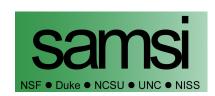
### Flooding to Financial Disaster: An Introduction to Extreme Value Theory

### **Grant Weller** Department of Statistics Colorado State University

Joint work with: Dan Cooley, CSU Steve Sain, NCAR







### Extreme Events

Q: What does it mean to be an 'extreme event'?

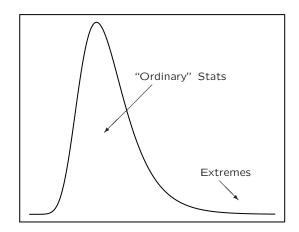
Q: What does it mean to be an 'extreme event'?

A: It depends whom you ask:

- Financial analysts: 'shocks' to a time series (market crash)
- Insurance companies: costly (though perhaps not rare) phenomena (hurricanes)
- Meteorologists: rare weather phenomena (a brown Christmas in Moorhead?)
- Applied mathematicians: extremes of Gaussian/other processes, large deviations
- Statisticians (like me!): Extreme Value Theory (this talk: data perspective)
- Many possible answers!

"Ordinary" Statistics: try to describe a distribution of data; possibly ignore very large or small values (outliers)

*Extreme Value Theory:* try to characterize the tail of the distribution; uses only the extreme observations



While infrequent, extremes often have large human impact.

*Goal of an extreme value analysis:* to quantify the magnitude of a <del>worst-case</del> really-bad-case scenario.

Application areas:

- hydrology (stream/river flows)
- climate variables: precipitation, wind, heat waves, ...
- finance
- insurance/reinsurance
- engineering (structural design, failure)
- not much done (yet) in medicine, biology, ecology

### Why study extremes?

In fact, you don't have to go very far to find an example...



Larry Hansen; Fargo Flood homepage

2009 Red River of the North flood

- Highest ever recorded water level at Fargo (40.8 ft on 3/28/09) flood stage is 18 ft
- Tens of millions of dollars in damages (1997: \$3.5 billion)
- No classes at Concordia for two weeks!
- Related question: how unlikely was this event?
- We'll come back to this later...

# ...first, let's go back in time

#### Concordia College 2004-2008 Cobber football

- 2007 school records for total points, touchdowns, yards in a season
- 2005 & 2007 teams #1 and #2 in season rush yds and tds



#### Mathematics & Economics major

- Advised by Dr. Zeng
- Other influences: Dr. Doug Anderson, Dr. Jim Forde, Dr. Haimeng Zhang, Dr. Dan Biebighauser



I had the best undergraduate advisor!

2007 vs. Carleton

# Then on to graduate school...

### Department of Statistics, CSU Fort Collins, 2008-present

- M.S. in Statistics, 2010
- Ph.D. expected 2013
- Advisor: Dr. Dan Cooley
- learned to ski!



# Research opportunities

NCAR, Boulder, CO

- Graduate Student Visitor (Summer 2011)
- Climate change research
- Extreme precipitation from climate models - more later



Mesa Lab

#### SAMSI, RTP, NC

- Visitor for Fall 2011
- Participant in 2011-2012 Uncertainty Quantification program
- Lived in Chapel Hill, NC



2012 National Champions?

# Outline

This talk will mainly focus on applications and examples, with very basic introduction to theoretical results.

- Univariate extremes Red River flooding example
- Brief introduction to bivariate extremes describing dependence
- Pineapple Express project
- Suggestions for current Concordia math majors

# So what about the 2009 Red River flood?

The Fargo Red River station has daily peak flow measurements (cfs) going back to 1901.

Q: How 'rare' was the 2009 event, in terms of daily flow?

Let's use springtime (MAM) data from 1901-2008 to try to answer this.

