

Multivariate Flood Frequency Analysis through Copulas in a Partially Gauged Watershed

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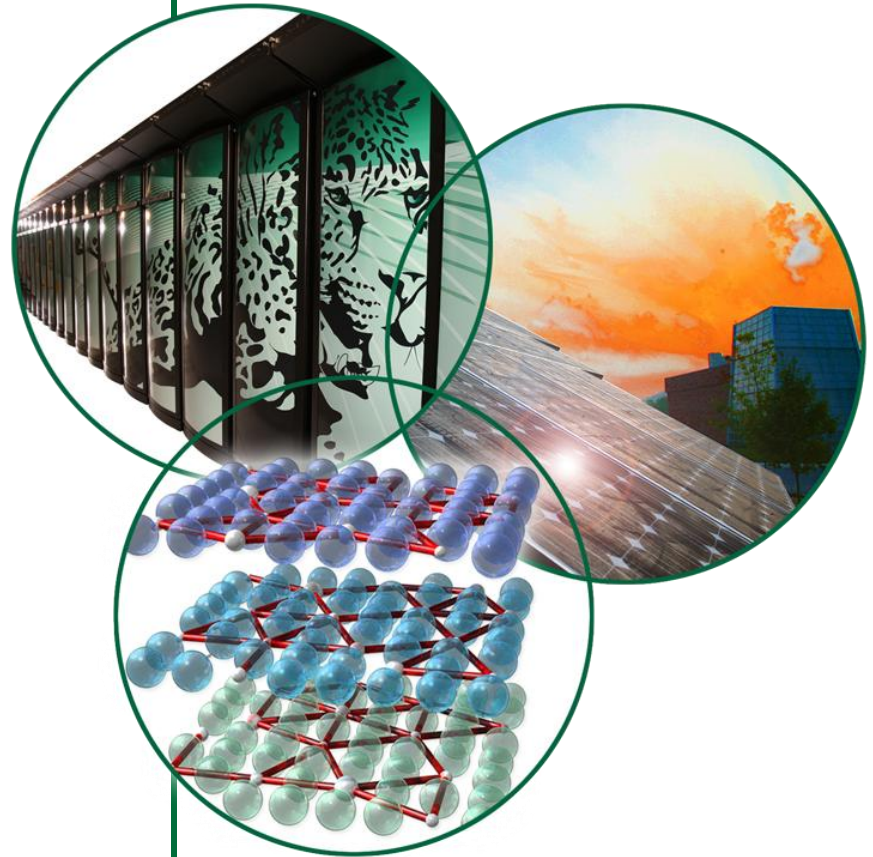
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Background

- **25000 USGS Gauge Stations**
 - Enough?
- **Estimate Flood Frequency at Ungauged Locations**
 - Modeling Approach
 - Statistical Approach
- **Limitations of the Univariate Flood Frequency Analysis**
 - How to account for river confluences?
 - What if a river has been partially regulated?
 - How to account for major land use and land cover change?
- **Multivariate Flood Frequency Analysis Could be a Solution**
 - *But, can we make it easier?*

Joint Distribution and Copulas

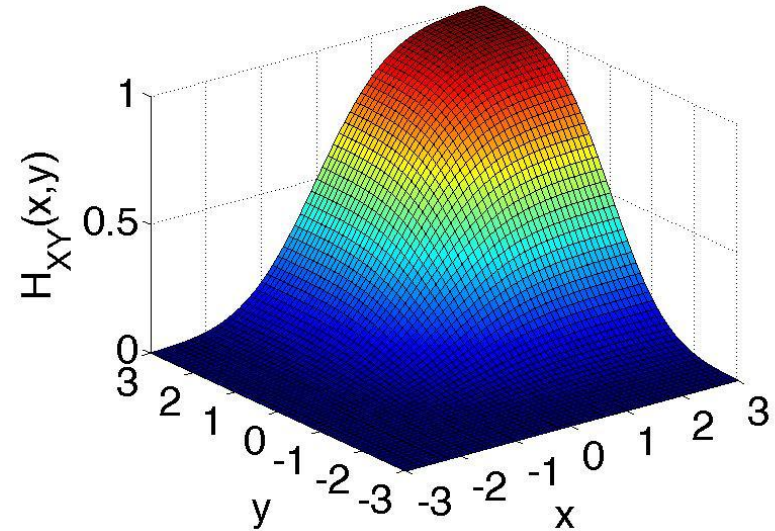
- One may formulate any joint distribution in terms of copulas and marginals

