

Multivariate Frequency Analysis of Extreme Rainfall Events using Copulas

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Extreme Events



Feb. 5, 2008
Delphi, Indiana
Flooding of Tippecanoe River



(AP Photo/Journal & Courier, Michael Heinz)

Sept., 2007
George H. Sparks Reservoir
Lithia Springs, Georgia



(Barry Gillis, <http://www.drought.unl.edu/gallery/2007/Georgia/Sparks1.htm>)



Outline



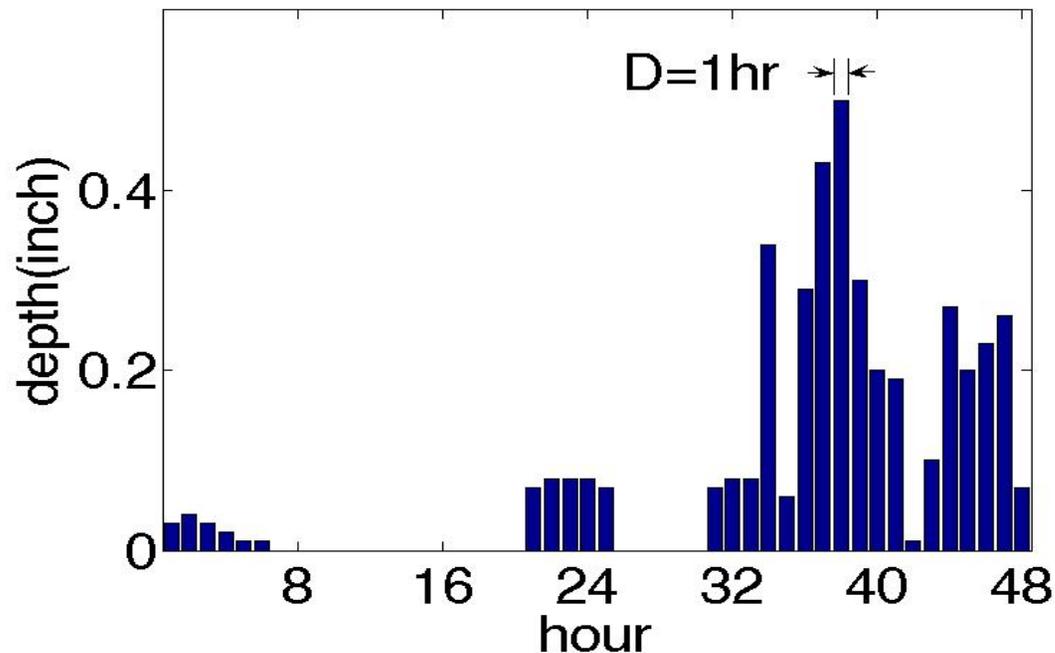
- Background and motivation
 - Limitations in univariate approach
- Introduction to copulas
- Research objectives
 - Topic 1: Probabilistic structure of surface runoff
 - Topic 2: Extreme rainfall frequency analysis
 - Topic 3: Drought frequency analysis
- Summary and concluding remarks



Limitations in Univariate Approach



- Eg: Selection of annual maximum precipitation events
 - *Durations* are not the actual durations of rainfall events
 - Long-term maximum may cover multiple events
 - Short-term maximum encompasses only part of the extreme event



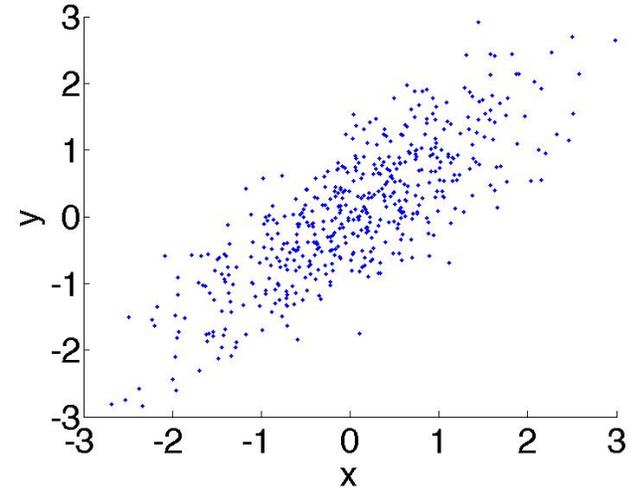
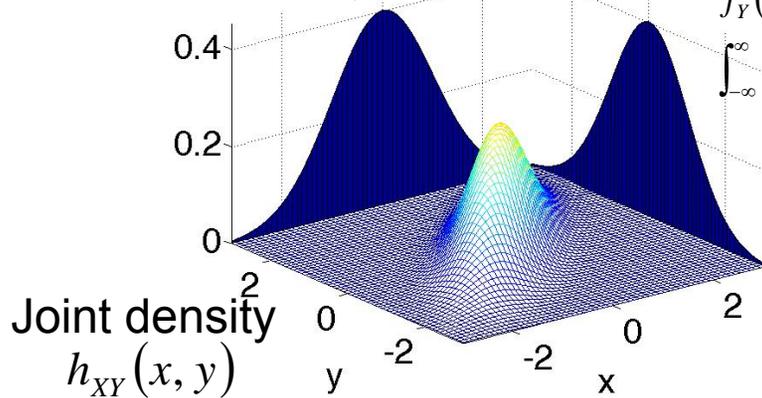
Bivariate Distribution Example



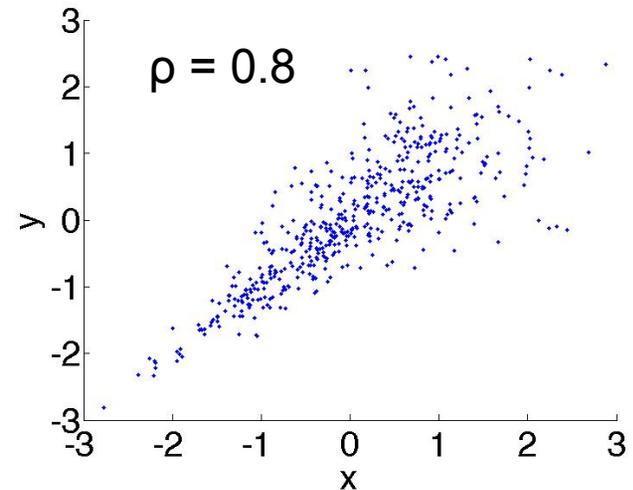
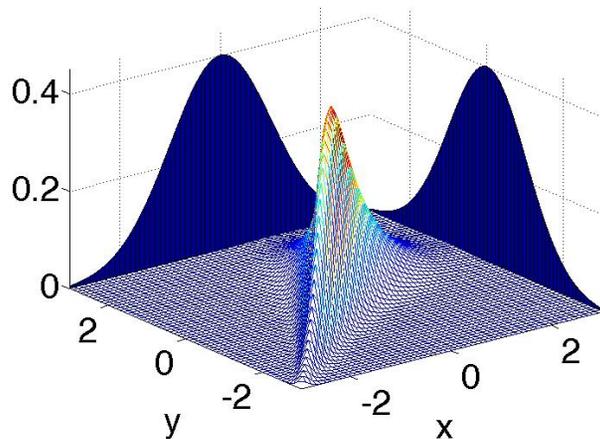
Bivariate Gaussian distribution, $\rho = 0.8$