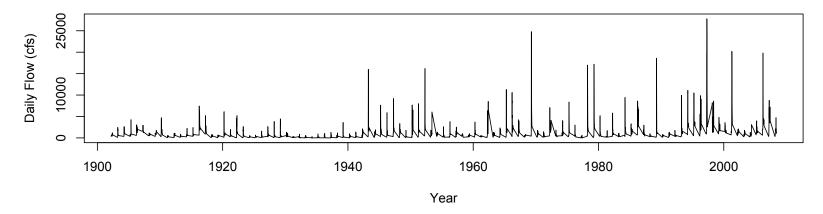
Two approaches:

- 1. Analyze all the data; fit a gamma distribution
- 2. Use only the annual max data; fit a GEV

Warning: *Both* analyses are likely incorrect, for different reasons. Bonus points if you can tell me why later.

# Modeling all springtime daily data

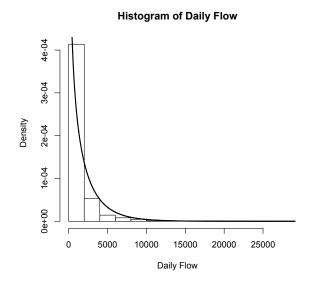




Let  $X_t$  be the daily peak flow on day t. Assume that  $X_t \sim \text{Gamma}(\alpha, \beta)$  (ignore zero-flow days).

Estimate the parameters via maximum likelihood:  $\hat{\alpha} = 0.62$ ,  $\hat{\beta} = 2765.08$ .

# Modeling all springtime daily data



Fit looks ok.

Maximum flow in 2009 was 29100 cfs on March 28<sup>th</sup>. What is the probability that the maximum daily flow in 2009 would be *at least* this much, assuming this is the right model?

 $\mathbb{P}(X_t > 29100) = 1 - F_X(29100) = 7.400995 \times 10^{-6}$ 

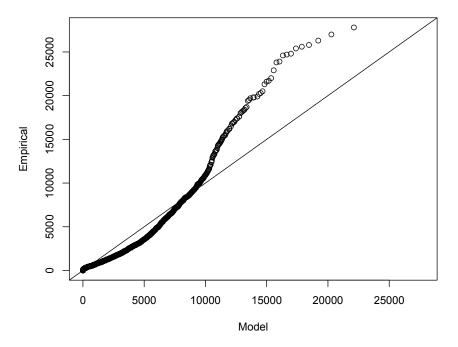
 $\mathbb{P}(\text{ann. max} > 29100) = 1 - \mathbb{P}(\text{entire year's obs} < 29100) \\= 1 - (1 - \mathbb{P}(\text{one obs} > 29100))^{92} \\= 1 - (1 - 7.400995 \times 10^{-6})^{92} \\= 6.807 \times 10^{-4}$ 

Associated "return period":  $(6.807 \times 10^{-4})^{-1} = 1469$  years. But is this the right way to analyze the data?

# All daily data model

Two main problems with this model:

- 1. Assumes daily river flow rates are *iid*
- 2. Underestimates the tail of the distribution

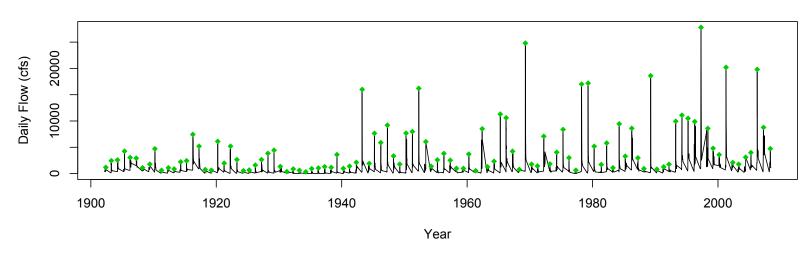


Gamma Q-Q Plot

99.4% of data and 99.8% of model's mass are < 15000.

