

Going to extremes to predict natural disasters

July 10 2017

Predicting natural disasters remains one of the most challenging problems in simulation science because not only are they rare but also because only few of the millions of entries in datasets relate to extreme events. A systematic method for comparing the accuracy of different types of simulation models for such prediction problems has recently been developed by a working group at KAUST.

"Extreme events like dust storms, droughts and floods can affect communities and damage infrastructure," said Sabrina Vettori, a doctoral student cosupervised by Distinguished Professor Marc Genton and Assistant Professor Raphaël Huser of Applied Mathematics and Computational Science. "Modeling and forecasting extremes is very challenging and requires flexible, yet interpretable, models with sound theoretical underpinning—criteria that are exponentially more difficult to meet as the data dimensionality increases," she explained.

Increasing the dimensionality or number of observation variables (like temperature and wind speed) dramatically increases the predictive power of a simulation model, but the statistical dexterity needed to correctly pick out and predict the combination of conditions leading to extreme events is immense.

"We are exploring the boundaries of extreme value theory," said Genton. "The aim of our work is to provide a greater understanding of the performance of existing estimators for modeling extreme events over multiple variables and to develop a new statistical [method](#) for

nonparametric estimation in higher dimensions."

Multivariate simulations generally follow one of two approaches. The first are parametric approaches that configure the model by using a set of variables to best approximate the behavior described by the data. The second are nonparametric approaches, which are statistical methods that fit a function to data but use no underlying assumptions or constraints.

Both approaches have pros and cons, and the best method depends on the application," said Huser.

"Nonparametric methods are typically more flexible than parametric methods, making them less prone to bias, but they are usually limited to small dimensions," explained Huser. "Parametric methods can be applied to much higher-dimensional problems, such as spatial applications with data recorded at a large number of monitoring sites, but are sensitive to errors in the underlying parameters and assumptions.

In their research, the team developed a computational tool to implement nonparametric methods and conducted a vast and systematic [simulation](#) to compare nonparametric and parametric estimator performance in up to five dimensions under various scenarios. These methods provided significant insight into higher dimensional settings.

"These estimators can be used to better [model](#) the location and magnitude of [extreme events](#) and to assist in risk assessment and the identification of trends and variability estimates," said Vettori.

More information: Sabrina Vettori et al, A comparison of dependence function estimators in multivariate extremes, *Statistics and Computing* (2017). [DOI: 10.1007/s11222-017-9745-7](https://doi.org/10.1007/s11222-017-9745-7)

Provided by King Abdullah University of Science and Technology

Citation: Going to extremes to predict natural disasters (2017, July 10) retrieved 16 April 2024 from <https://phys.org/news/2017-07-extremes-natural-disasters.html>

This document is subject to copyright. Apart from any fair dealing for the purpose of private study or research, no part may be reproduced without the written permission. The content is provided for information purposes only.