- $H_{XYZ}(x,y,z) = C_{UVW}(u,v,w)$
 $u = F_X(x), v = F_Y(y), w = F_Z(z)$
- Copulas is a “distribution-free” dependence structure

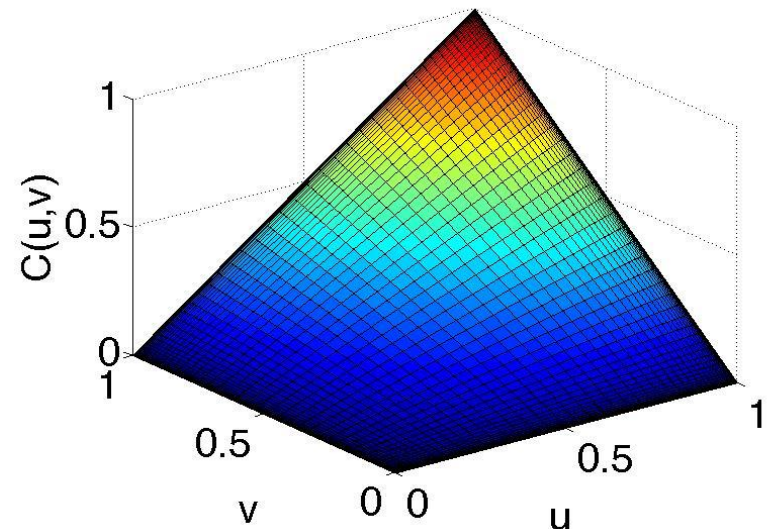
- Use copulas to construct joint distributions

- Marginal distributions => selecting suitable PDFs
- Dependence structure => selecting suitable copulas
- **Together they form joint distribution with no specific marginals**

Bivariate Gaussian distribution, $\rho = 0.1$



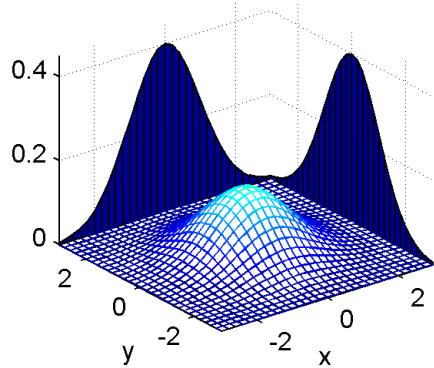
Gaussian Copulas, $\rho = 0.1$



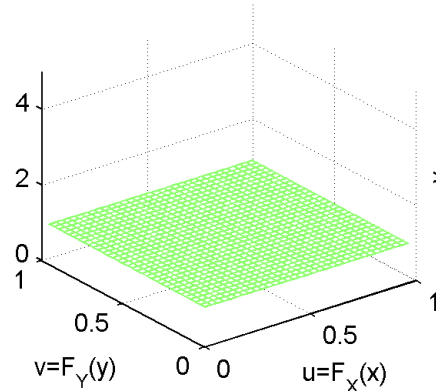
Copula Density

- Joint PDF versus copula density
 - Positive dependence: main diagonal ($u = v$)
 - Independence: flat surface
 - Negative dependence: secondary diagonal ($u - v = 1$)

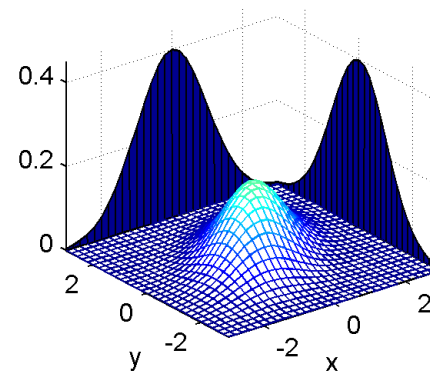
(a) Bivariate Gaussian PDF, $\rho = 0$



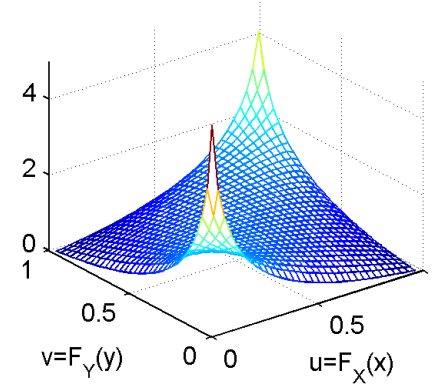
(b) copula density, $\partial^2 C / \partial u \partial v$