# Modeling Annual Maxima

Alternative approach: retain only the largest value from each year, and fit a generalized extreme value distribution



**Red River at Fargo Daily Peak Flow** 

#### Why?

- 1. Intuitively, river usually only floods once each year
- 2. Mathematically, justified by theory

# Fitting a GEV

Let 
$$M_n = \max(X_t), t = 1, ..., n$$
. Assume  $M_n \sim \text{GEV}(\mu, \sigma, \xi)$ .  
 $F_{M_n}(x) = \mathbb{P}(M_n \le x) = \exp\left\{-\left[1 + \xi\left(\frac{x-\mu}{\sigma}\right)\right]^{-1/\xi}\right\}$ 

PWM estimates:  $\hat{\mu} = 2367.54$ ,  $\hat{\sigma} = 2496.05$ ,  $\hat{\xi} = 0.32$ .  $\mathbb{P}(\text{ann. max} > 29100) = 1 - F_{M_n}(29100) = 0.0095$ Associated return period:  $(0.0095)^{-1} = 105$  years

### 100-year return level: 28561

95% confidence interval (delta method): (8909, 48213)

#### 500-year return level: 51536

95% confidence interval (delta method): (1718, 101354)

# A word of caution

Because we use one observation per year, uncertainty associated with return period/level estimates are *very high*.

Q: Why is uncertainty so important?

Because we use one observation per year, uncertainty associated with the previous return period estimate is *very high*.

### Q: Why is uncertainty so important?

"There are known knowns; there are things we know we know. We also know there are known unknowns; that is to say we know there are some things we do not know. But there are also unknown unknowns - there are things we do not know we don't know."

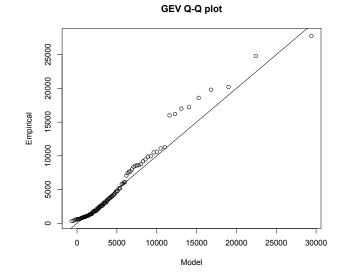
Former Secretary of Defense Donald Rumsfeld

Think of uncertainty as a 'known unknown'. Important to acknowledge that this exists and quantify it.

This is the crux of Mathematical Statistics.

# GEV model fit

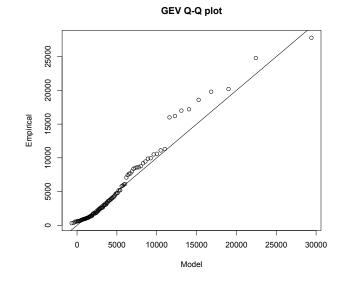
#### Plot shows annual maxima 1902-2008.



This model is a better fit. But are we missing something?

# GEV model fit

#### Plot shows annual maxima 1902-2008.



This model is a better fit. But are we missing something? Non-stationarity? Six of the seven largest flooding events occurred in the last 15 years.

Could be handled by a regression-type approach.

### Why use only 'extreme' observations?

Two approaches for extracting extreme observations:

- 1. Block-maximum approach (done above)
- 2. Threshold-exceedance approach (skipped today)

*Heuristic explanation:* Phenomena which generate extreme observations are different than those which generate typical observations (Red River floods?).

Mathematical explanation: Assume  $X_t$  has cdf  $F_X(x)$ .

$$F_{M_n}(x) = P(M_n \le x) = P(X_t \le x \text{ for all } t = 1, ..., n)$$
  
=  $P^n(X_t \le x)$   
=  $F_X^n(x)$ 

If we know  $F_X$  exactly, then we know  $F_{M_n}$  exactly. But if we have to estimate  $F_X$ , any errors in estimating the tail get amplified by a power of n.

# Generalized Extreme Value distribution

The GEV distribution captures three types of tail behavior.

$$F(x) = \exp\left\{-\left[1 + \xi\left(\frac{x-\mu}{\sigma}\right)\right]^{-1/\xi}\right\}$$

The parameter  $\xi$  determines tail behavior and is difficult to estimate in practice.

- $\xi < 0$ : Weibull case (bounded tail)
- $\xi = 0$ : Gumbel case (light tail), interpreted as limit
- $\xi > 0$ : Fréchet case (heavy tail)

Q: Why is the GEV the right distribution to fit to annual maximum data?

Recall from your introductory statistics course: the Central Limit Theorem says that the Normal distribution is the right distribution for sums of *iid* random variables...