Marginals $f_X(x) = \int_{-\infty}^{\infty} h_{XY}(x, y) dy$

$$f_Y(y) = \int_{-\infty}^{\infty} h_{XY}(x, y) dx$$



Gaussian marginals with Clayton Copulas



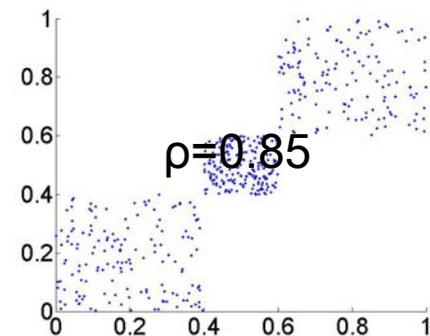
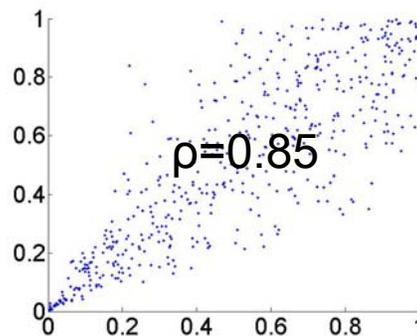
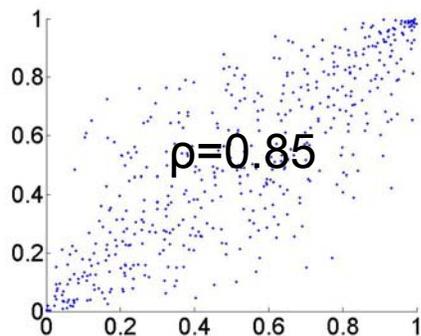
Concept of Dependence Structure



- Conventionally quantified by the linear correlation coefficient ρ

$$\rho_{XY} = \frac{E[(X - \bar{x})(Y - \bar{y})]}{Std[X]Std[Y]}$$

- Only valid for Gaussian (or some elliptic) distributions



- Can not correctly describe association between variables
- A better tool is required to characterize dependence

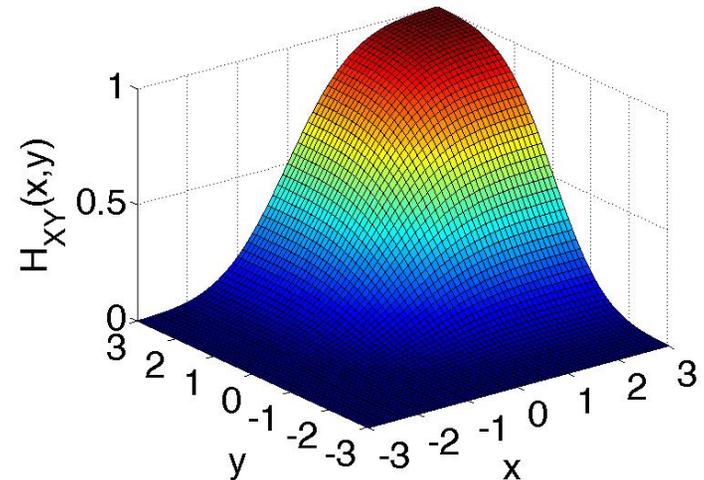


Copulas

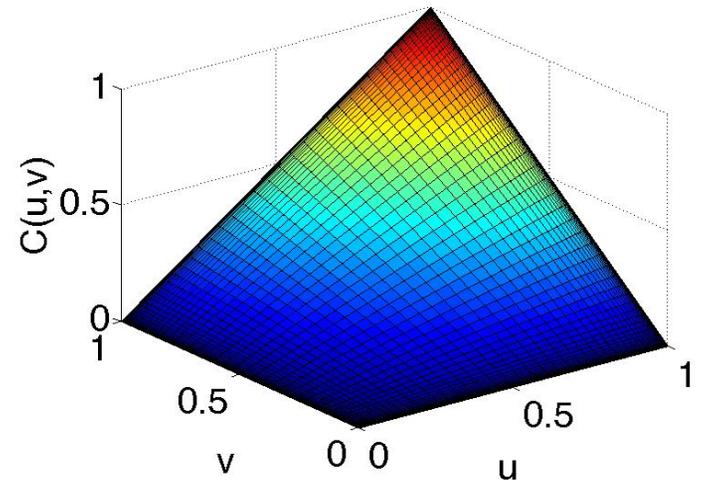


- Transformation of joint cumulative distribution
 - $H_{XY}(x,y) = C_{UV}(u,v)$
marginals: $u = F_X(x)$, $v = F_Y(y)$
 - Sklar (1959) proved that the transformation is *unique* for continuous r.v.s
- Use copulas to construct joint distributions
 - Marginal distributions => selecting suitable PDFs
 - Dependence structure => selecting suitable copulas
 - Together they form the joint distribution

Bivariate Gaussian distribution, $\rho = 0.1$



Gaussian Copulas, $\rho = 0.1$



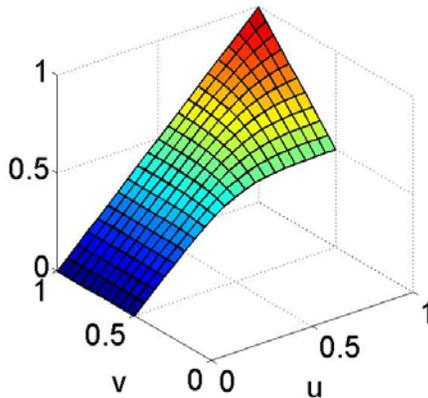
Example of Copulas – Frank Family



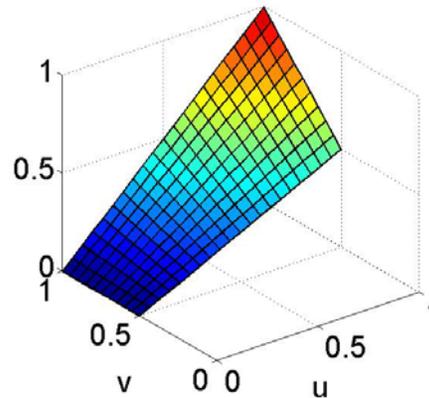
- Frank family of Archimedean copulas

$$C_{Frank}(u, v) = -\frac{1}{\theta} \ln \left(1 + \frac{(e^{-\theta u} - 1)(e^{-\theta v} - 1)}{e^{-\theta} - 1} \right)$$

