

On the Stationarity of Flood Peaks in the Continental US and Central Europe

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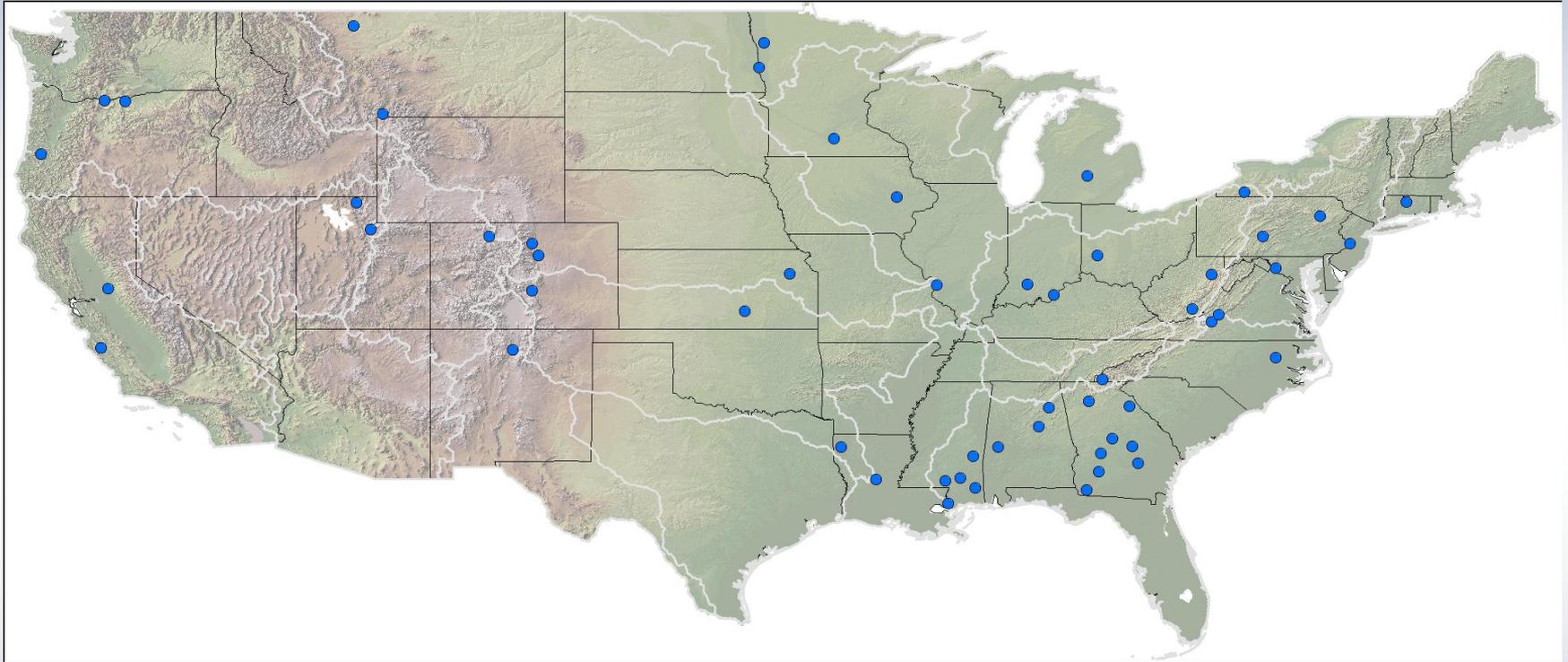
Problem Statement and Objectives

- **There is a large spatial variability in flood hazards over the continental United States and central Europe**
- **Profound anthropogenic changes (e.g., changes in land use / land cover, construction of dams, river engineering)**
- **Projected increasing frequency of extreme rain and flood with human induced climate change**
- **Rapidly changing flood hazards with urbanization**

Central Issues:

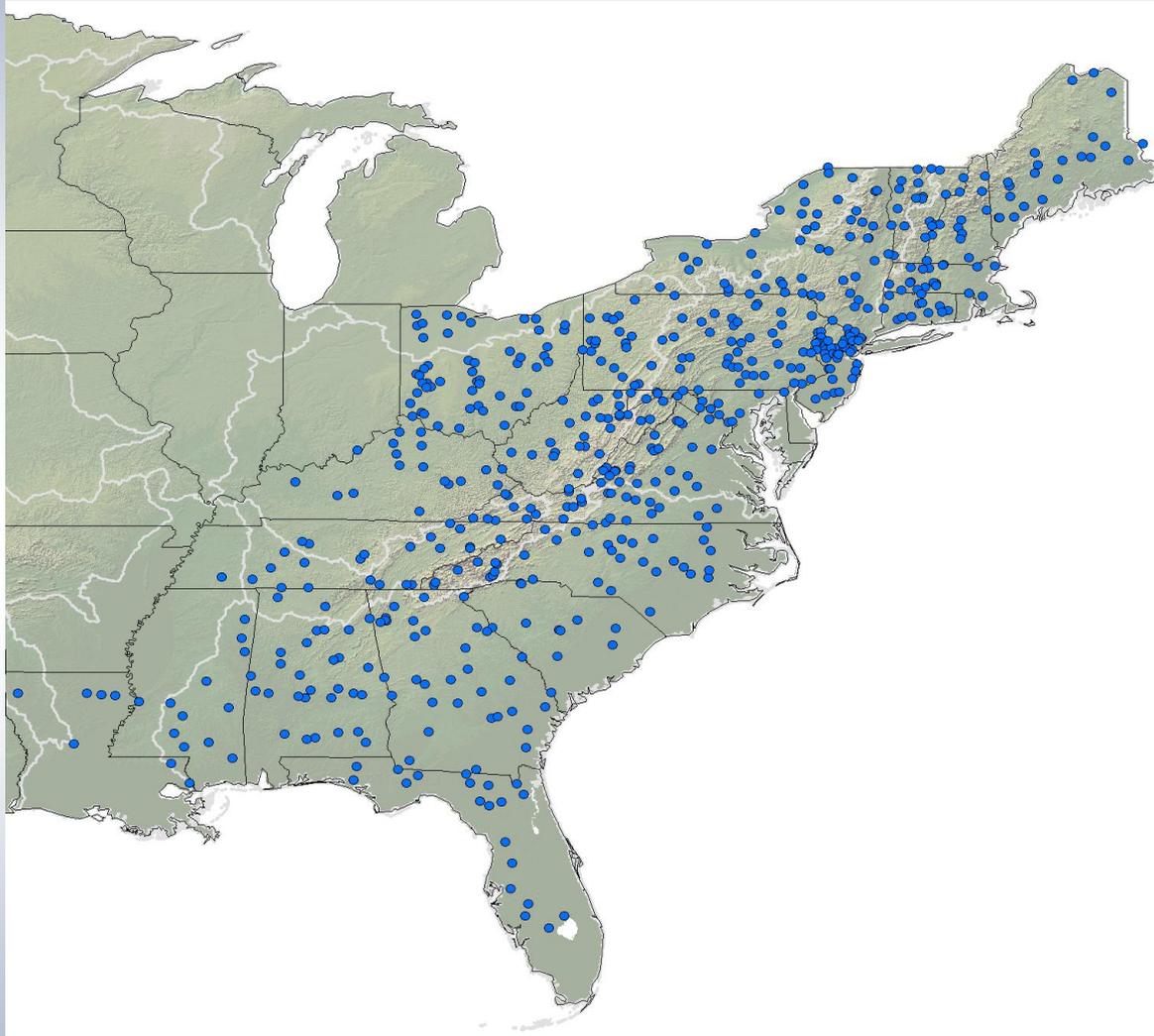
- **temporal non-stationarities of annual flood peaks**
- **changes in frequency of extreme flooding due to climate change**
- **flood hazard assessment in urban environment**
- **“Mixtures” of flood peaks and impact of tropical cyclones on flood peak distributions**

Continental US



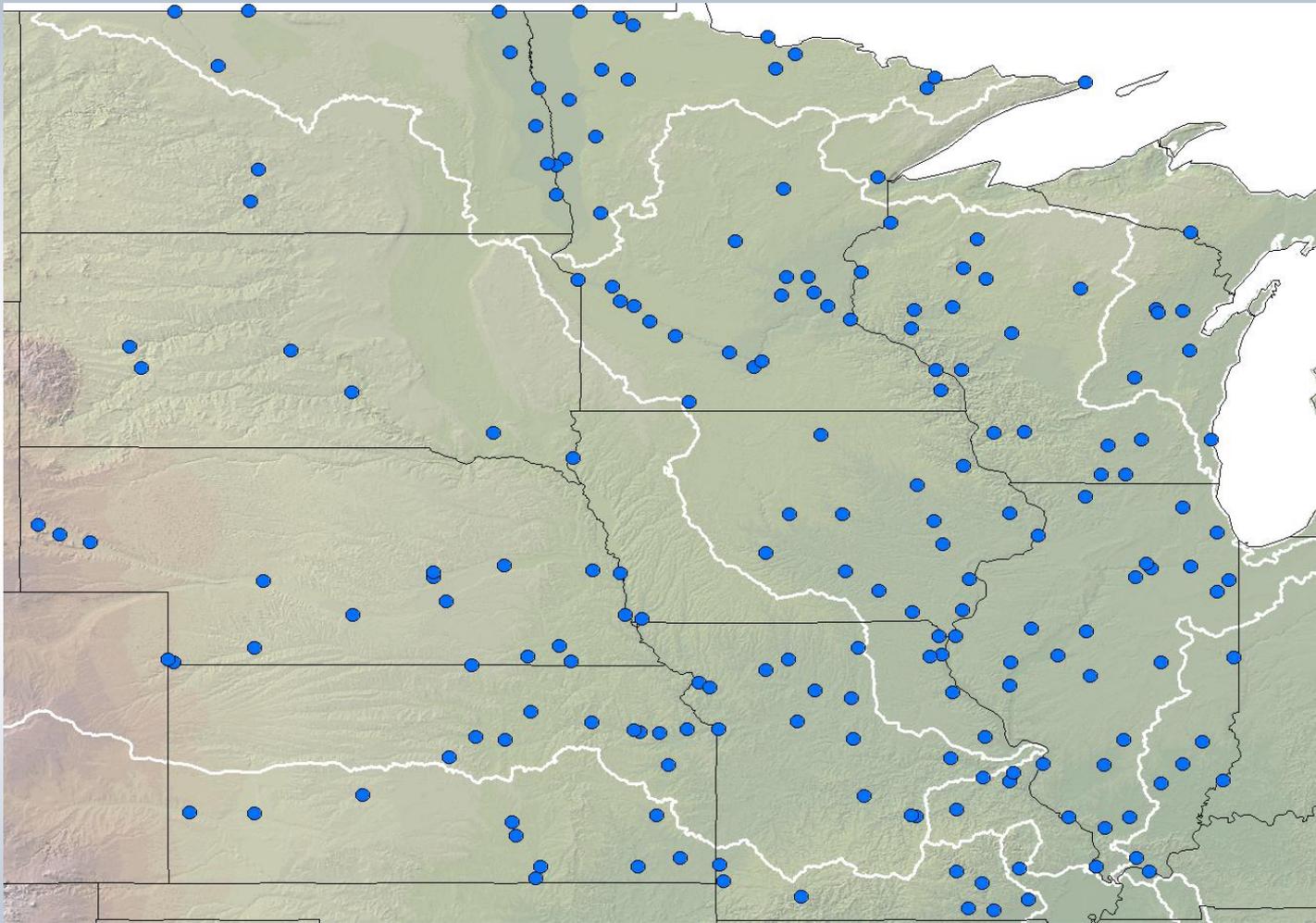
50 USGS streamflow stations with at least 100-year record length of annual maximum peak discharge