...there is a 'CLT-like' result for block maxima.

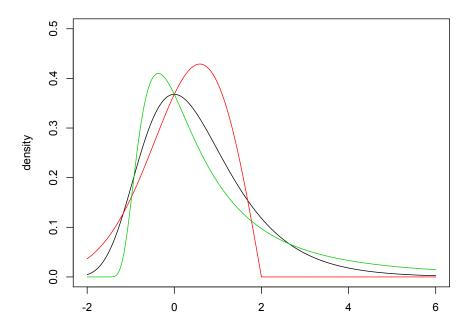
Let  $M_n = \max_{t=1,\dots,n} X_t$ , where  $X_t$  are iid. If there exist normalizing sequences  $a_n$  and  $b_n$  such that  $P\left(\frac{M_n-b_n}{a_n} \leq x\right) \to G(x)$ (nondegenerate) as  $n \to \infty$ , then

$$G(x) = \exp\left\{-\left[1 + \xi x\right]^{-1/\xi}\right\}.$$

We don't need information about the distribution of  $X_t$  to know about the distribution of  $M_n$ .

# Limiting Distributions

**GEV** distributions



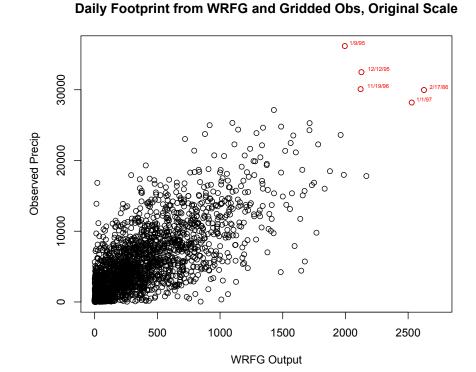
Weibull ( $\xi = -0.5$ ) Gumbel Fréchet ( $\xi = 0.5$ )

# **Bivariate Extremes**

Assume  $\mathbf{X}_t = (X_{t,1}, X_{t,2})$  iid.

Idea: Describe the joint tail of the distribution

Goal: Often extrapolation



# Bivariate Extremes: Example

Portfolio consisting of two securities. What is the probability that one goes bust, given that the other does?

The Gaussian copula was a popular model for portfolio risks *at least until 2008:* 

# Bivariate Extremes: Example

Portfolio consisting of two securities. What is the probability that one goes bust, given that the other does?

The Gaussian copula was a popular model for portfolio risks *at least until 2008:* 

$$Pr[T_{A} < 1, T_{B} < 1] = \varphi_{2}(\varphi^{-1}(F_{A}(1)), \varphi^{-1}(F_{B}(1)), \gamma)$$

Wired magazine: "Recipe for Disaster: The Formula that Killed Wall Street". February 23, 2009.

Basic problem: this model doesn't account for multiple securities experiencing an 'extreme event' at the same time.

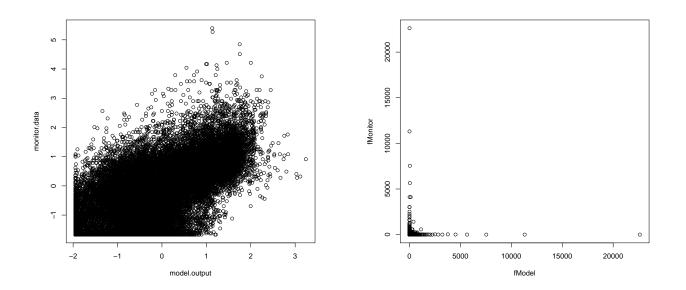
Some forward-thinking statisticians in extreme value theory were warning of this problem years ahead of time.

Their warnings went largely unheeded.

Marginal and dependence effects are typically handled separately.