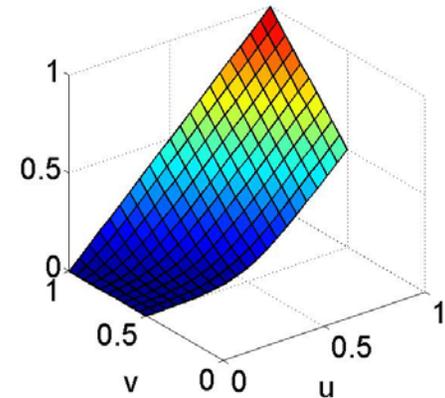
Frank family, $\theta=10$



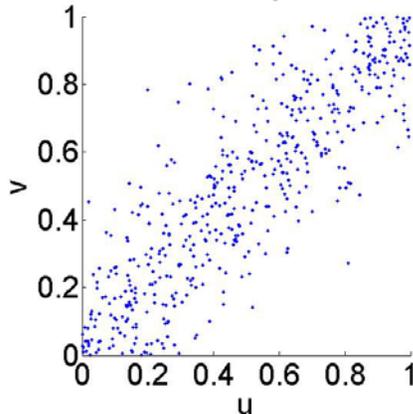
Frank family, $\theta=0.01$



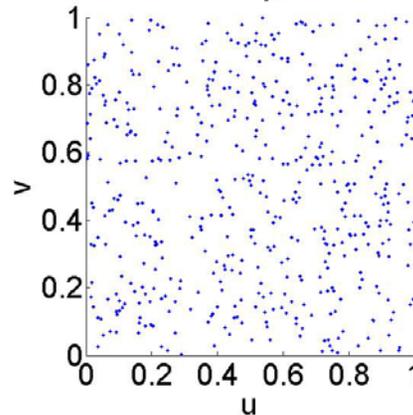
Frank family, $\theta=-10$



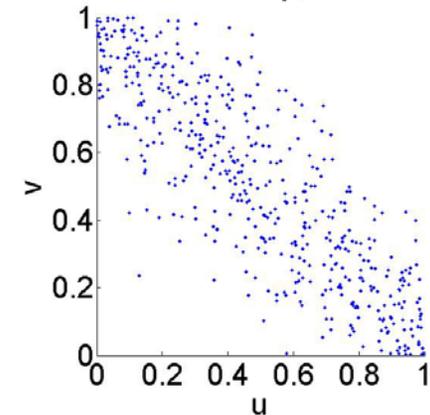
Frank family, $\theta=10$



Frank family, $\theta=0.01$



Frank family, $\theta=-10$



Use of Copulas in Hydrology



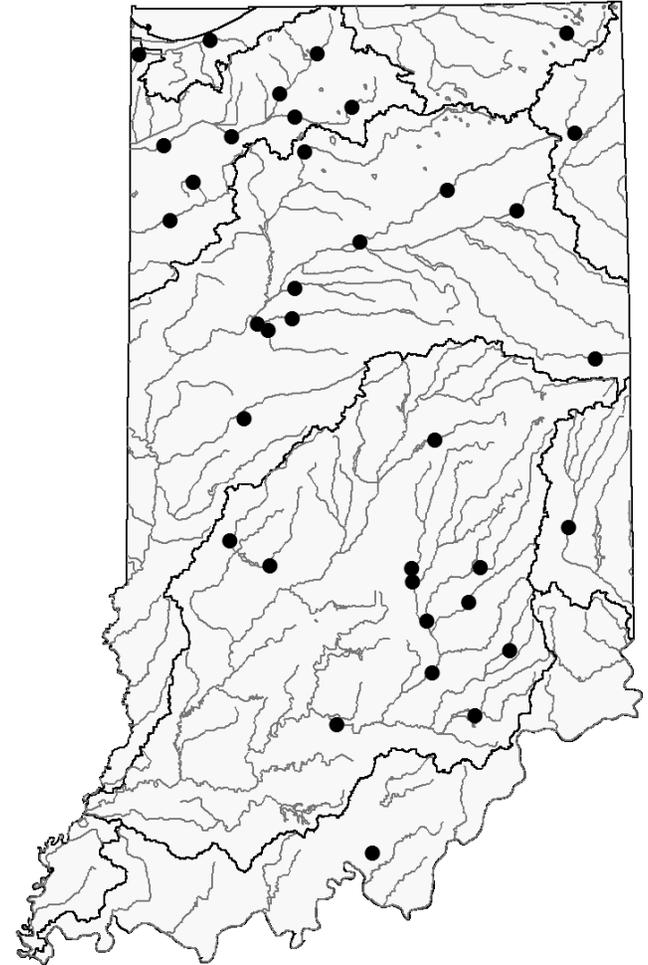
- Since 2003, over 20 research articles have been published in water resources related journals
 - Rainfall and flood frequency analysis, groundwater parameters estimation, sea storms analysis, rainfall IDF curves, and *etc.*
 - Full potential of copulas is yet to be realized (Genest and Favre, 2007)
- For copulas in rainfall frequency analysis:
 - How dependence affects rainfall depth and surface runoff?
 - The definition of extreme events was not clear
 - Few stations were examined
- For copulas in drought frequency analysis:
 - What's the dependence between various drought indices?



Data Sources & Study Area



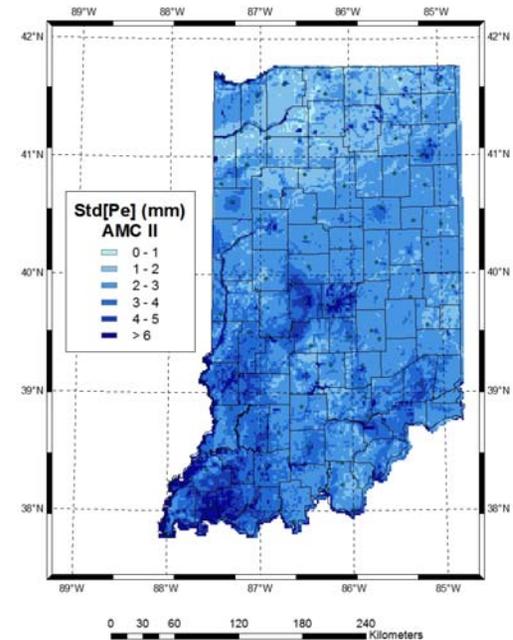
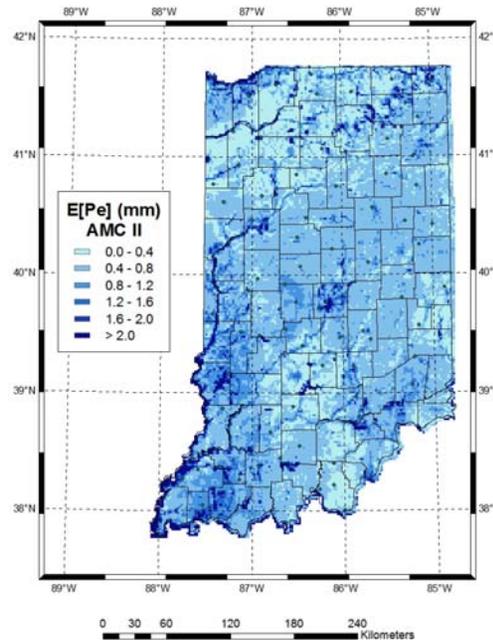
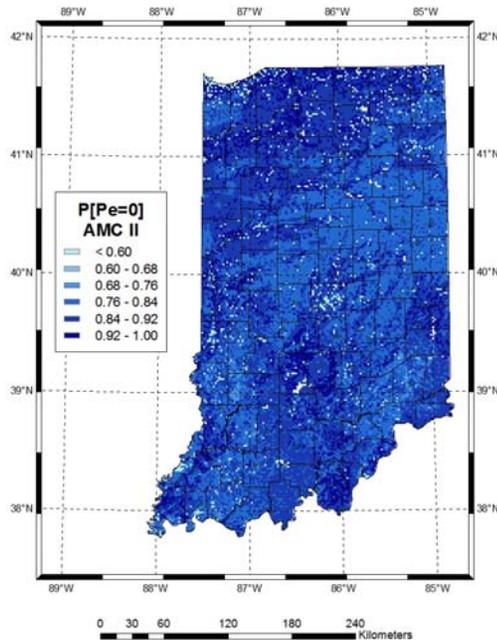
- Precipitation
 - NCDC hourly precipitation dataset
 - 53 stations with record length greater than 50 years
 - NCDC daily precipitation dataset
 - 73 stations with record length greater than 80 years
- Streamflow
 - USGS unregulated daily mean flow
 - 36 stations with record length greater than 50 years