Eastern US



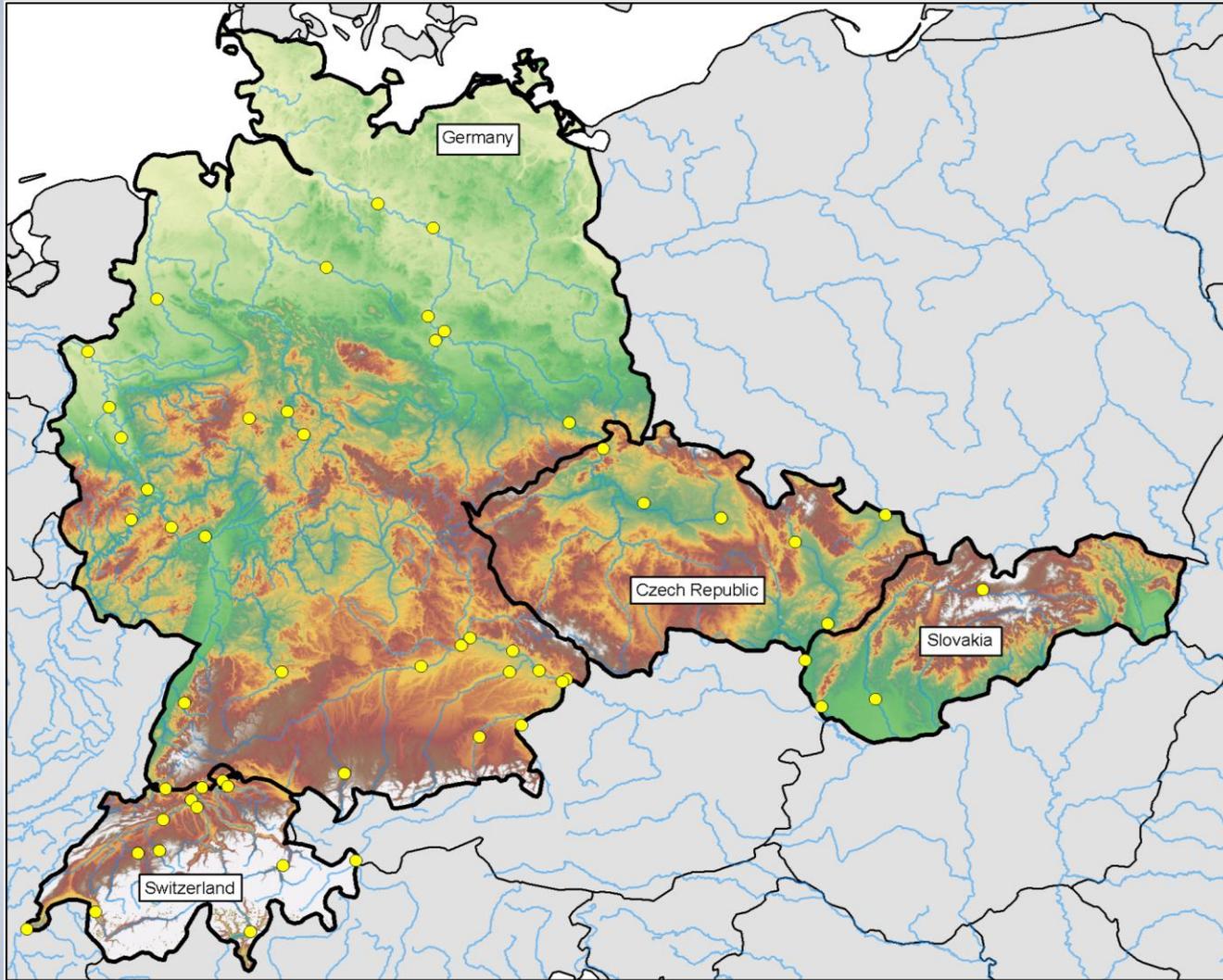
572 USGS streamflow stations with at least 75-year record length of annual maximum peak discharge

Central US



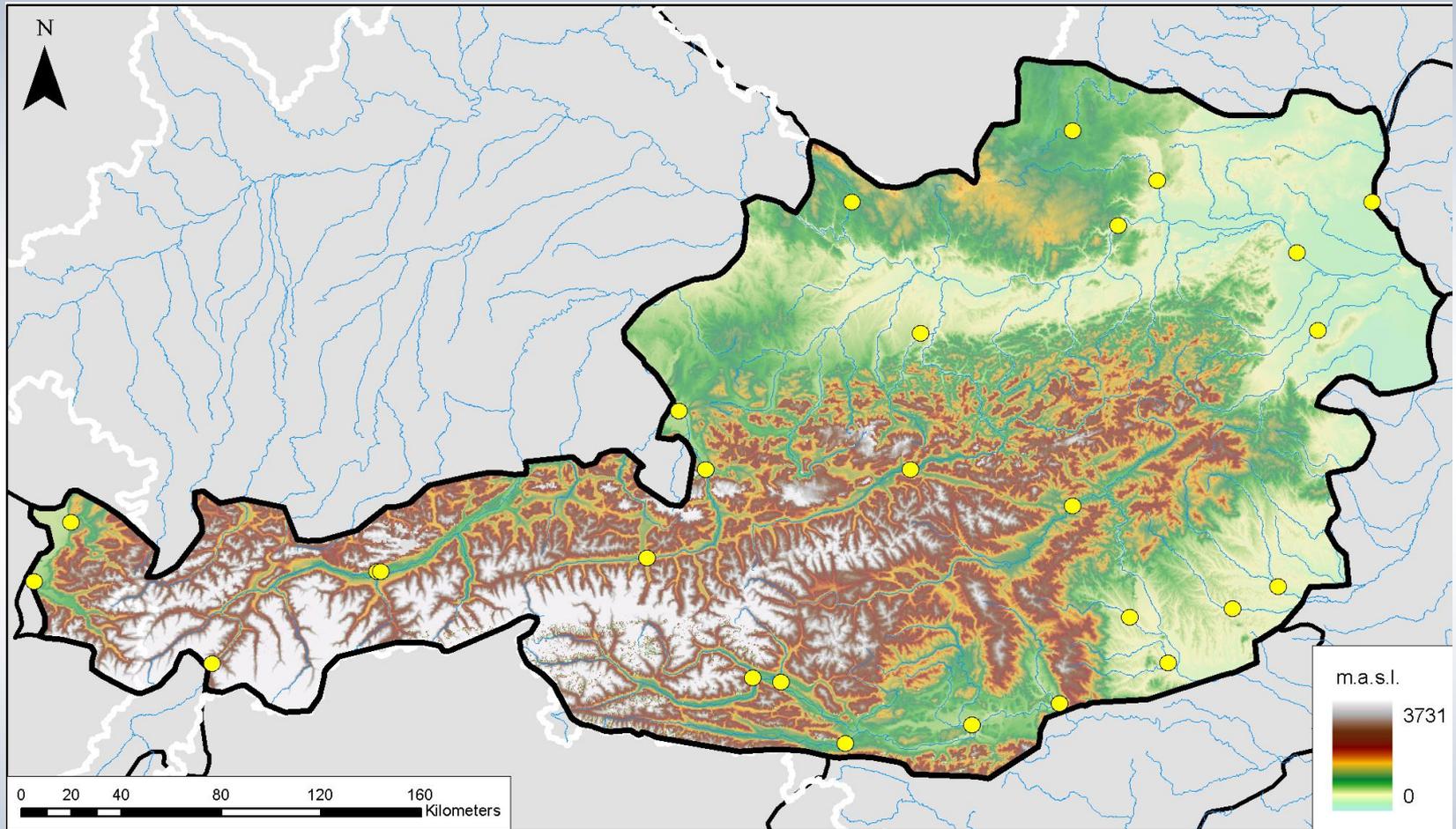
196 USGS streamflow stations with at least 75-year record length of annual maximum peak discharge

Central Europe



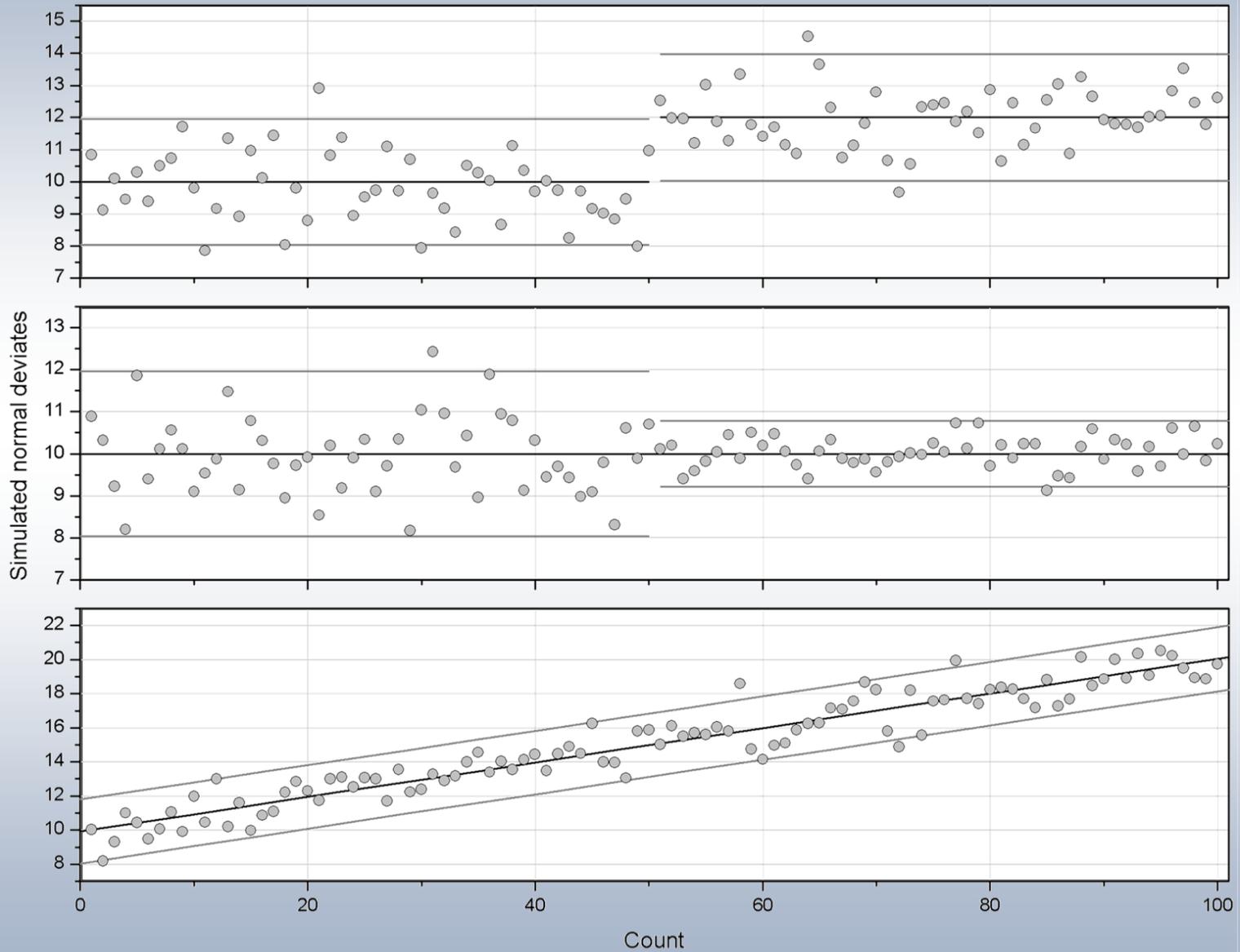
55 streamflow stations with at least 75-year record length of annual maximum daily peak discharge

Austria



27 streamflow stations with annual maximum daily peak discharge from 1951 to 2006 (56 years)

Stationarity

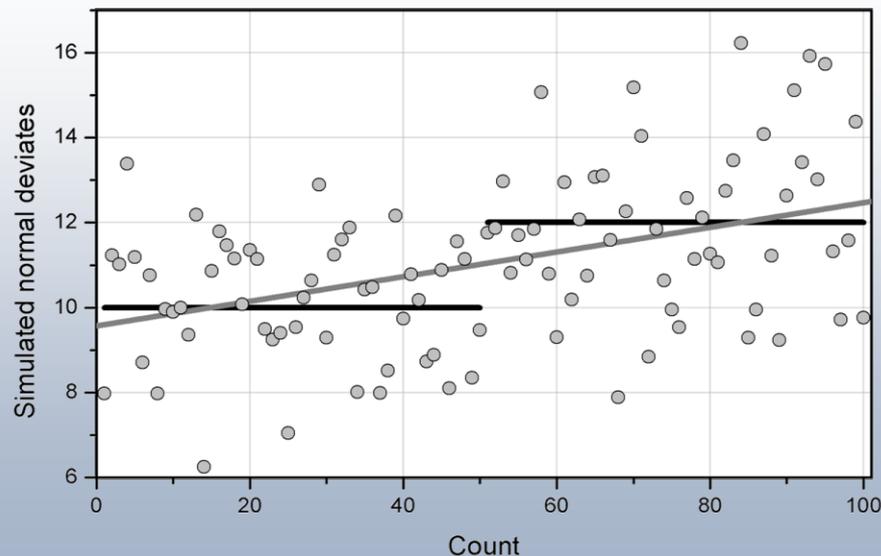


Stationarity

The stationarity of a time series is evaluated by checking for the presence of abrupt and/or slowly varying linear changes over time:

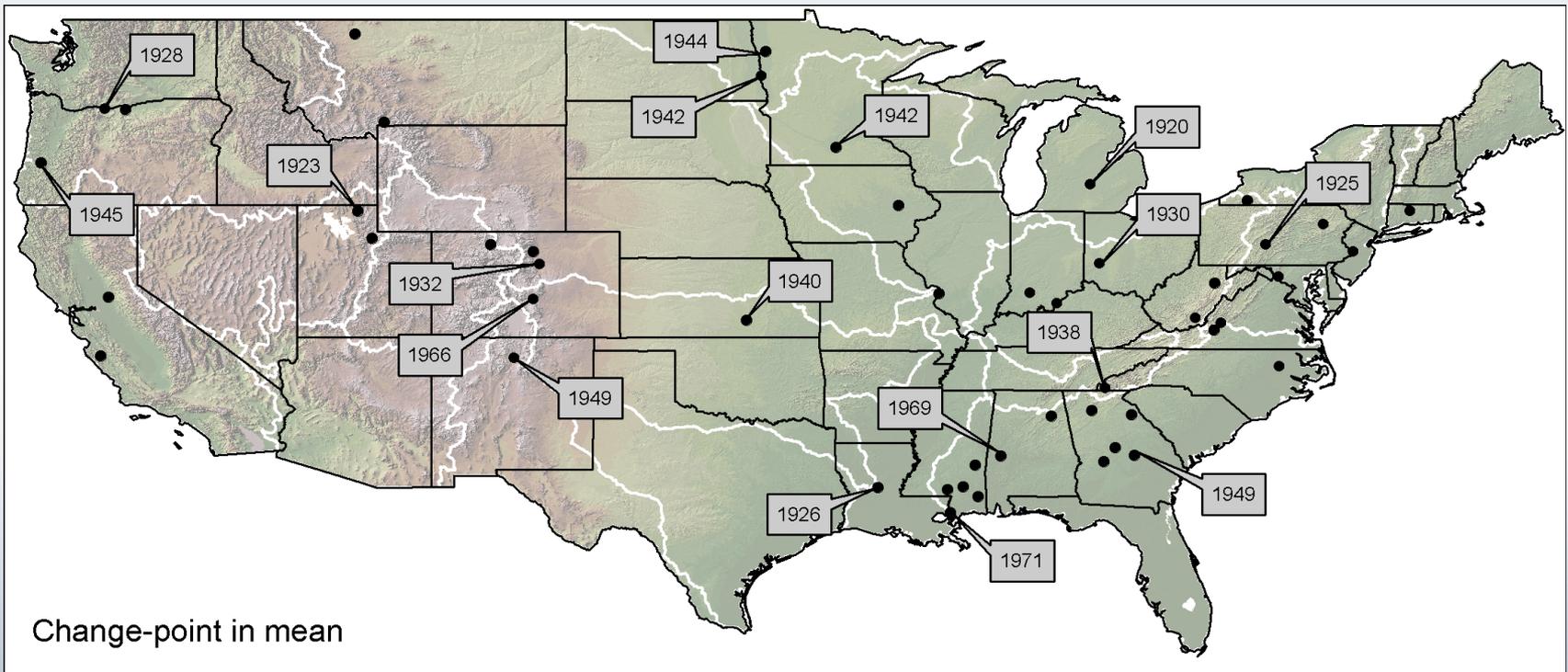
-Change-point (cp) in mean and variance investigated by means of the non-parametric Pettitt test

-Presence of monotonic trends investigated by means of the non-parametric Mann-Kendall and Spearman tests

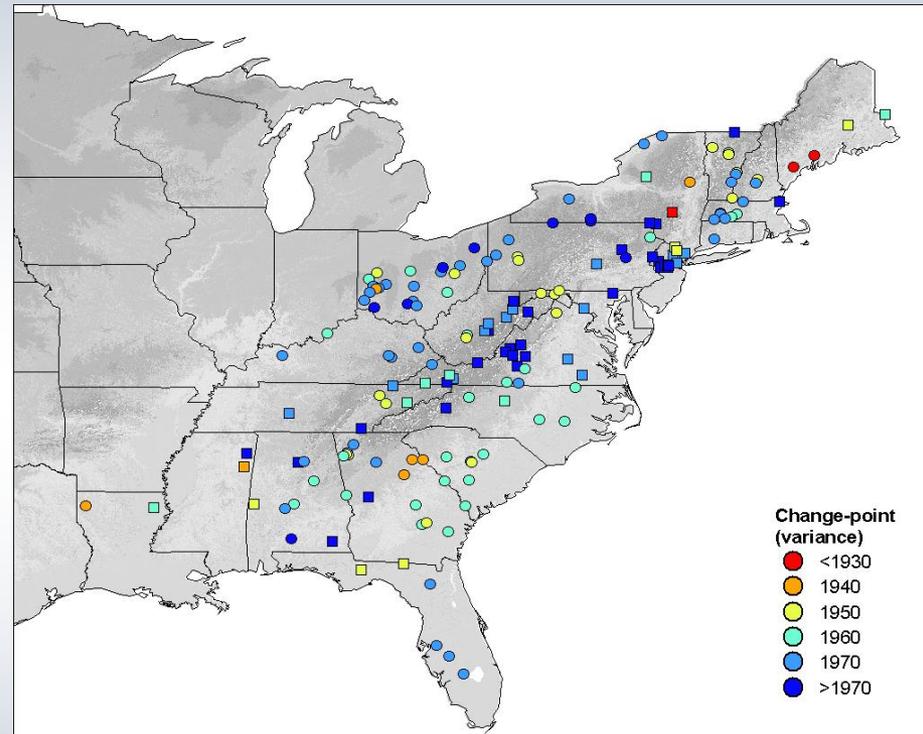
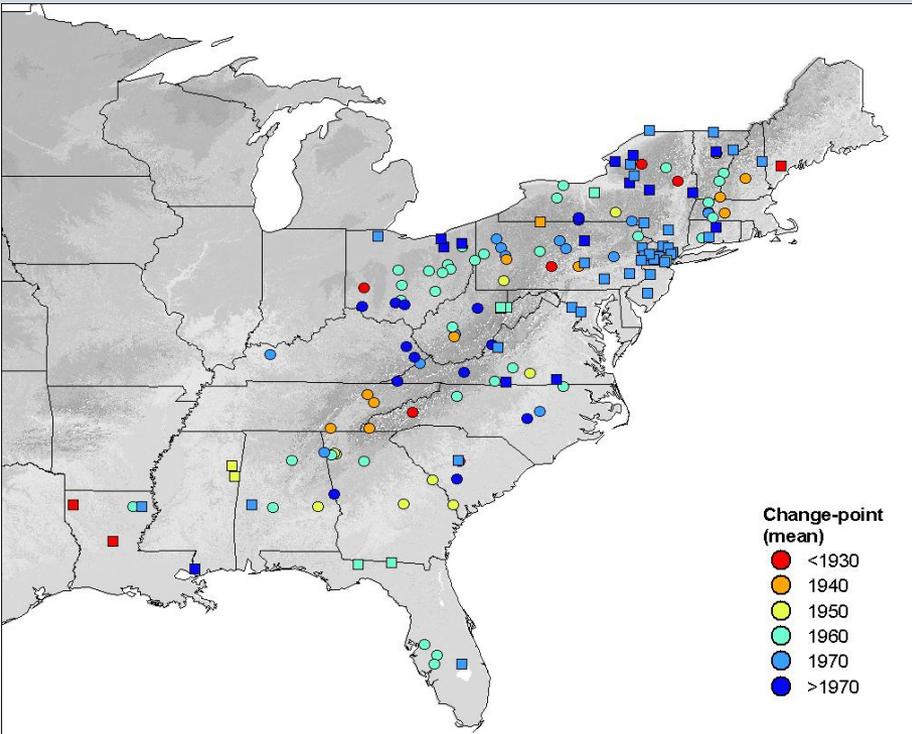


Stationarity: Continental US

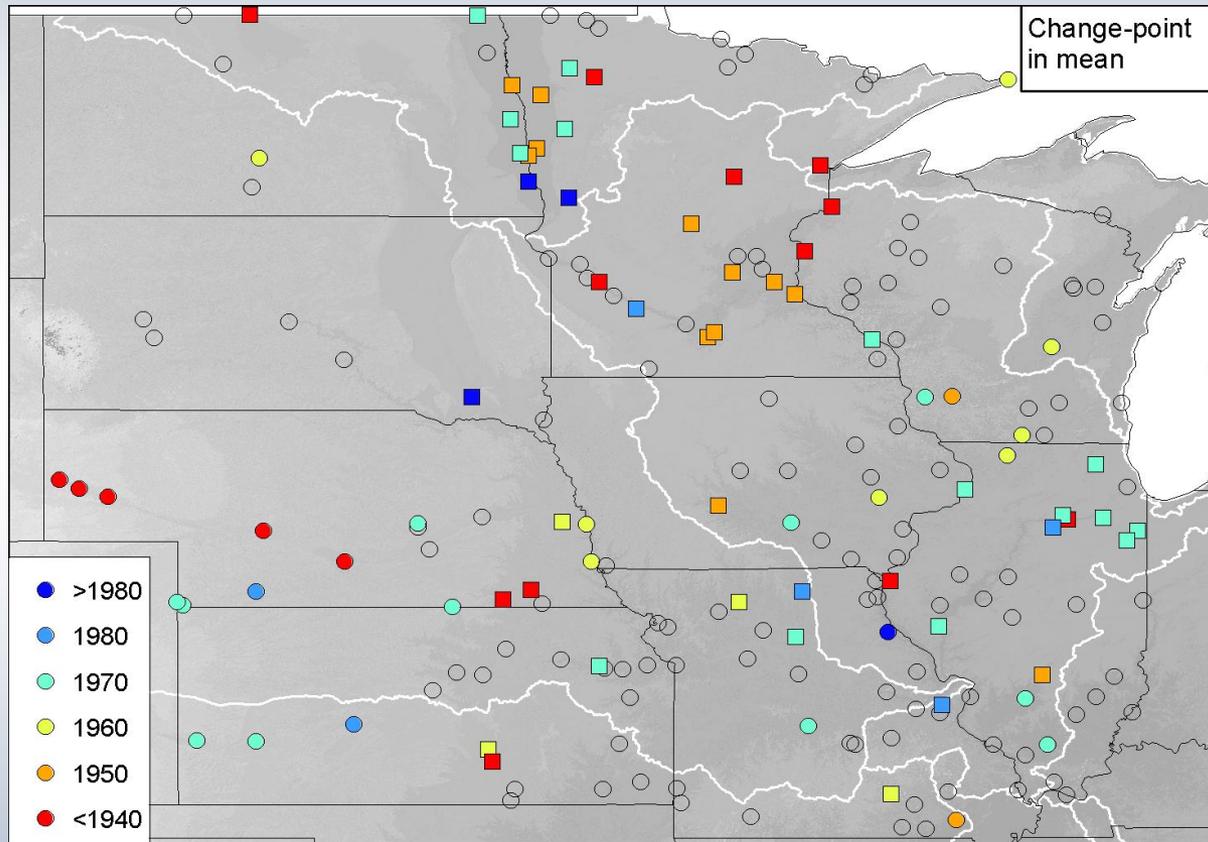
Most of the violations of the stationarity assumption are associated with construction of dams, streamgauge relocation, and changes in land use/land cover.