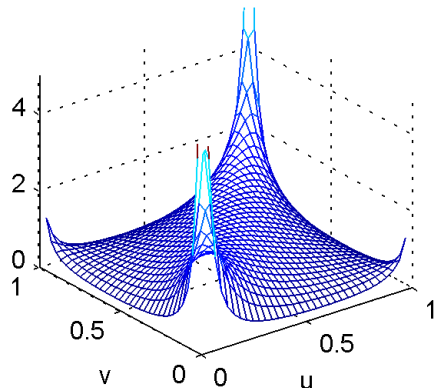
(d) Bivariate Gaussian PDF, $\rho = 0.5$



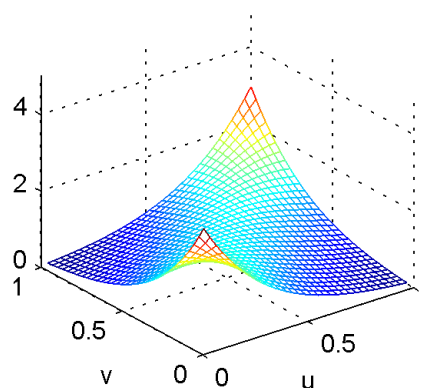
(e) copula density, $\partial^2 C / \partial u \partial v$



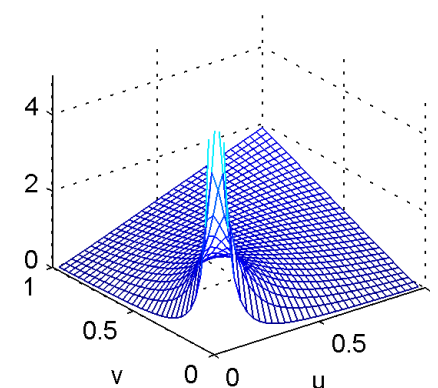
(a) Student t copula density ($\nu = 2$)