- SCS CN method for excess rainfall depth P_e

$$P_e = \frac{(I \cdot D - 0.2S)^2}{(I \cdot D + 0.8S)}$$

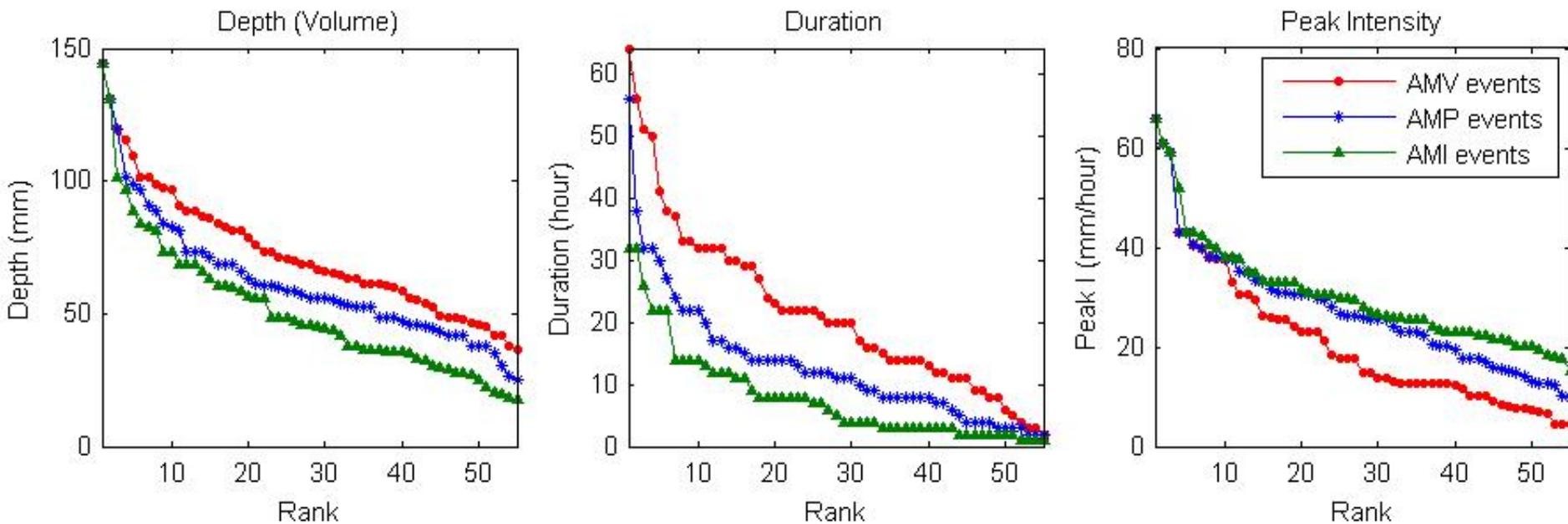
- A function of both average intensity I and rainfall duration D
- How does the dependence between I & D affect P_e ?
- GIS maps of $P[P_e=0]$, $E[P_e]$, and $\text{Std}[P_e]$ for Indiana



Extreme Rainfall Frequency Analysis (I)



- Definitions of Extreme Rainfall Events
 - Hydrologic designs are usually governed by depth (volume) or peak intensity
 - Annual maximum volume (AMV) events
 - Longer duration
 - Annual maximum peak intensity (AMI) events
 - Shorter duration
 - Annual maximum cumulative probability (AMP) events
 - The use of empirical copulas between volume and peak intensity
 - Wide range of durations





- Bivariate distribution H_{PD} , H_{DI} , H_{PI}
 - Marginal: Extreme Value Type I (EV1), Log Normal (LN)
 - Dependence: Frank Family

- Applications

- Estimate of depth for known duration

$$F_p(p_T | d - 1 < D \leq d) = 1 - 1/T$$

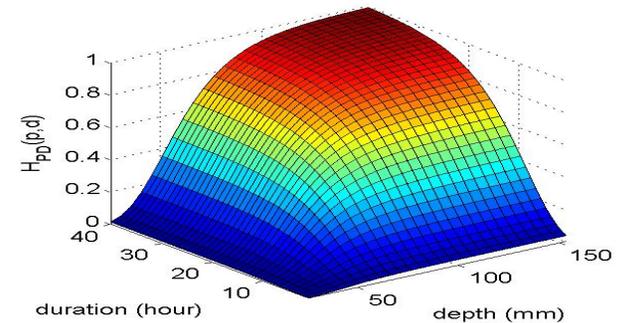
- Estimate of peak intensity for known duration

$$F_I(i_T | d - 1 < D \leq d) = 1 - 1/T$$

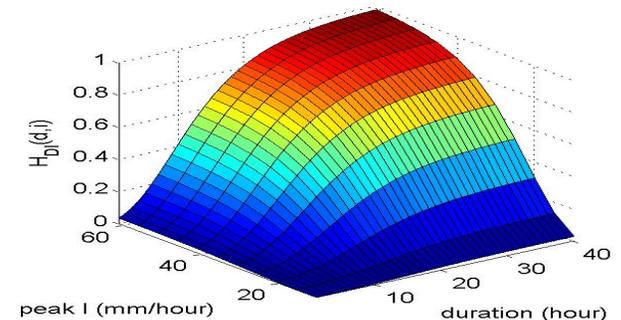
- Estimate of peak intensity for known depth

$$E[I | P > p]$$

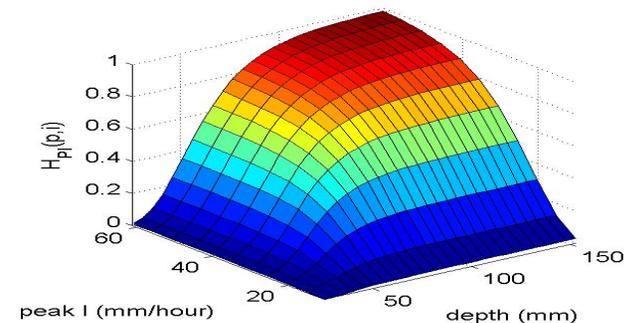
$H_{PD}(p,d)$, AMP events, Station: 120132