Stationarity: Eastern US

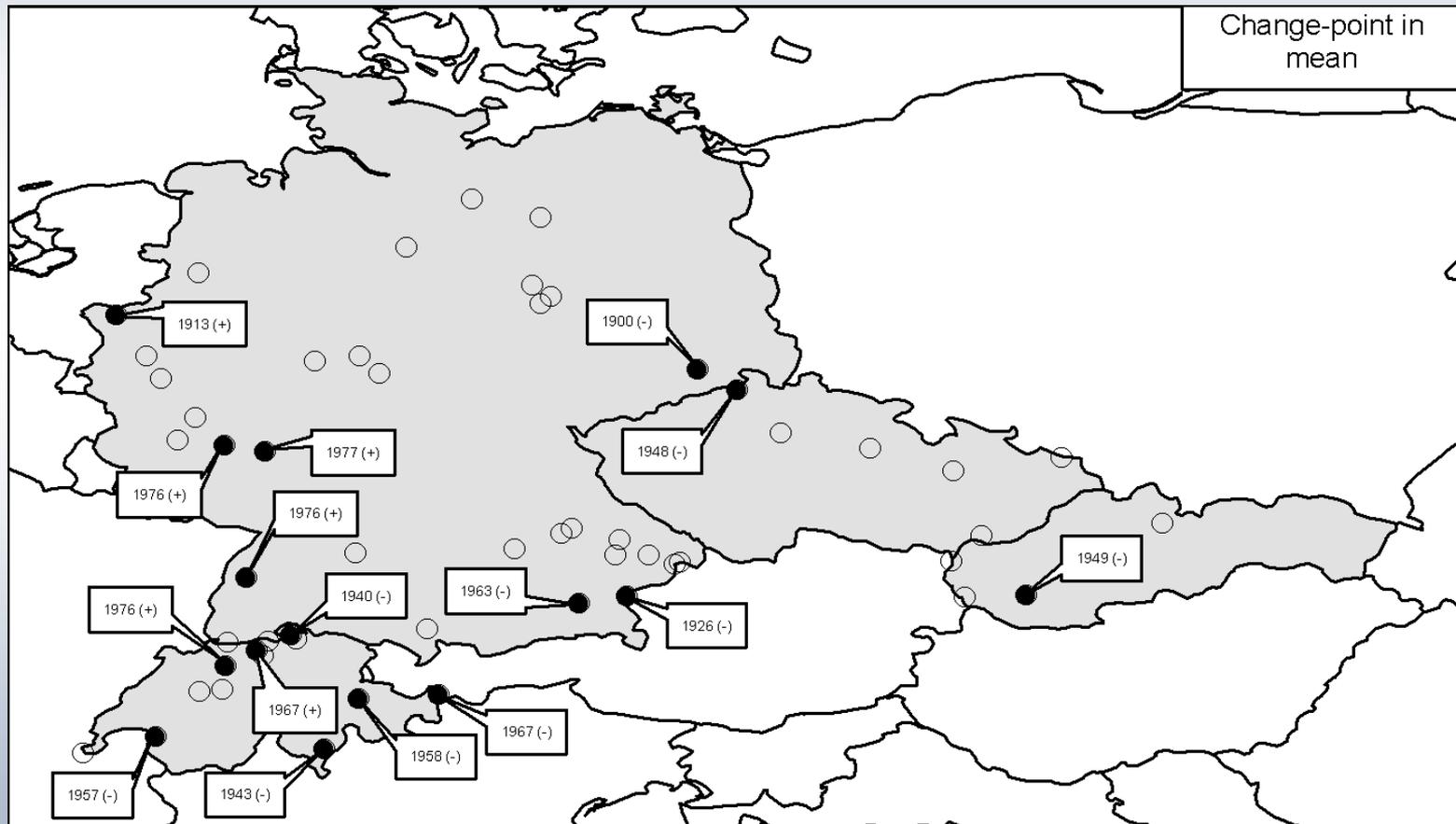


Stationarity: Central US

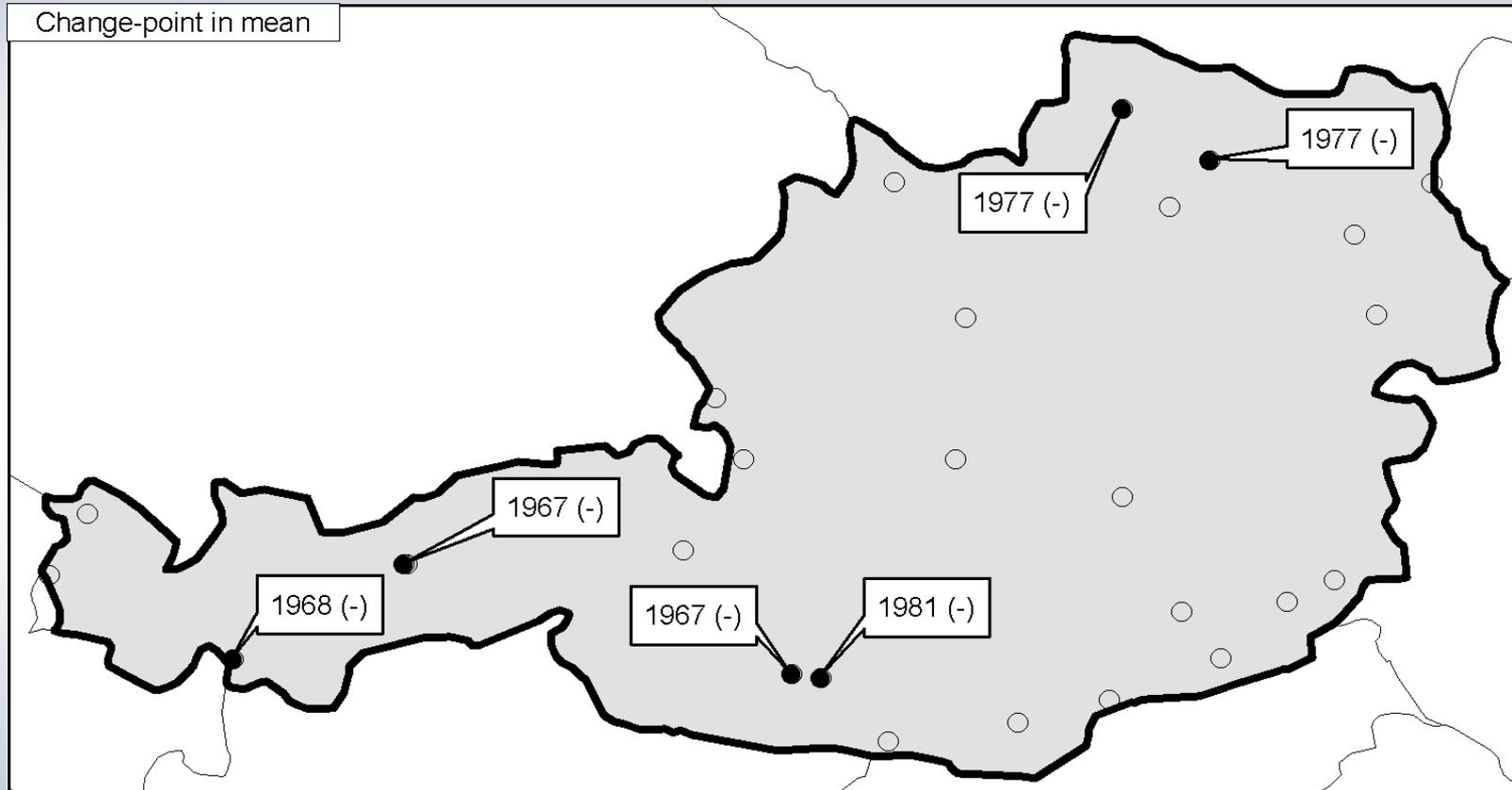


Stationarity: Central Europe

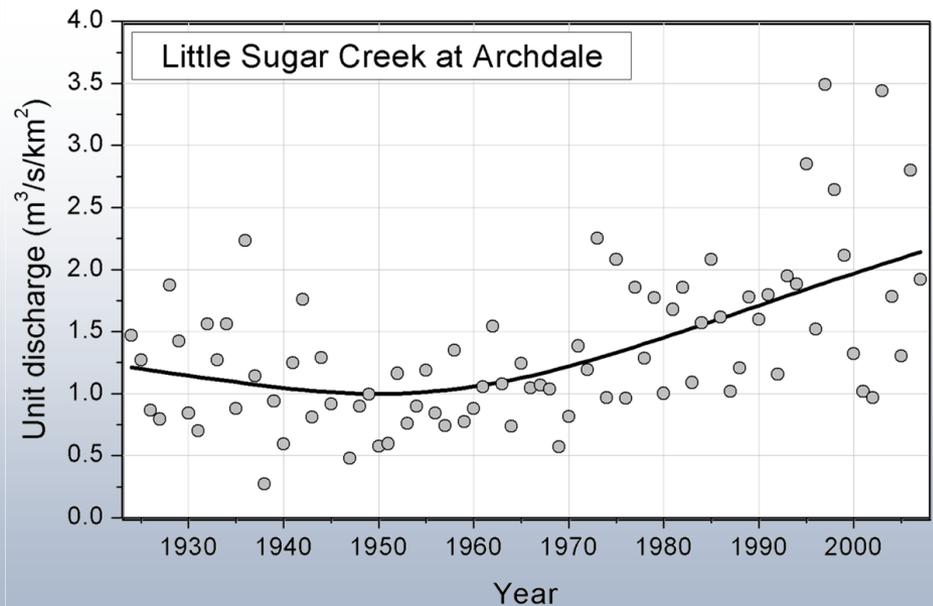
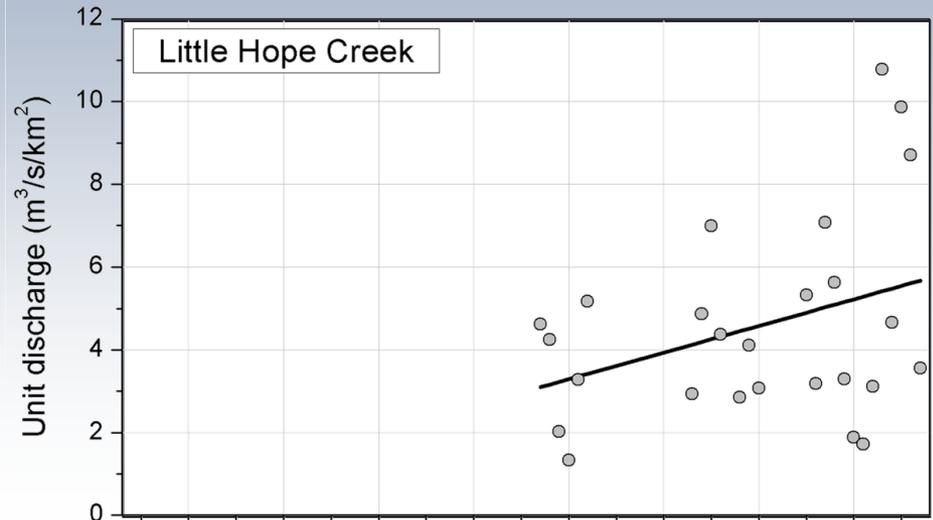
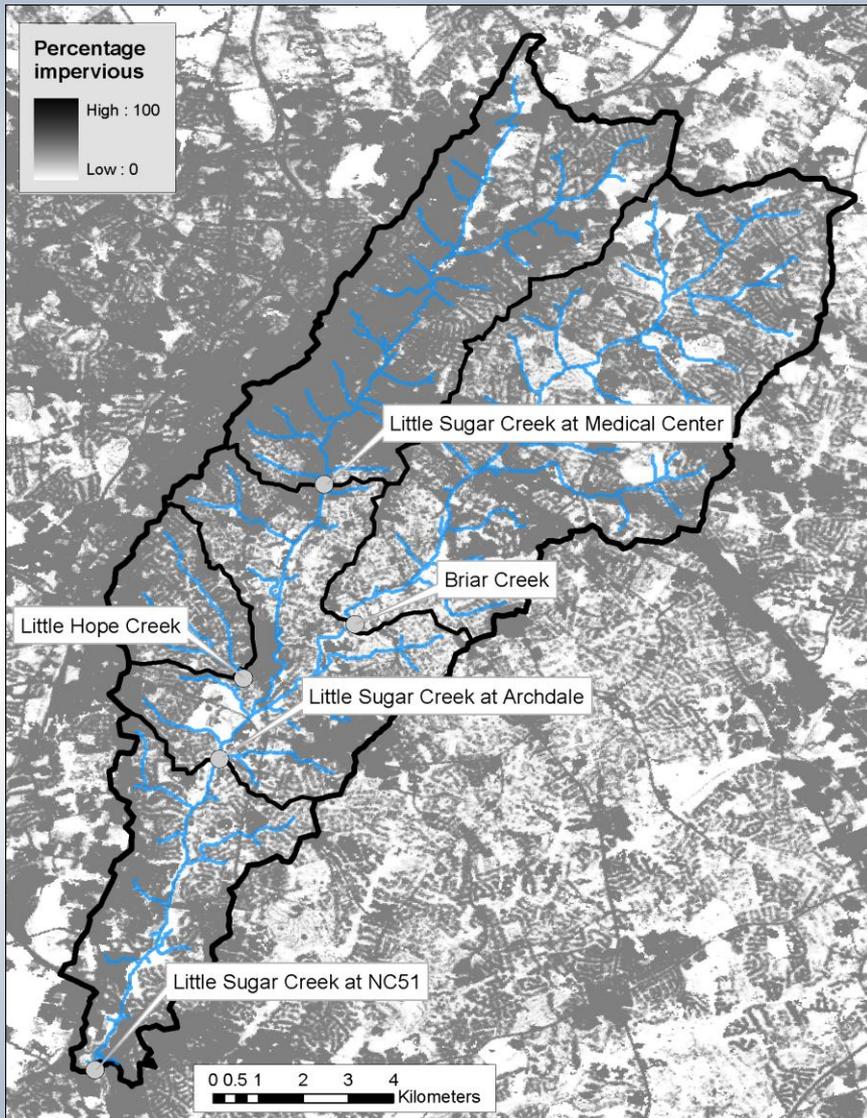
Most of the violations of the stationarity assumption are associated with construction of dams and river engineering (e.g., straightening, diversion, meander cut-off).