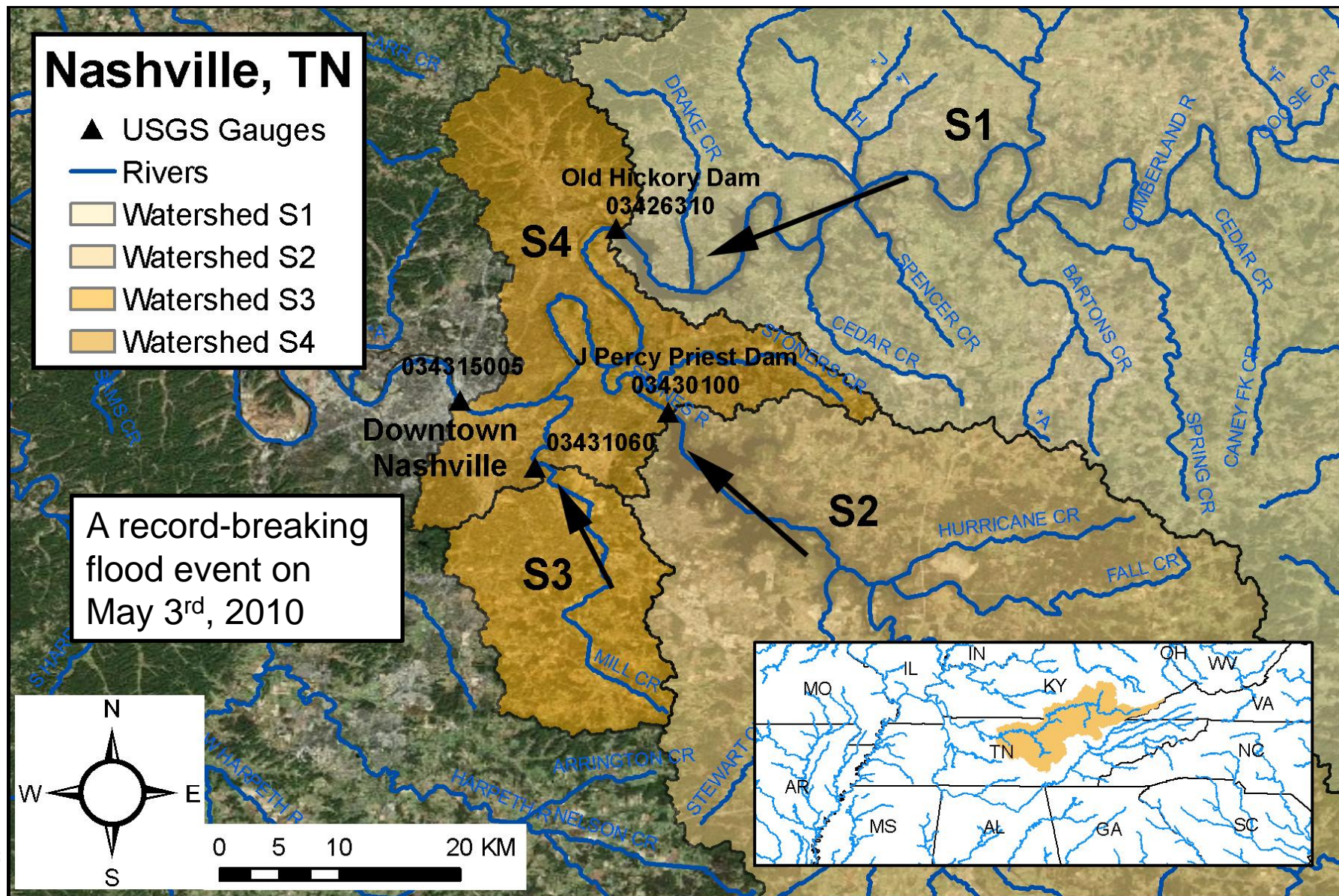
(b) Frank copula density



(c) Clayton copula density



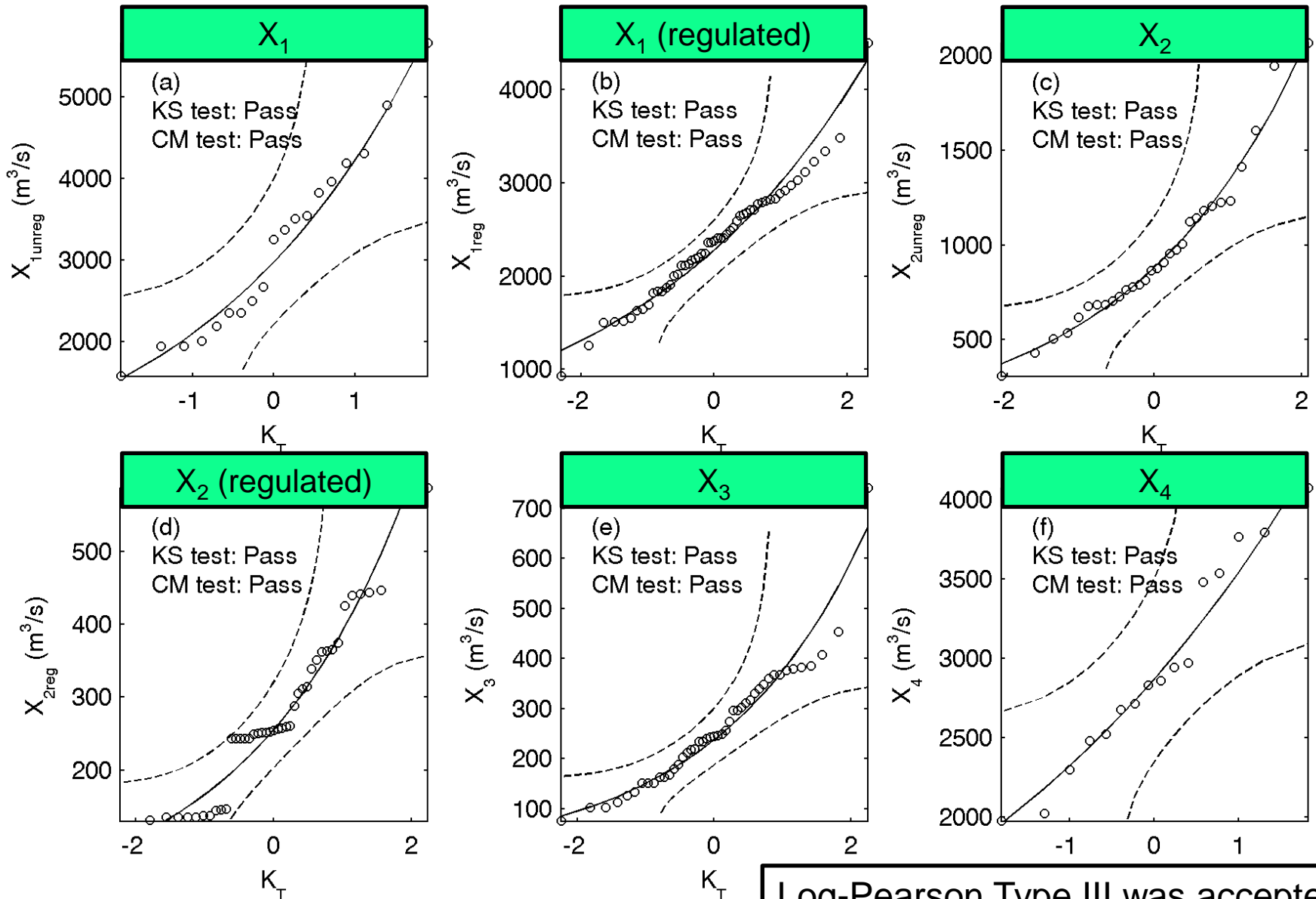
Case Study



Data Availability

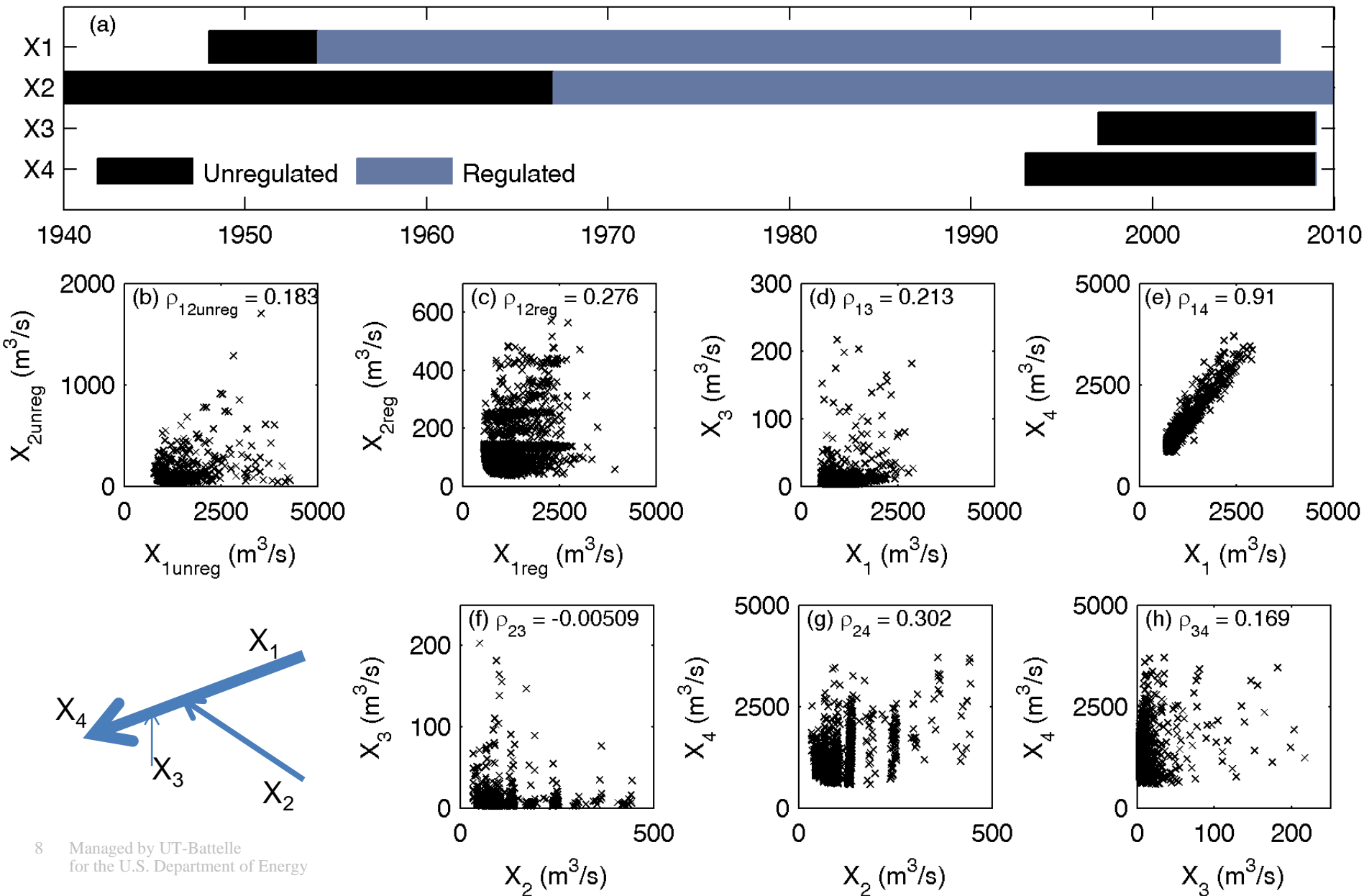
	X_1	X_2	X_3	X_4
USGS ID	03426310	03430100	03431060	034315005
Gage Name	Cumberland River at Old Hickory Dam	Stones River below J Percy Priest Dam	Mill Creek at Thompson Lane near Woodbine	Cumberland River at Woodland St at Nashville
Drainage Area (km²)	30233	2310	241.9	33307
Corresponding Watersheds	S_1	S_2	S_3	$S_1, S_2, S_3, \text{ and } S_4$
Data Coverage	WY1948~WY2007	WY1940~WY2010	WY1997~WY2009	WY1993~WY2009
# of Annual Peak Flow	19 (pre-regulated) 53 (pos-regulated)	30 (pre-regulated) 43 (post-regulated)	45 (peak flow since WY1965)	16
Mean Annual Flow (m³/s)	526.12	39.93	4.08	589.98
Note	Old Hickory Dam regulated at 1954	J Percy Priest Dam regulated at 1967		