$H_{DI}(d,i)$, AMP events, Station: 120132



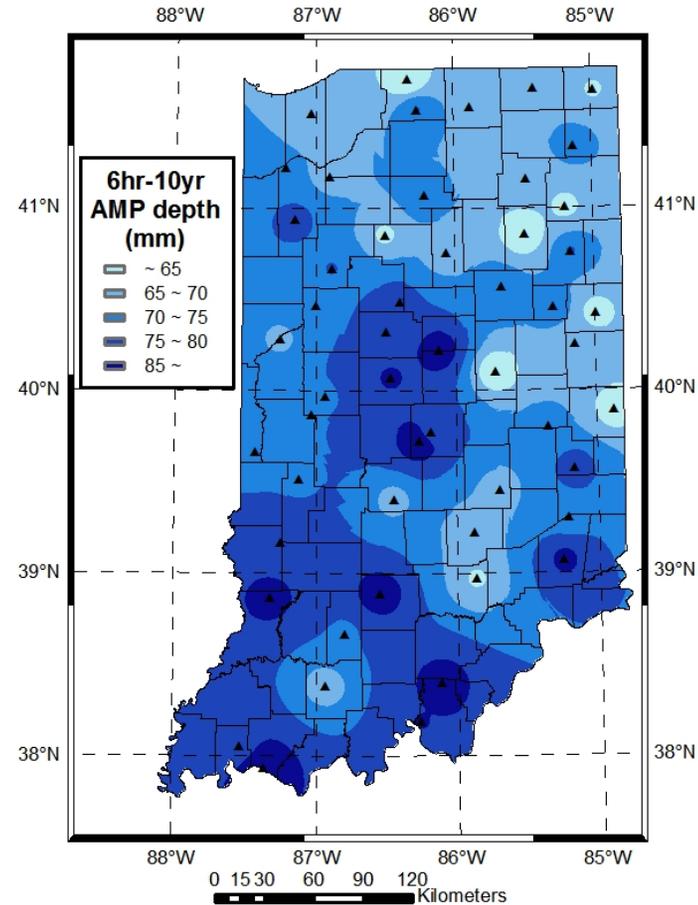
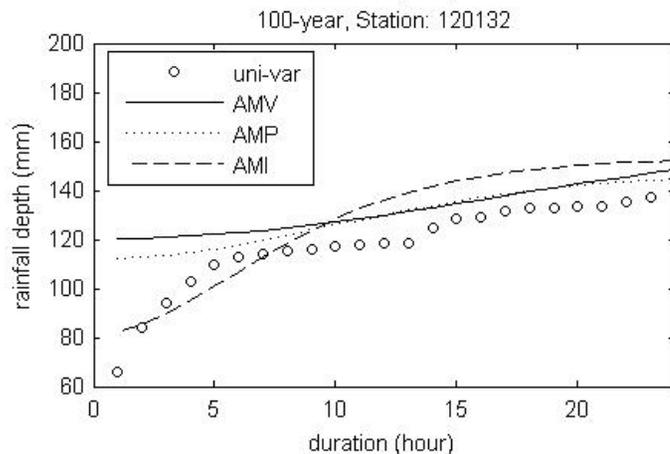
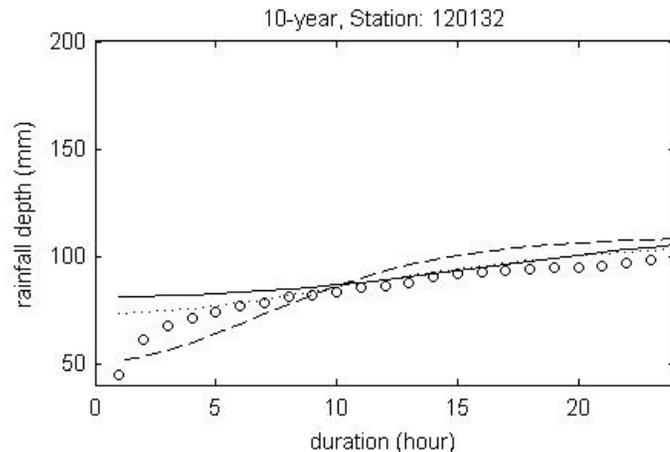
$H_{PI}(p,i)$, AMP events, Station: 120132