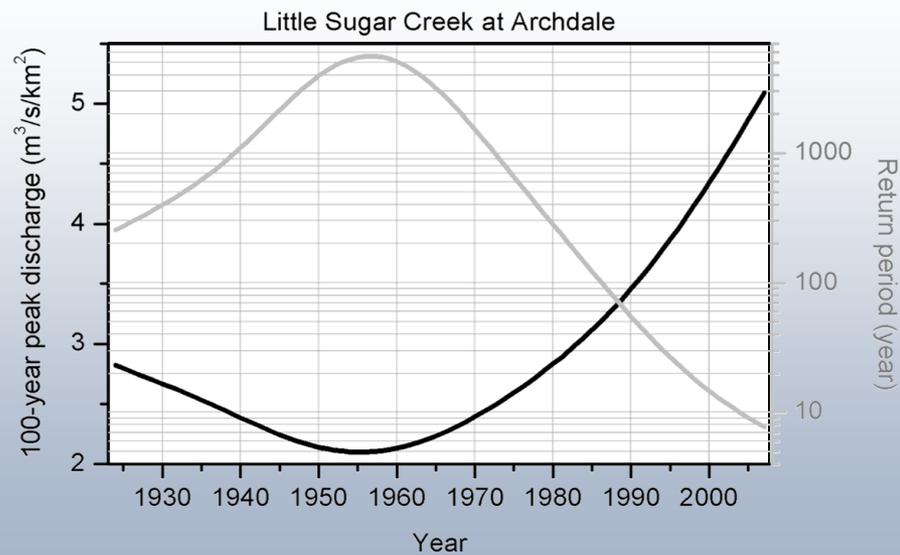
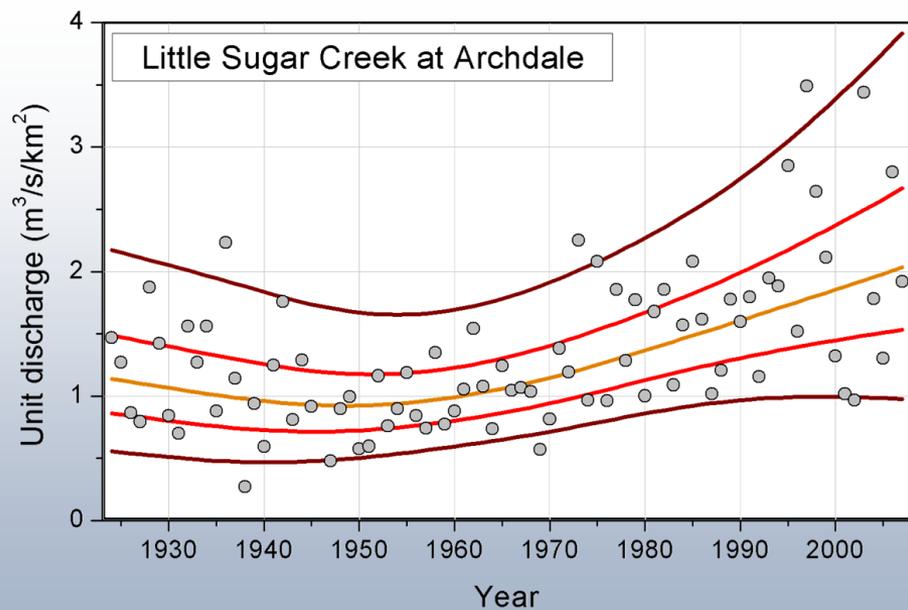
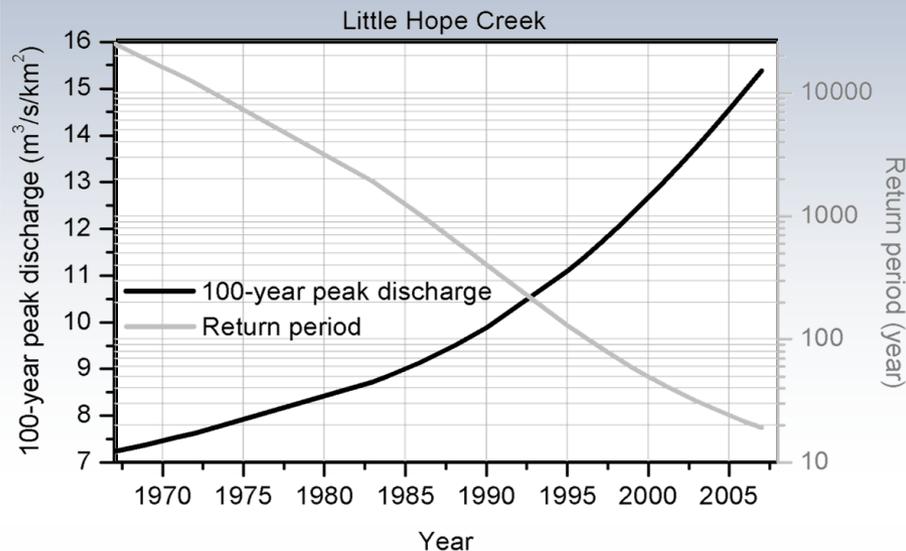
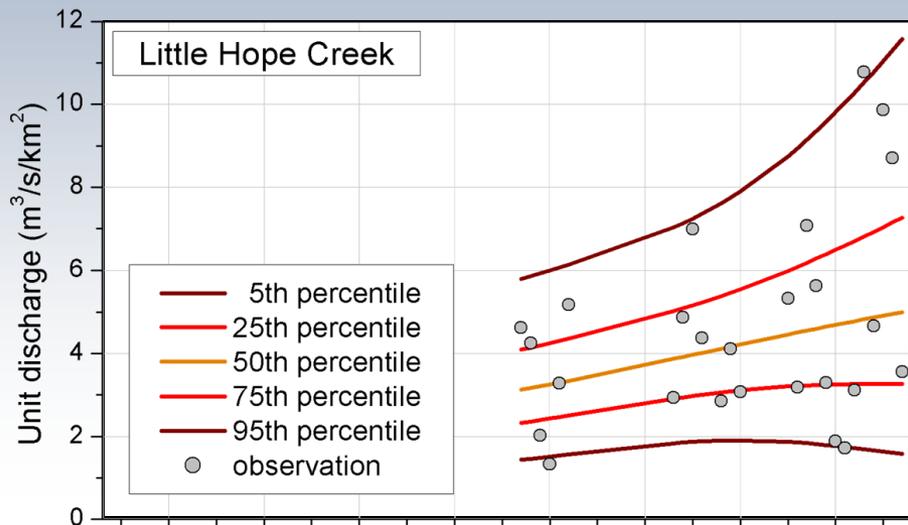
Stationarity: Austria



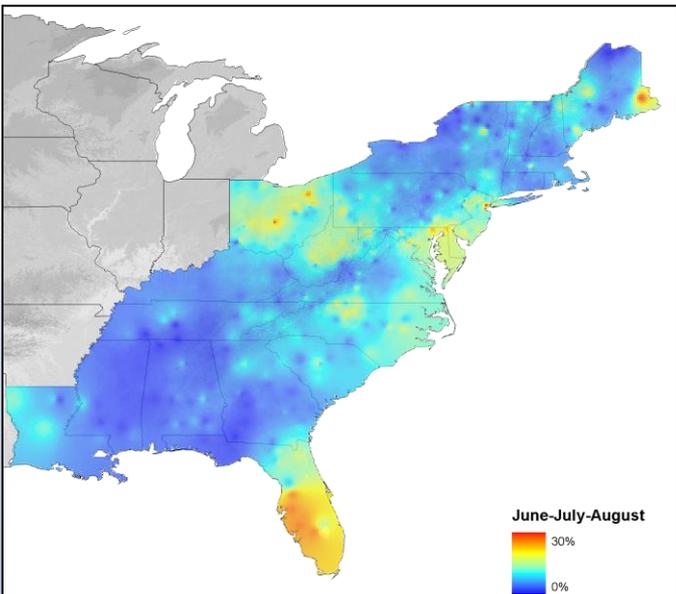
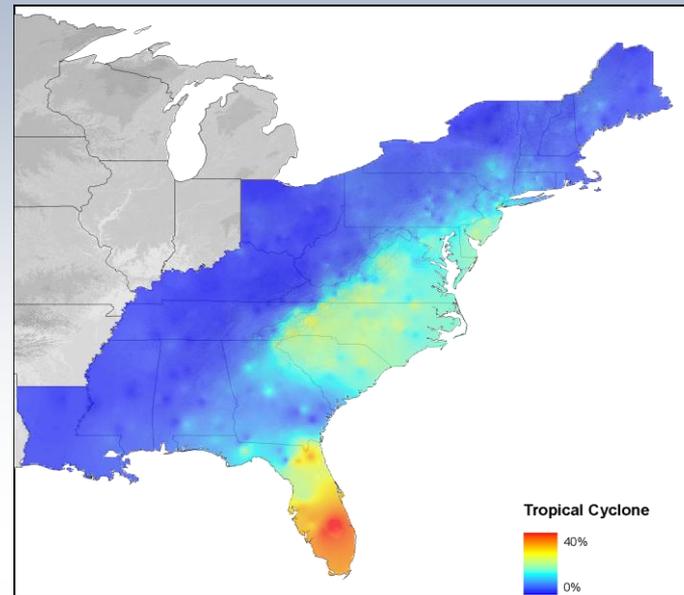
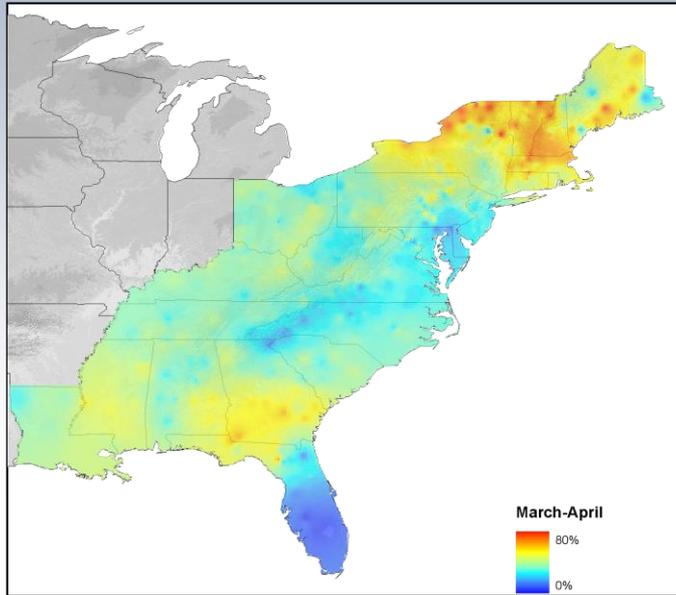
Flood Frequency – Charlotte NC