Fitting of Marginal Distributions



Log-Pearson Type III was accepted

Correlation between High Flow Pairs



Flow Synthesization through Copulas

- **Gaussian copulas is chosen for simplification**

- **Multivariate normal distribution (MVN)**

$$\Phi_d(\phi^{-1}(u_1), \dots, \phi^{-1}(u_d)) = \int_{-\infty}^{\phi^{-1}(u_1)} \dots \int_{-\infty}^{\phi^{-1}(u_d)} (2\pi)^{-\frac{d}{2}} |\Sigma|^{-\frac{1}{2}} \exp(-\mathbf{z}\Sigma^{-1}\mathbf{z}^T / 2) dz_1 \dots dz_d$$

- **Gaussian Copulas**

$$C_{U_1, \dots, U_d}(u_1, \dots, u_d) = \Phi_d(\phi^{-1}(u_1), \dots, \phi^{-1}(u_d)) == \Phi_d(\phi^{-1}(F_{X_1}(x_1)), \dots, \phi^{-1}(F_{X_d}(x_d)))$$

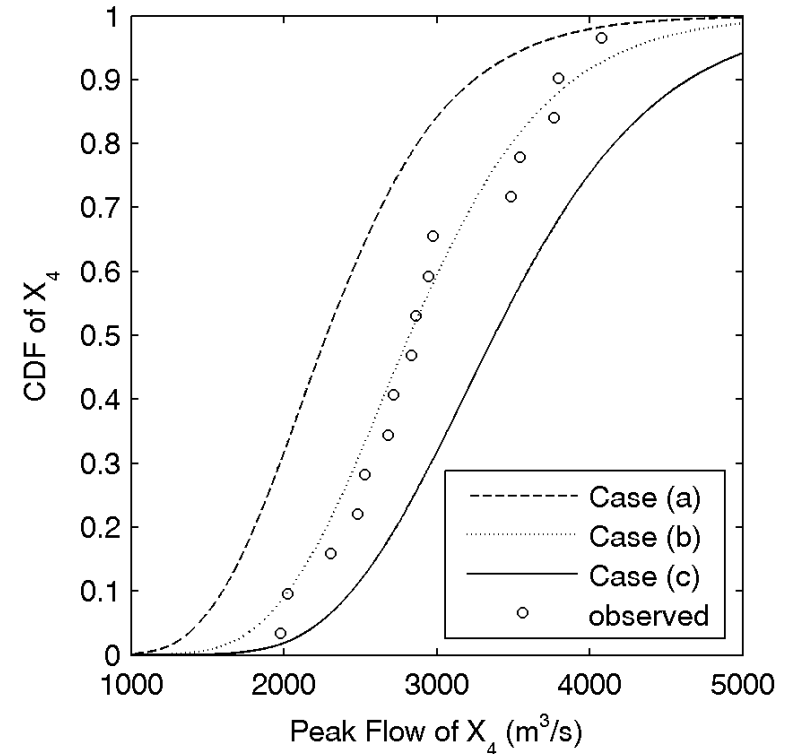
- **Existing MVN generators are easy to use**

- **Procedures**

- (1) Calculate the correlation matrix for MVN
- (2) Generate 100,000 MVN samples
- (3) Transform the MVN samples to Gaussian copulas, and then to different marginals
- (4) The synthesized (x_1, x_2, x_3) flow are then used to estimate the flood frequency at downstream reaches.

Evaluation

- Three synthesizing functions were tested:
 - (1) $X_4 = X_1$
 - (2) $X_4 = X_1 + X_2 + X_3$
 - (3) $X_4 = X_1 + X_2 + w * X_3$
- Validate by observed X_4 flow
 - Function (2) works the best
 - More suitable function can be considered in the future