Approach:

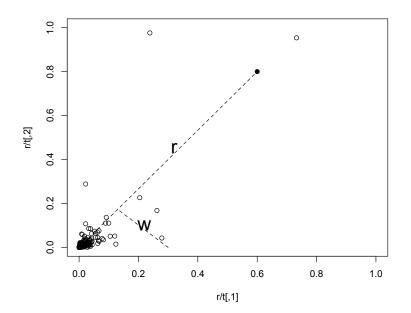
- 1. Transform marginal to something nice.
- 2. Describe dependence.



Note: high correlation does not imply tail dependence!

# Describing tail dependence

Limit result: extremes occur according to a Poisson process.



Intensity measure factors into 'radial' component (r) and 'angular' component (w).

Dependence is described by a probability measure on w.

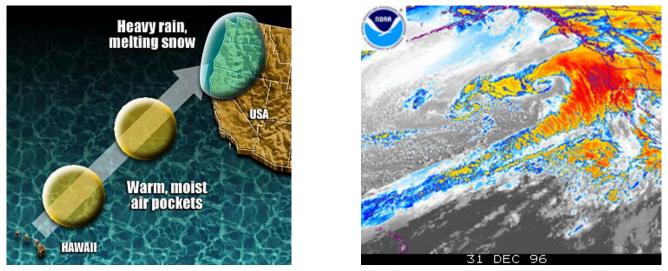
# Pineapple Express project

#### What this section is *not* about



PE storms: caused by atmospheric rivers hitting the west coast in winter

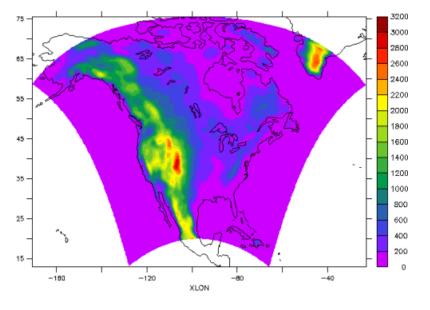
- Often bring heavy rain and warm temperatures
- Great impact on water resources of western US



Question of Interest: How well are Regional Climate Models able to represent extreme precipitation caused by this phenomenon?

# **Regional Climate Models**

Use input from a low-resolution global model and known physics of the Earth system to produce simulated weather over long periods of time at finer scales ( $\sim$ 50km).



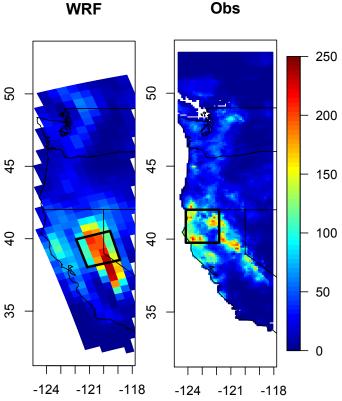
NARCCAP domain

Aim: study regional impacts of climate change scenarios For evaluation, RCMs forced by reanalysis for 1979-2004.

# RCM output vs. observations

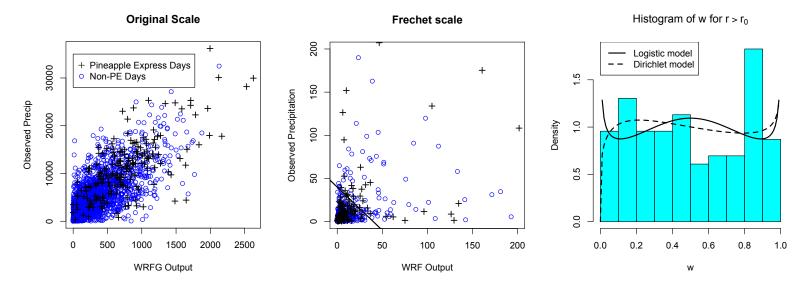
Idea: compare daily precipitation amounts from RCM to observed data

- Identify region and quantity a that capture PE events
- NDJF days 1981-1999
- Right: January 1, 1997 (big PE event)
- Method: bivariate extreme value analysis



