Multivariate Heavy Tailed Phenomena: Modeling, Diagnostics and Applications

Sidney Resnick School of Operations Research and Information Engineering Rhodes Hall, Cornell University Ithaca NY 14853 USA

> http://people.orie.cornell.edu/sid sir1@cornell.edu

MURI Natick Nov 21, 2016

November 15, 2016

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Legs of the Stool

- 1. Mathematical modeling.
- 2. Data exploration, inference and network analysis.
- 3. Scheduling and control in multivariate heavy tailed environments.
- 4. Software.



1. Mathematical modeling

- Network growth under preferential attachment. (UMass, Columbia, Cornell)
- Generating mechanisms for multivariate heavy tails. (UMass, Cornell)
- Models for mobility, eg wireless networks. (OSU)
- Competition models under cumulative advantage; related to preferential attachment and generates games with heavy tailed durations. Cumulative advantage like preferential attachment may be linear or non-linear function of *fitness*. (UMass)
- Models to aid visual search. (UMass, Illinois)
 - Searching for suspects from a huge number of images.
 - Matching images when resolution is low in one image.
- Semi-parametric and asymptotic techniques for risk estimation. (Cornell)
- Methods for dimension reduction for multivariate heavy tailed data (ICA, PCA, ...). (Columbia, American)

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2. Data exploration, inference and network analysis.

• Sampling issues: how to sample large networks under budget constraints to achieve estimation of summary quantities? (UMass, Columbia, Cornell, Minn)





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- Algorithms for matching large networks. (Illinois, UMass, Mn)
 - When are two networks similar?
 - Heavy tailed node degrees help the algorithm; match large degree nodes first.







Software.







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- Calibration, simulation of a preferential attachment model. (UMass, Columbia, Cornell)
 - Efficiency loss from estimation using one snapshot vs network evolution;
 - stationary vs changepoint analysis;
 - asymptotic semiparametric methods (more robust) vs MLE parametric methods (more fragile but more accurate).



• Methods for exploring reciprocity (% of node pairs with bi-directional edges) in large directed graphs. How do reciprocal node pairs affect evolution and social structure? They form a core subgraph to which other nodes are attached. (Minn, UMass, Cornell)



- Mining geoMobility data (wireless users, bike-share, ambulance movement). Framework for study: EPIC. (Minn)
- Inference for heavy tail data (Columbia, Cornell)
 - Large values may be missing, lost, not-recorded.
 - Large values may be censored by rules. Must still do infence.
 - Limitations of minimum distance (Clausett) methods for threshold selection to decide what data has information about the tail.



3. Scheduling and control in multivariate heavy tailed environments.

- Emulating round robin scheduling in wireless networks. (Illinois)
 - Used for wireline where it is not sensitive to distributional properties of the file size distribution.
 - Military scenarios.
 - Goals to achieve: fairness, good throughput, small delay.



- Massive data analysis is often needed to prepare for battle, to track adversaries, and for other military applications.
- Routing algorithm that is insensitve to service time distribution.
- Goals: Fast data retrieval, fast computation.
- Influence propagation in evolving networks. (Minn, OSU)





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- Mobile data offloading: (OSU)
 - Cellular networks often highly constrained.
 - Suggests offloading some traffic to wifi or wired LANS.
- Context aware application scheduling for increasing battery lifetime and improving application response times in smartphones. (OSU)
 - Probability of launching an app follows Zipf's law.
 - The inter-run time and run time of an app follow multivariate heavy tail law.
 - Which apps held in memory and which should be closed?

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4. Software.

Nolan (American) packages in R environment posted on CRAN (Comprehensive R Archive Network) at https://cran.r-project.org.

- 1. **mvmesh.** a package to define and work with MultiVariate Meshes in n dimensions.
 - Define common shapes hollow and solid spheres, simplices, rectangles, and tubes.
 - Plot these objects in 2 and 3 dimensions.
 - Define and manipulate shapes in higher dimensions.
 - Multivariate histogram functions
 - Count the number of data points in partitions of any of the above shapes,
 - Directional histograms that tally how many points are in a list of cones.





- 2. SphericalCubature. a package to evaluate integrals over spheres in n dimensions. Gives exact formulas for polynomial integrands and adaptive methods for general functions.
- **3.** SimplicialCubature. a package to evaluate integrals over n dimensional simplices. Gives exact formulas for polynomial integrands and adaptive methods for general functions.
- 4. gensphere. a package to define probability distributions that have level sets that are all scaled versions of a fixed contour.
 - Specify a large family of contours.
 - Define a probability distribution in terms of that contour and a radial decay function.
 - Arose out of previous Natick visit over attempts to model fragment dispersion from explosions.



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- 4. ecdfHT. a package for computing and plotting a transformed empirical cumulative distribution function (ecdf) for heavy tailed data.
 - Uses log-log transform with extremes pulled in to get visual diagnostic for heavy tails.
 - If plot suggests power law behavior, there are functions to estimate the exponent of the decay and the scale constant.
 - Multivariate generalizations are developed.
- 5. mvevd. (Under development:) Implements several dense classes of multivariate extreme value distributions.
 - discrete angular measure,
 - generalized logistic,
 - piecewise constant and linear angular measures.
 - Uses tools from the mvmesh and SimplicialCubature packages to work in dimension n > 2.

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