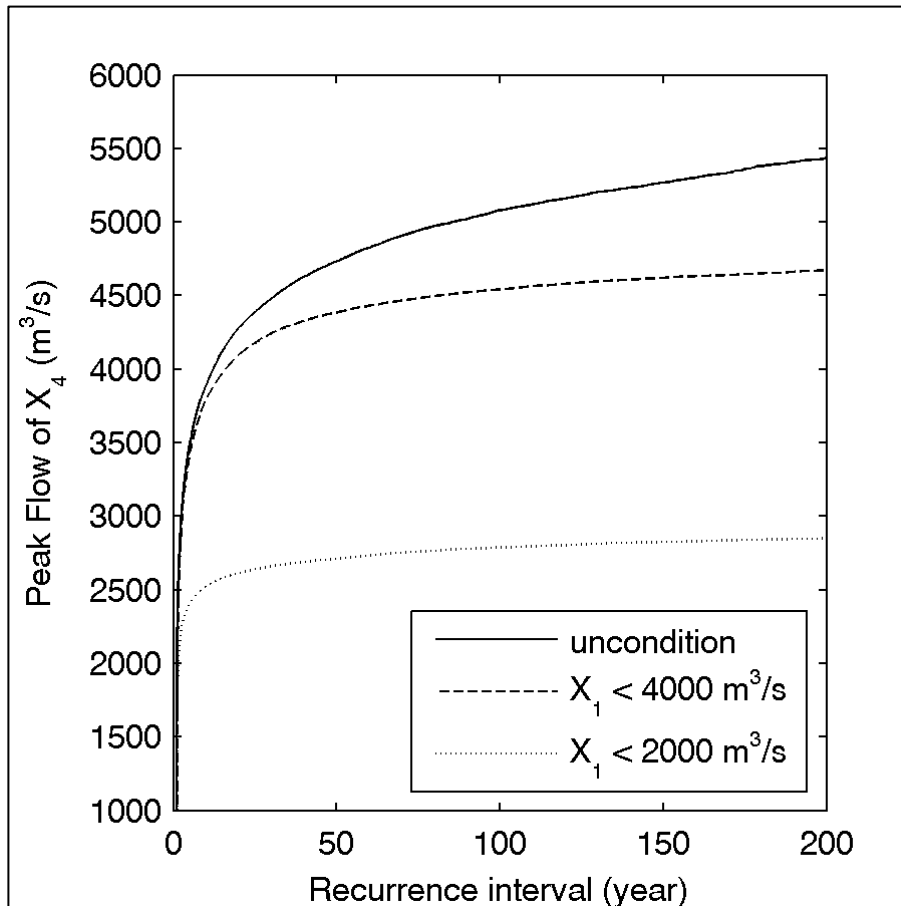


	KS (5%)	CM (5%)	Nash E	R ²	RMSE (m ³ /s)	PE (%)	May 3 rd , 2010
(a) X_1	Reject	Reject	0.0703	0.961	592.9	-20.5	885.3 yrs
(b) $X_1 + X_2 + X_3$	Accept	Accept	0.9363	0.964	155.2	-1.8	166.3 yrs
(c) $X_1 + X_2 + w * X_3$	Reject	Reject	0.0923	0.963	585.9	17.5	29.5 yrs

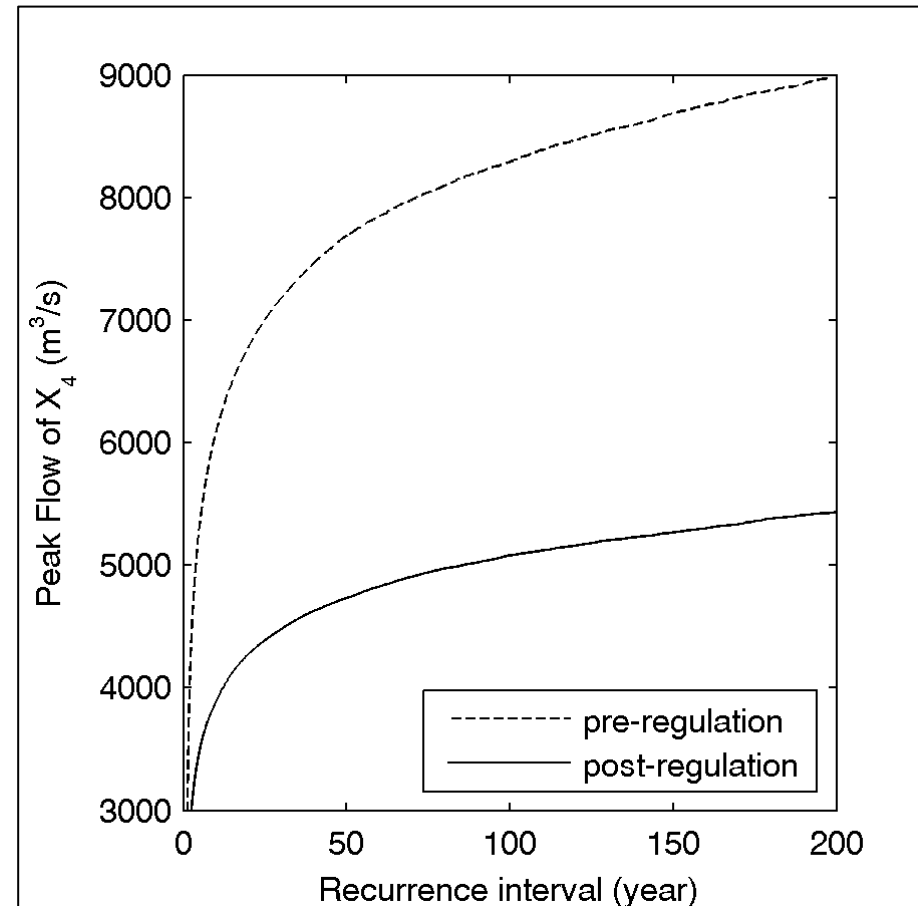
Flexibility of the Multivariate Approach

Frequency of Flood Considering Dam Regulation

$$H_{X_2, X_3 | X_1 \leq a} = C_{U_2, U_3 | U_1 \leq F_{X_1}(a)}$$



Flood Frequency before and after dam regulation



Conclusions

- **The multivariate flood frequency is more flexible, especially for river confluences considering dependence structure (comparing to the univariate statistical approach)**
- **It requires minimal data and moderate computation efforts (comparing to the modeling approach)**
- **Challenges and Future Works**
 - **What will be the best criteria to construct dependence structure for multivariate flood frequency analysis?**
 - **Dimensionality remains a major challenge. Gaussian (and t) copulas are the easiest but may not be the best solution.**
 - **How can we consider climate and land use / land cover change into this framework?**

**Thank you
Questions?**

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