Estimate of depth for known duration



T-year depth p_T given duration d

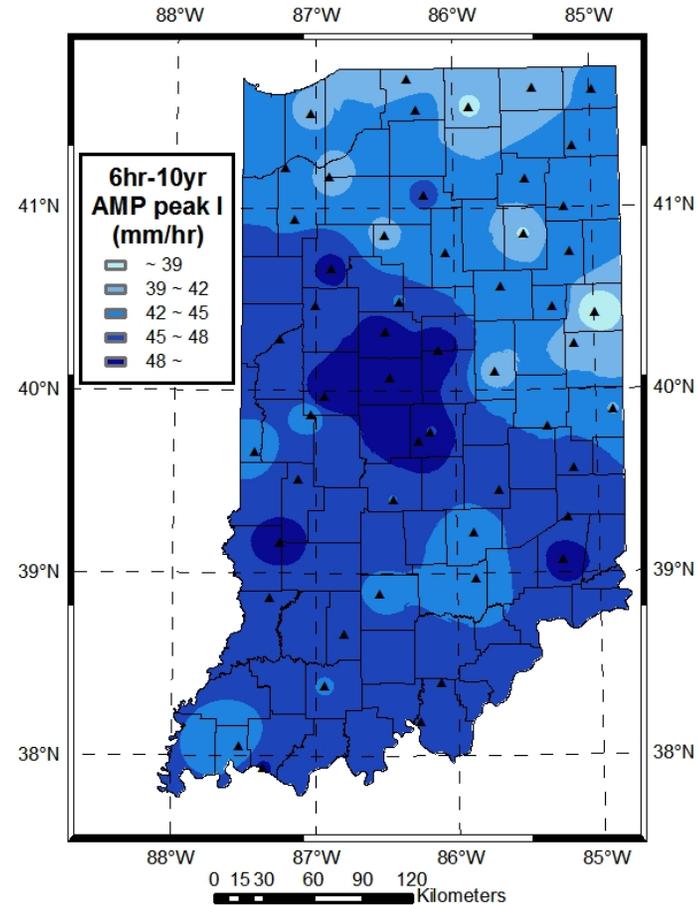
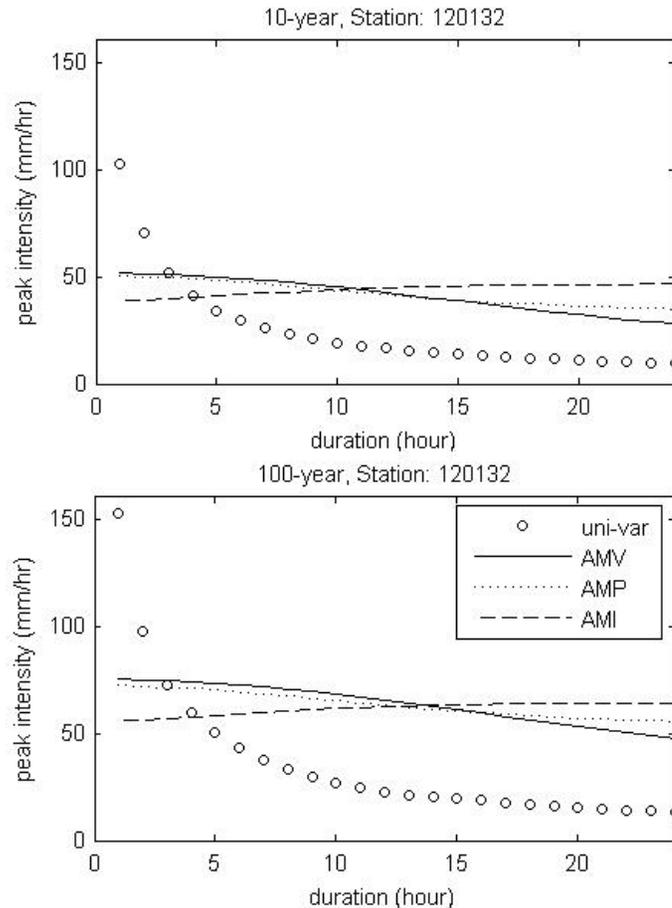


- The estimates of depth are similar for durations larger than 10 hours
- AMP definition seems to be an appropriate indicator for defining extreme events

Estimate of peak intensity for known duration



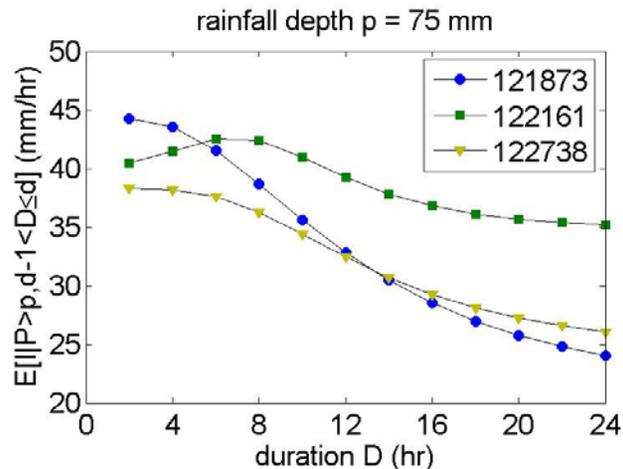
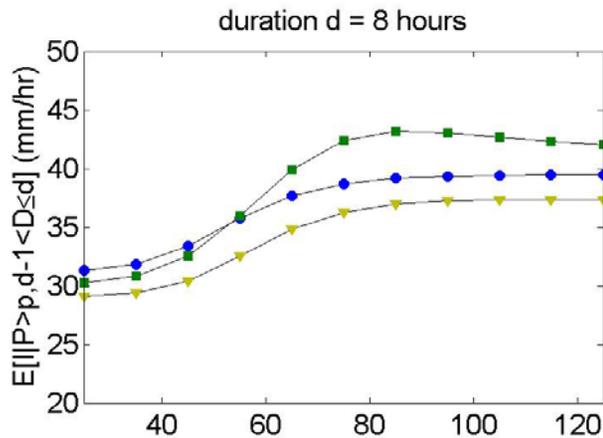
T-year peak intensity i_T given duration d



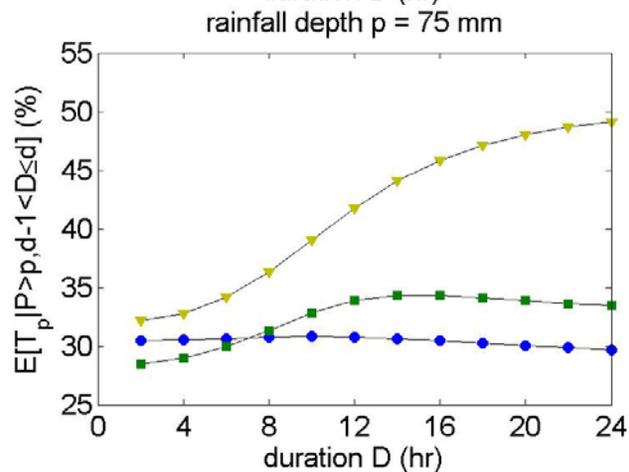
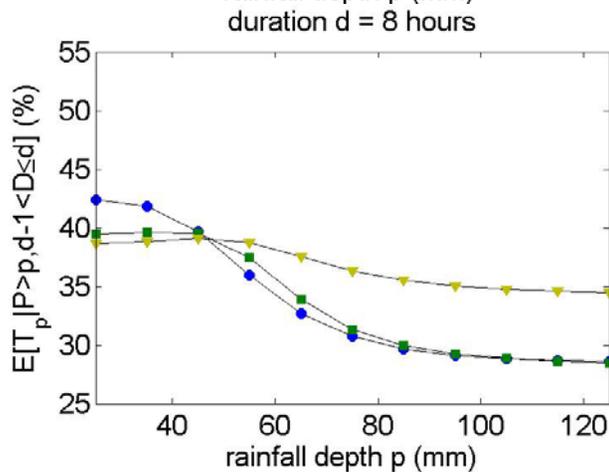
- Conventional approach fails to capture the peak intensity

Rainfall Peak Attributes

- Given depth (P) and duration (D), compute the conditional expectation of peak intensity (I) and percentage time to peak (T_p)