Nonstationary Modeling



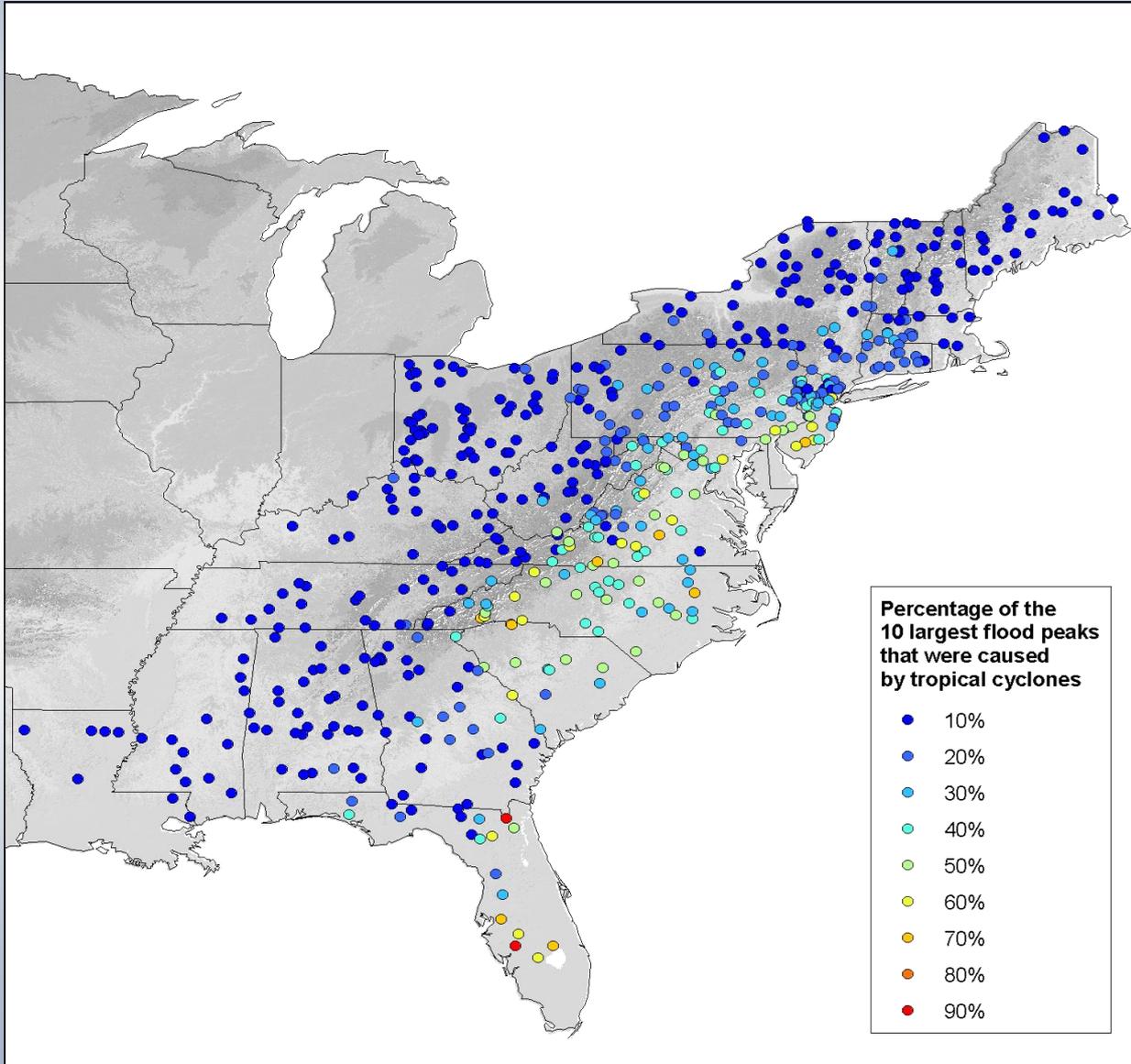
Mixture Distributions



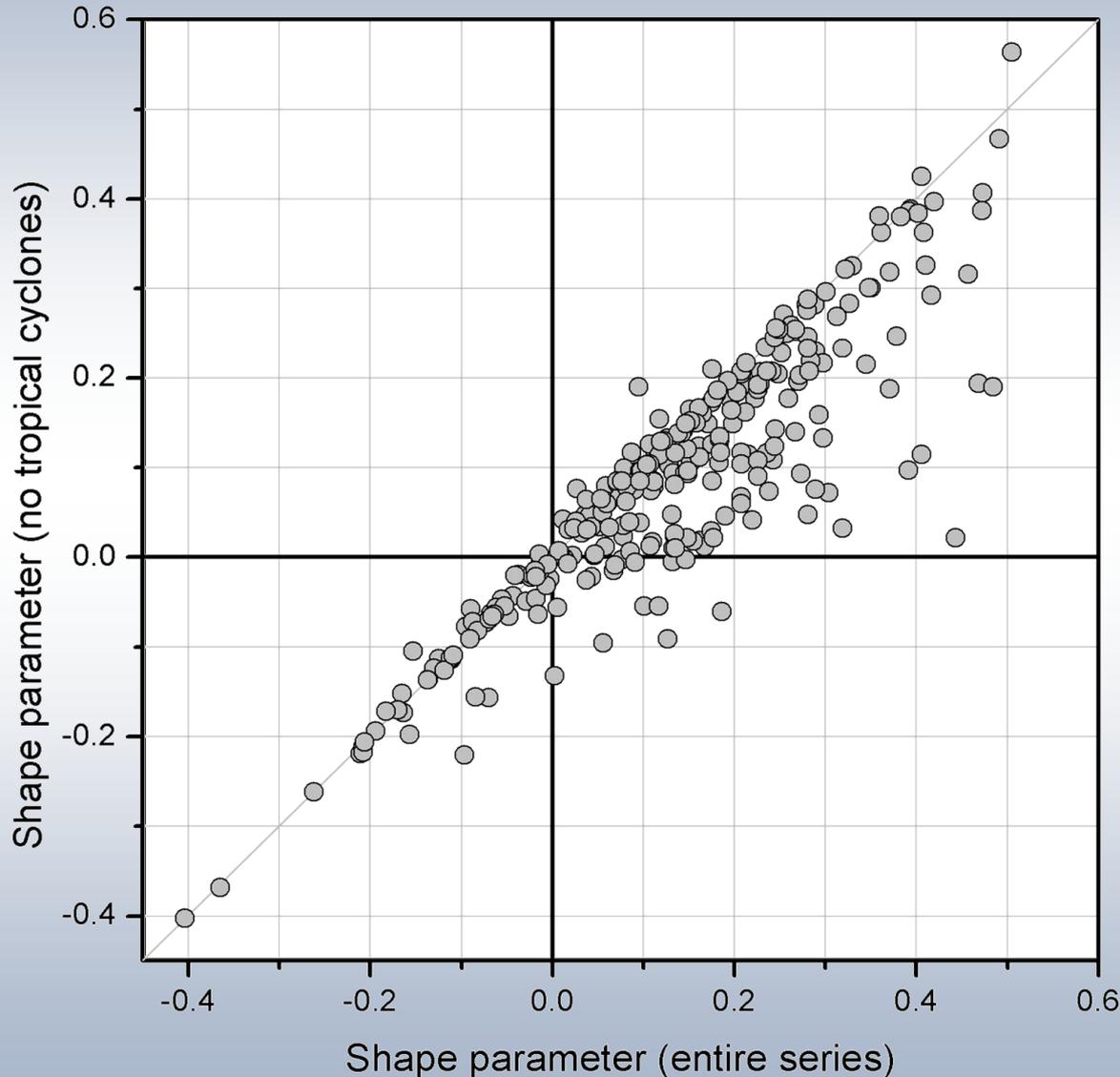
“Mixtures” and spatial heterogeneities of flood peak distributions

Villarini, G., and J.A. Smith, Flood peak distributions for the Eastern United States, *Water Resources Research*, doi:10.1029/2009WR008395, 46, W06504, 2010

Impact of Tropical Cyclones

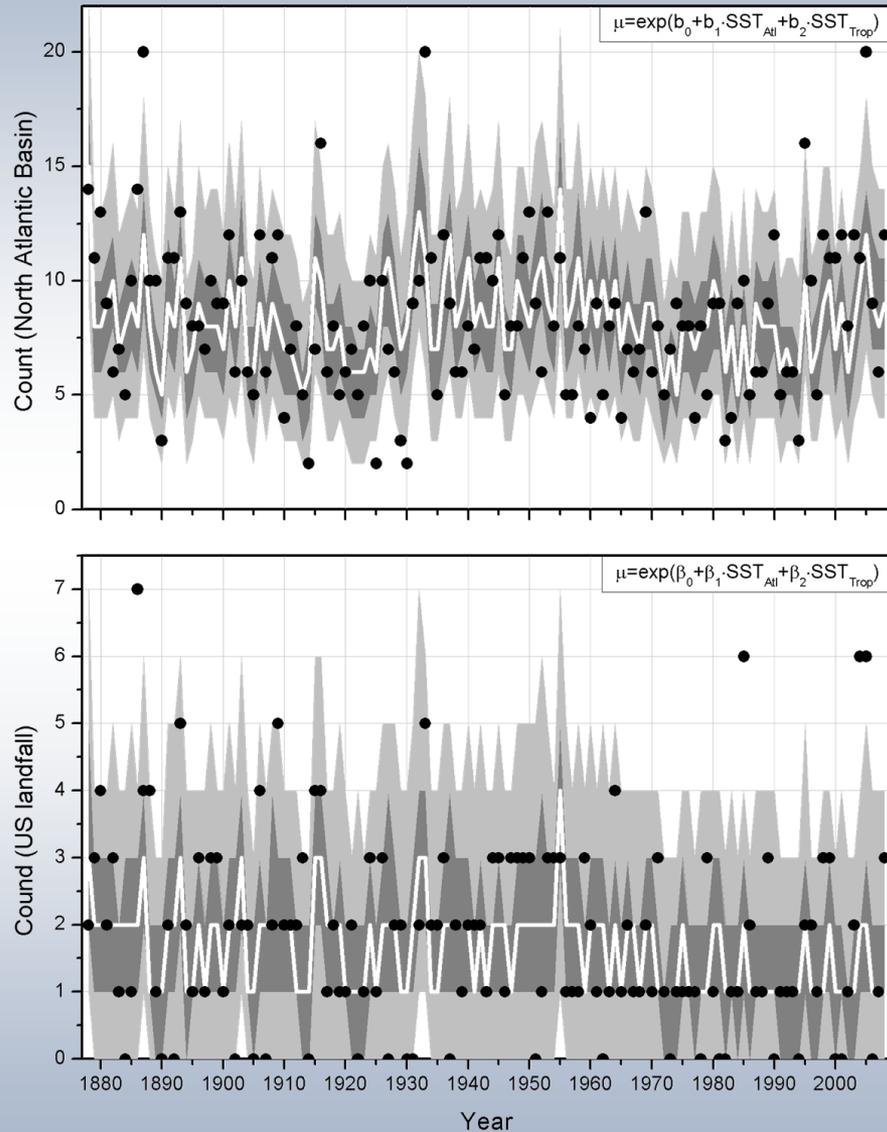


Upper Tail and Tropical Cyclones



The shape parameter of the GEV distribution is used to evaluate the impact of tropical cyclones on the upper tail of the flood peak distribution

