saramäki (Aalto University)

Confirmed invited speakers (alphabetical order)

Yong-Yeol Ahn (Indiana University Bloomington, USA) Meeyoung Cha (KAIST, Korea & Facebook, USA) Petter Holme (SKKU, Korea) Hang-Hyun Jo (POSTECH, Korea) Márton Karsai (ENS Lyon, France) Kimmo Kaski (Aalto University, Finland) János Kertész (CEU, Hungary) Pádraig Mac Carron (University of Oxford, UK) Yohsuke Murase (RIKEN, Japan) Takashi Shimada (University of Tokyo, Japan)