# Bivariate extremes analysis

Transform each marginal to unit Fréchet and examine dependence

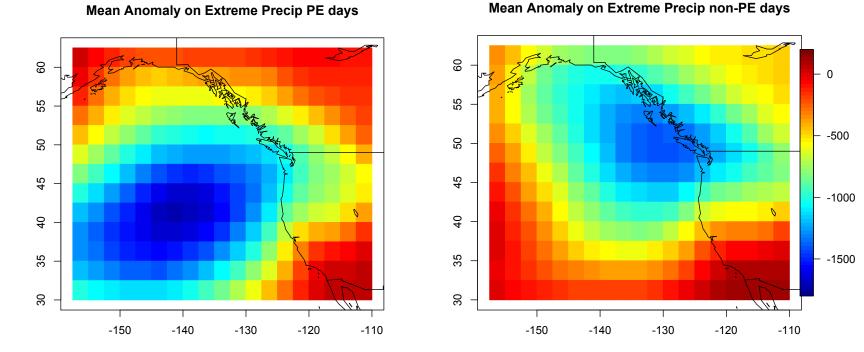


Find strong tail dependence - the WRF model represents the largest events quite well

Not all 'extreme' events are PE - aim to link to processes

# Pineapple Express precipitation index

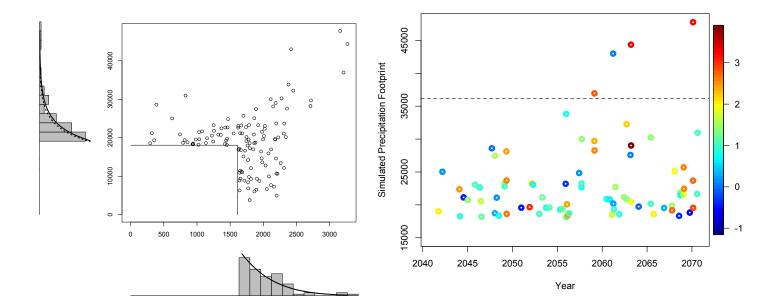
### Developed a PE index from daily sea-level pressure fields



A first step toward linking regional extreme precipitation to large-scale processes

# Future PE events from Regional Climate Models

Use fitted dependence model to study future extreme events



Findings: increase in both frequency and intensity of PE storms as produced by the WRF regional model forced by CCSM global model

# Summary

- 'Usual' distributions don't always capture tails correctly.
- Aim of EVT: study the tail of a distribution.
- An extreme value analysis utilizes only data considered 'extreme'.
- Goal of the analysis is often extrapolation.
- Foundation provided by results from probability theory.
- Communication of uncertainty is critical.
- Dependence *not* described with correlations/covariances.
- Application areas: climate, finance, engineering, ...

- More and better job opportunities
  - Often more flexible and interesting
  - Not just a 'number cruncher'
- Almost every day presents a new challenge
- Flexible (although busy) schedule
- You still get to be a student!



2009: Colorado State 23, Colorado 17 (in Boulder)

# For Today's Graduate, Just One Word: Statistics

By STEVE LOHR Published: August 5, 2009

- NY Times article: "the sexy job in the next ten years"
- Data is abundant, but people who can analyze it properly are not
- Computing power allows for new solutions to difficult problems
  - Need to quantify uncertainty in computer simulations
  - SAMSI 2011-2012 UQ program
- Wide variety of applications can 'play in everyone else's backyard'

# Preparing for grad school while at Concordia

As a Concordia mathematics major...

- Talk to your professors they're always willing to help!
- Explore research opportunities
  - Research experiences for undergraduates (REU)
  - Undergraduate research with Concordia professors
  - SAMSI Undergraduate Workshop: February 24-25
- Take real analysis & other proof-based courses
- Get experience teaching/tutoring

When looking at schools...

- Ask a lot of questions!
- Money is important too funding, fees, insurance, etc.
- If possible, make a visit

# Reference and contact

Weller G., Cooley D., Sain S. An investigation of the pineapple express phenomenon via bivariate extreme value theory. *Submitted*.

Website: www.stat.colostate.edu/~weller Email: gbweller@cord.edu