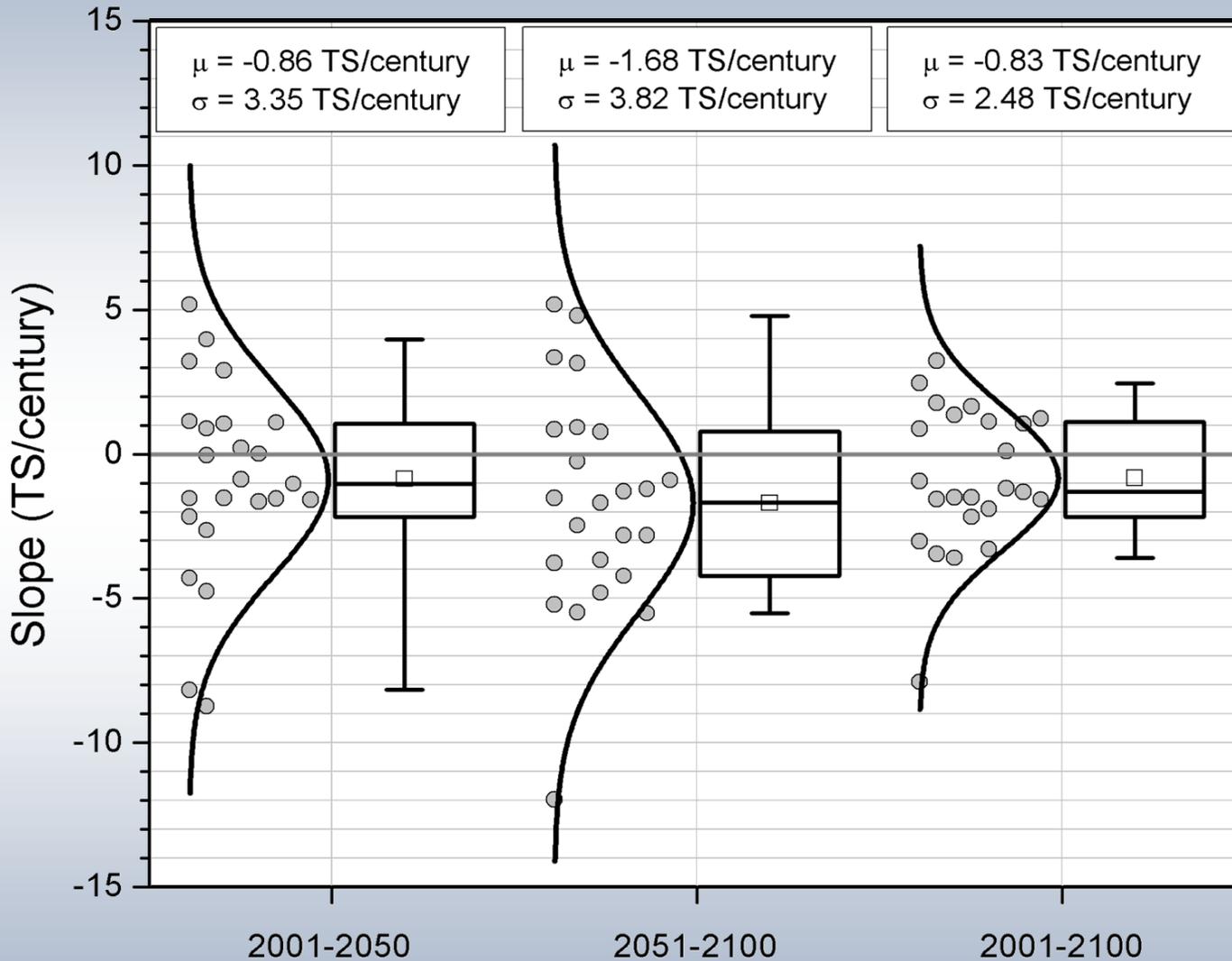
Modeling of Tropical Storm Count



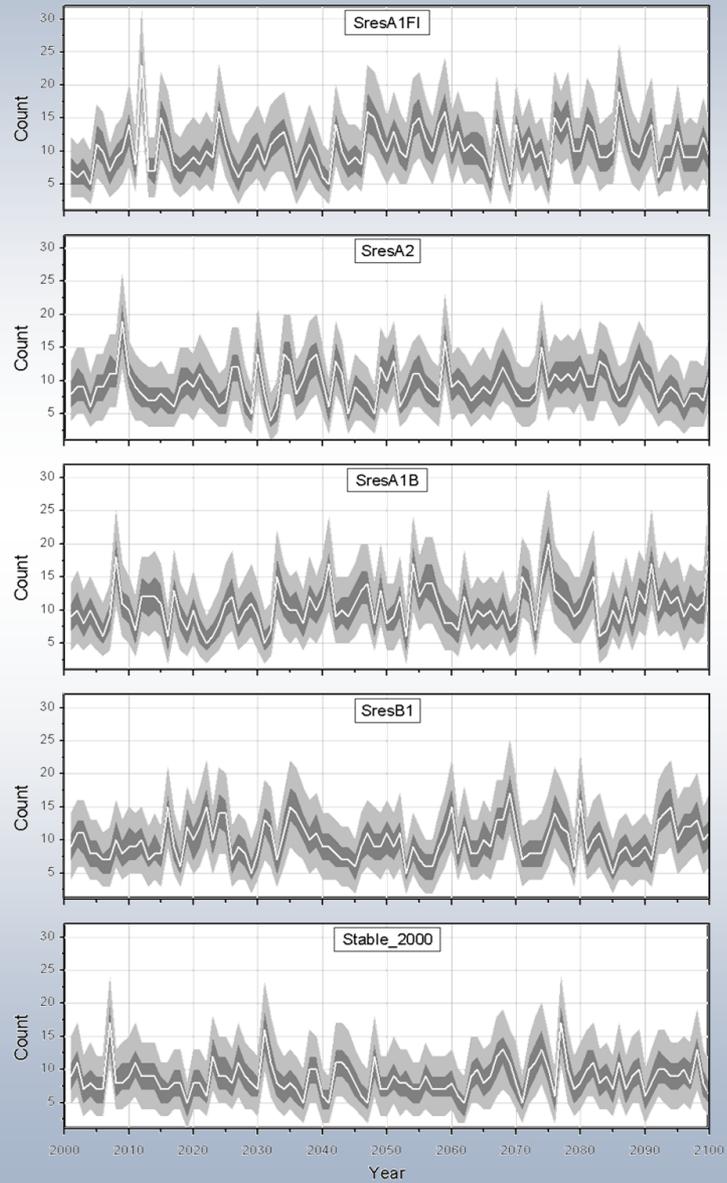
Statistical modeling of the tropical storm count for the North Atlantic basin using a Poisson regression model and the following covariates:

- Southern Oscillation Index
- North Atlantic Oscillation
- Tropical Atlantic SST
- Tropical mean SST

North Atlantic TS Activity (SresA1B)



North Atlantic TS Activity (5 scenarios)



Conclusions

- **Change-points rather than monotonic trends are responsible for the violation of the stationarity assumption.**
- **Analyses of historical discharge records do not point at changes in the flood peak distribution due climate change.**
- **Non-stationary modeling of flood peaks in urban environment**
- **Under nonstationary conditions, alternative definitions of return period have to be sought.**
- **Spatial heterogeneity in “mixtures” of flood peaks.**
- **Tropical cyclones control extreme flooding in much of the eastern US. Analyses of tropical storm frequency in a warmer climate do not suggest a significant increase in tropical storms in the North Atlantic basin over the 21st century.**

The background of the slide is a composite image. The top half shows a bright blue sky with large, fluffy white clouds. The bottom half shows a close-up of a surface covered in numerous water droplets of various sizes, creating a textured, shimmering effect. The text "Thank You!" is centered in the upper half, and "The End" is centered in the lower half, separated by a horizontal line.

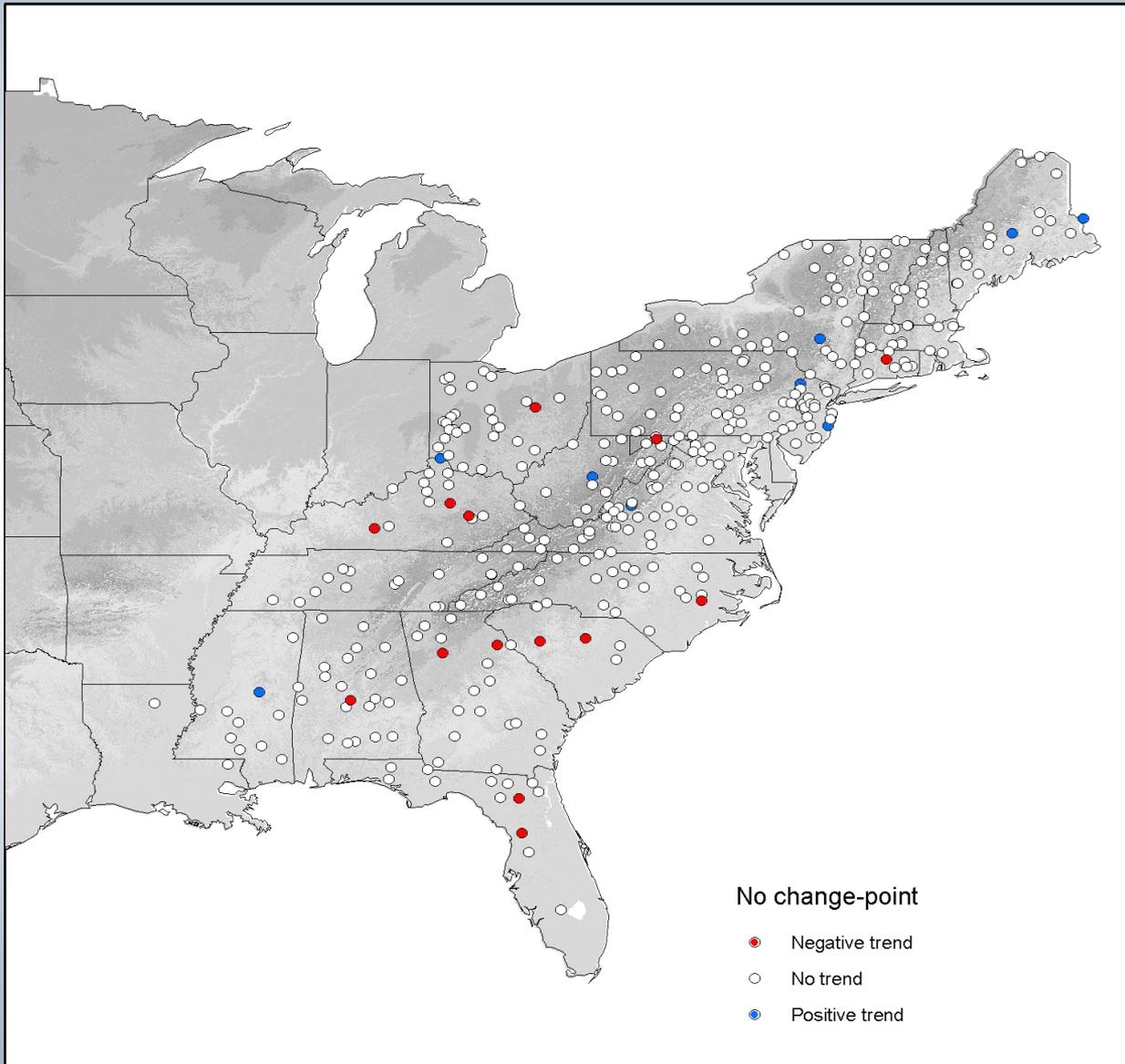
Thank You!

The End

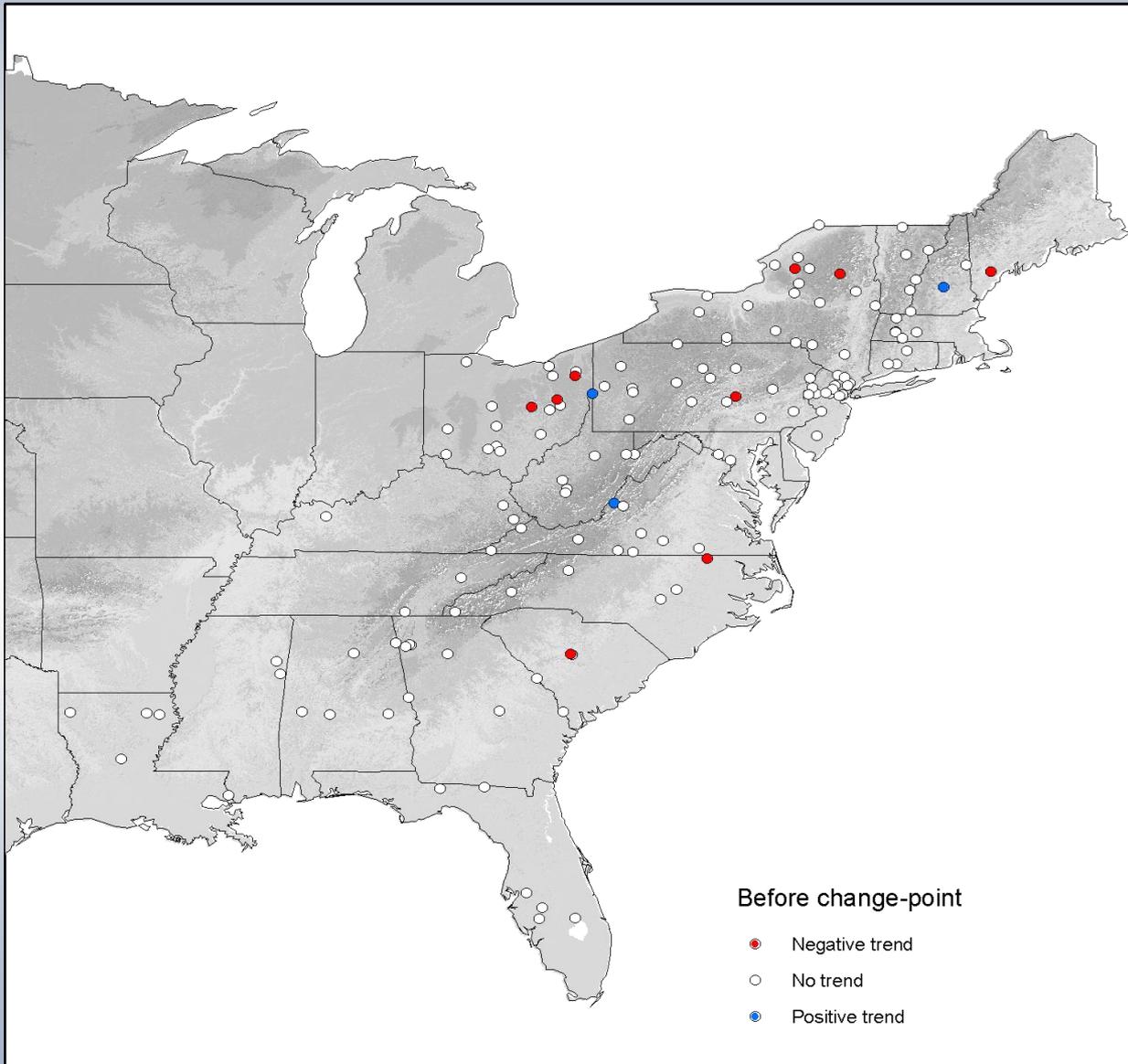
The image features a blue sky with large, white, fluffy clouds in the upper half. The lower half shows a close-up of a textured surface covered in numerous water droplets of various sizes, creating a shimmering effect. The text 'Extra slides' is centered in the middle of the image, with a dark blue horizontal line below it.