Expectation of peak intensity given P & D



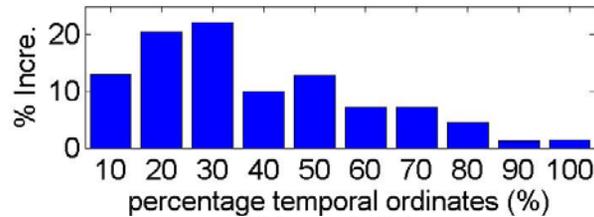
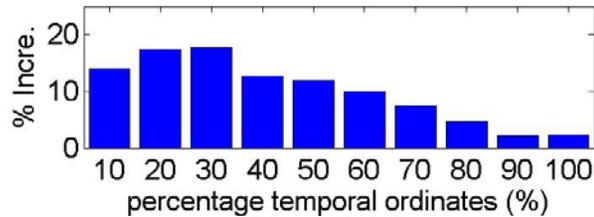
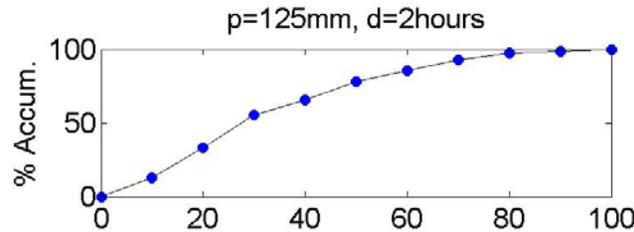
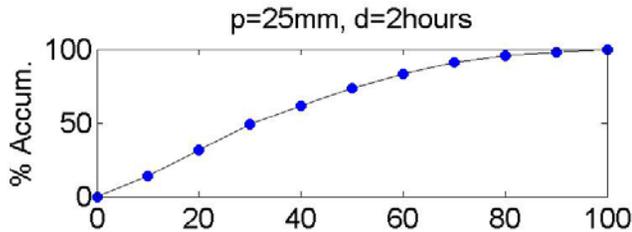
Expectation of time to peak (%) given P & D



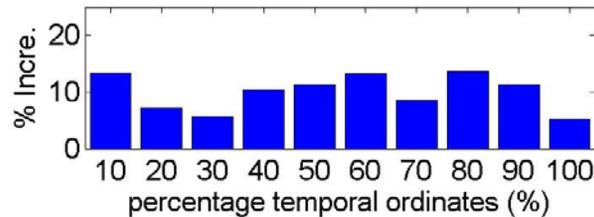
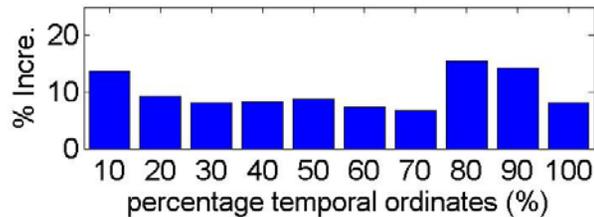
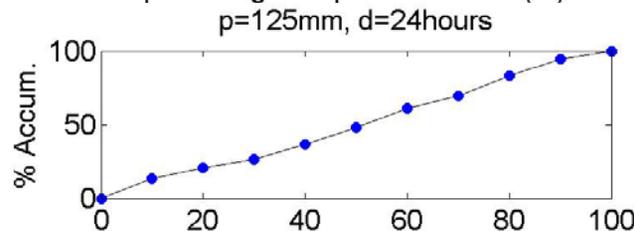
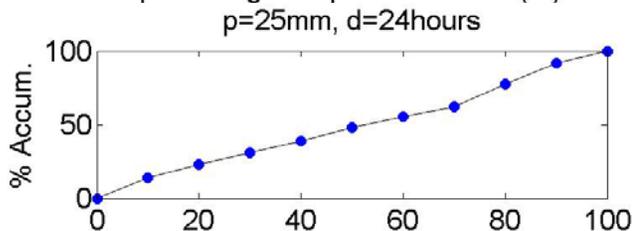
Temporal Accumulation Curves



- Given depth (P) and duration (D), compute the conditional expectation of percentage accumulations at each 10% temporal ordinates ($A_{10}, A_{20}, \dots, A_{90}$)



Expectation of % accumulation given P & D



Drought Frequency Analysis



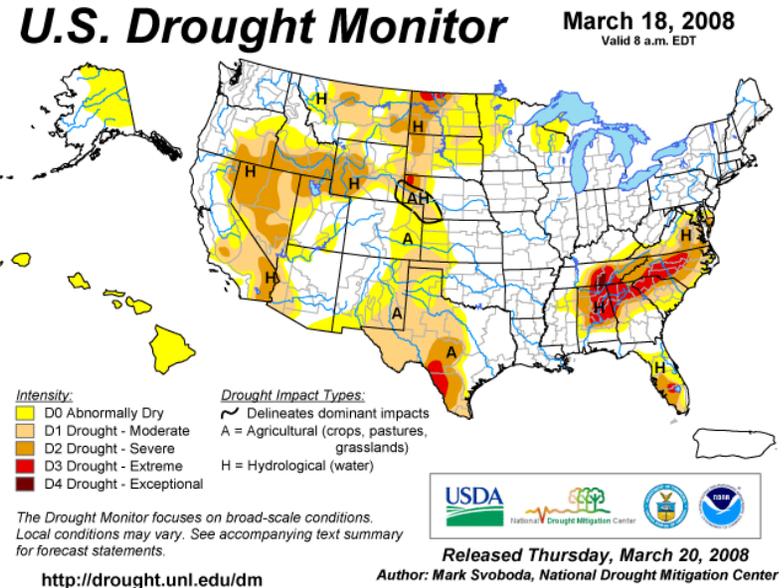
- Challenges in characterizing droughts
 - No clear (scientific) definition: deficit of water for prolonged time
 - Phenomenon dependent in time, space, and between various variables such as precipitation, streamflow, and soil moisture
- Classification of droughts
 - Meteorological drought: precipitation deficit
 - Hydrologic drought: streamflow deficit
 - Agricultural drought: soil moisture deficit
- Various drought indices
 - Palmer Drought Severity Index (PDSI), Crop Moisture Index (CMI), Surface Water Supply Index (SWSI), Vegetation Condition Index (VCI), CPC Soil Moisture, Standardized precipitation index (SPI)



US Drought Monitor



- Overall drought status (D0 ~ D4) determined based on various indices together (Svoboda *et al.*, 2002)
 - PDSI
 - CPC Soil moisture
 - USGS weekly
 - Percentage of normal
 - SPI
 - VCI

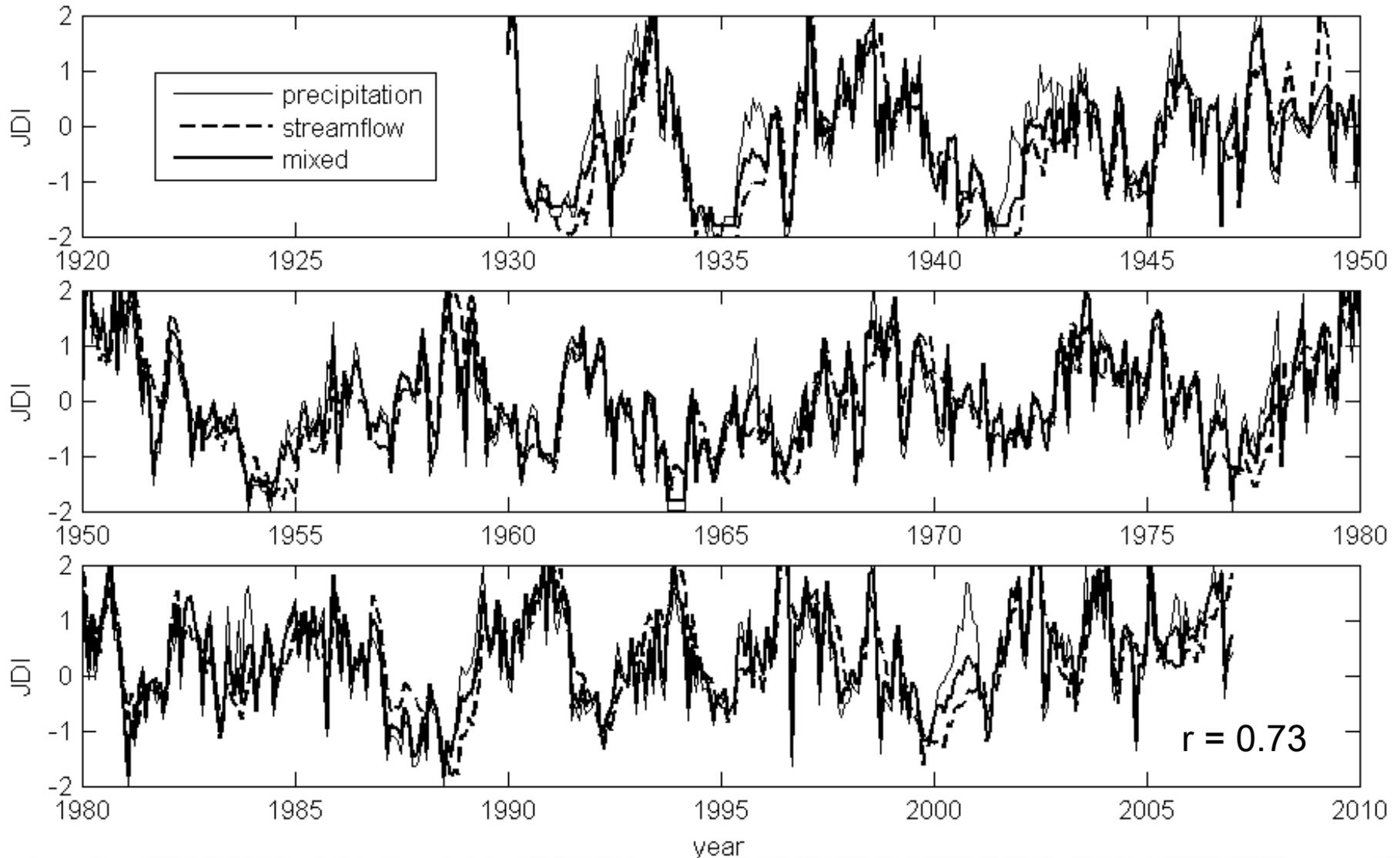


<http://drought.unl.edu/dm/monitor.html>

- Linear combination of selected indices (OBDI, objective blend of drought indicator) was adopted as the preliminary overall drought status
- The decision of final drought status relies on subjective judgment
- Copulas can be adopted to compute the joint distribution of drought indices, and then used to compute the joint deficit index (JDI)



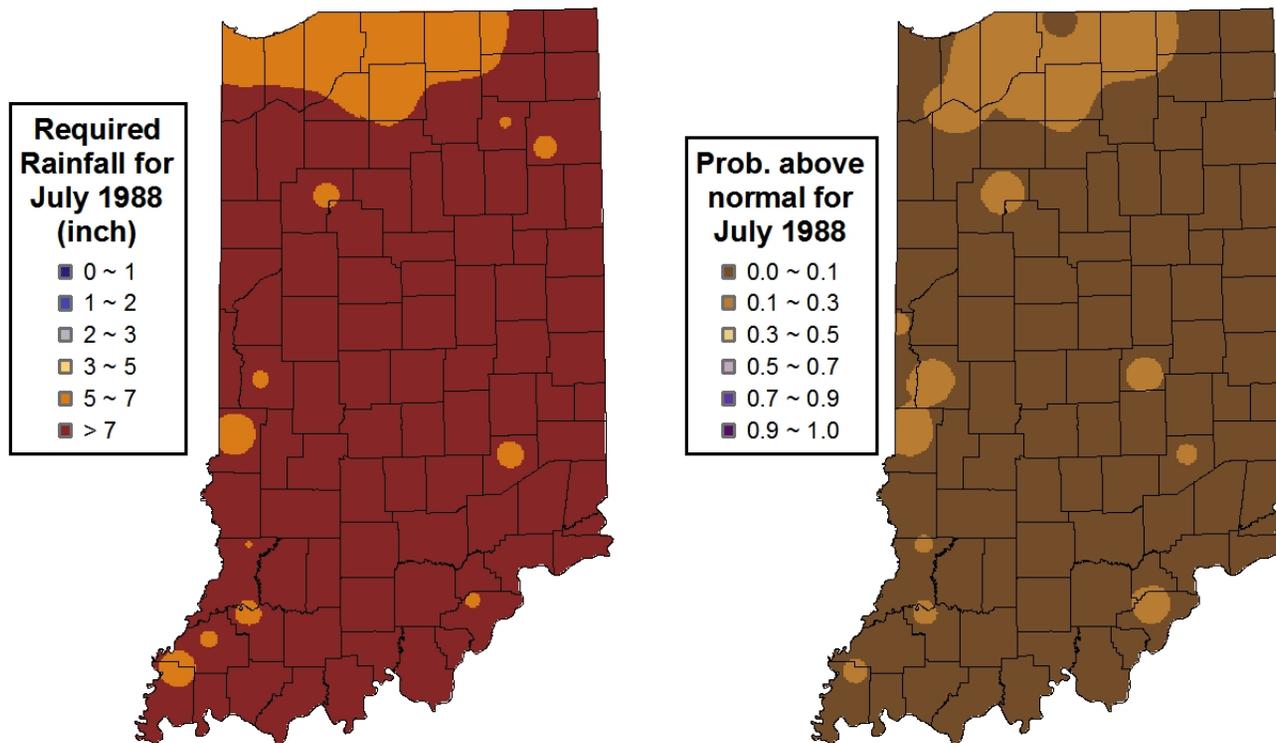
Precipitation and Streamflow JDIs



Potential of Future Droughts



- Required precipitation for reaching joint normal status (cumulative probability to be 0.5) in the future
- Probability of drought recovery



Summary and Concluding Remarks



- Copulas are found to be flexible for constructing joint distributions (no specific marginals are required).
- The dependence structure can be faithfully preserved
- Copula-based multivariate algorithm is potential for general statistic-related engineering problems
- Caution when using copulas
 - Need sufficient historic records
 - Intrinsic assumption of stationary
 - Difficulties arise in higher dimensions
 - Mathematical complexity
 - Hard to preserve all lower level mutual dependencies
 - Compatibility problem



Acknowledgements



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- Thank my family and friends



Thank you
Questions?