Extra slides

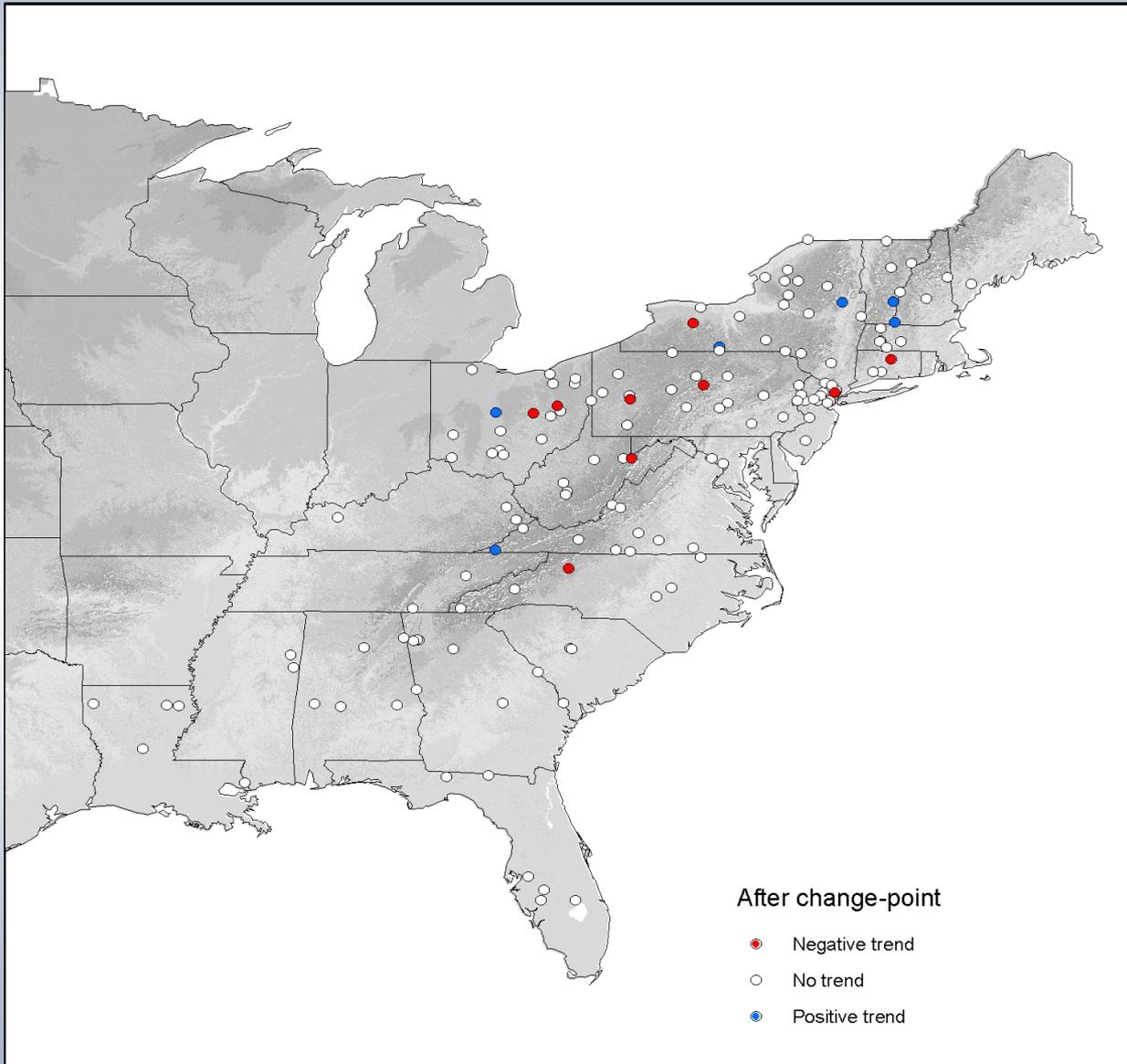
Trend Analyses: no change-point (Eastern US)



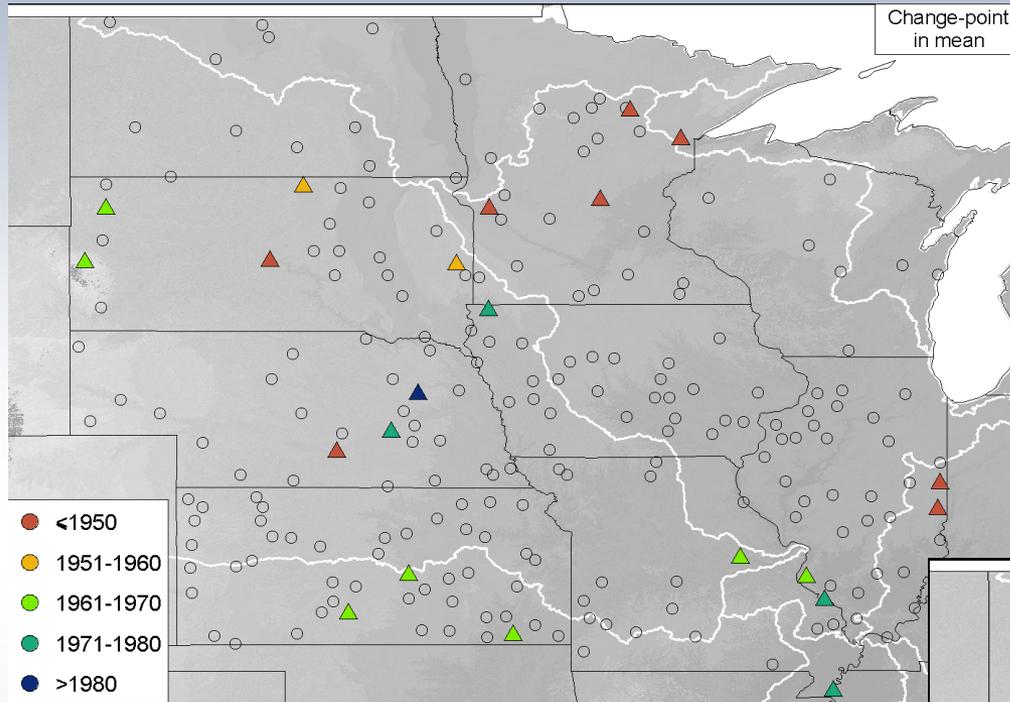
Trend Analyses: before cp (Eastern US)



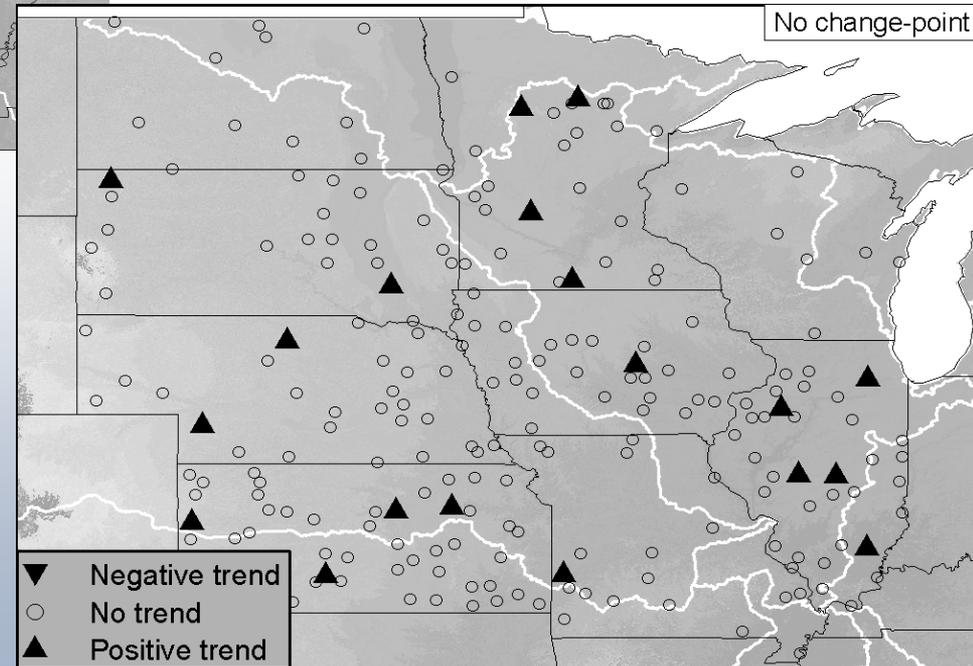
Trend Analyses: after cp (Eastern US)



Extreme Rainfall: Central US



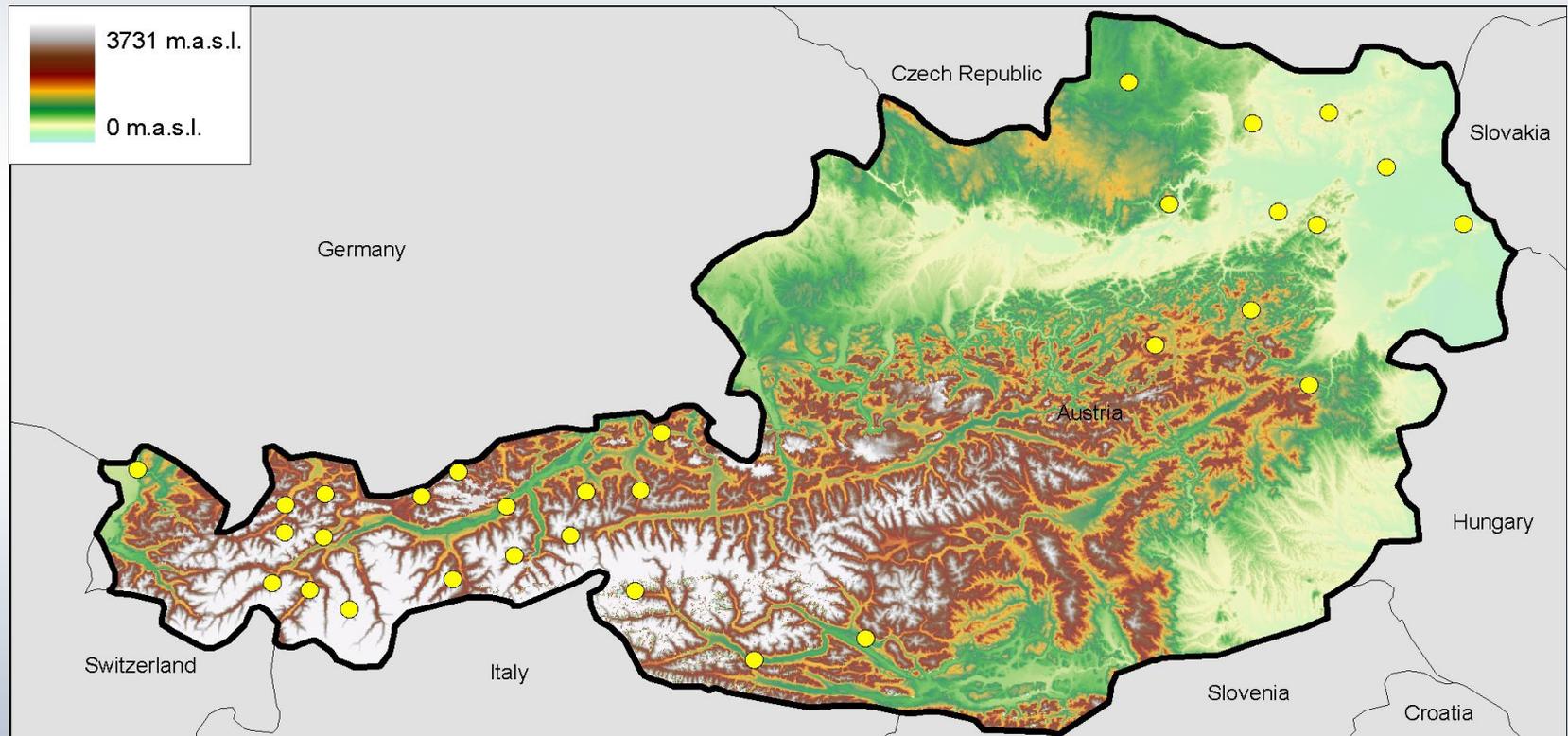
Analyses of 221 rain gages with at least 75 years of daily rainfall.



Villarini, G., J.A. Smith, M.L. Baeck, R. Vitolo, D.B. Stephenson, and W.F. Krajewski On the frequency of heavy rainfall for the Midwest of the United States, *Journal of Hydrology*, 2010 (submitted).

Extreme Rainfall: Austria

Analyses of 31 rain gages with at least 50 years of daily rainfall.

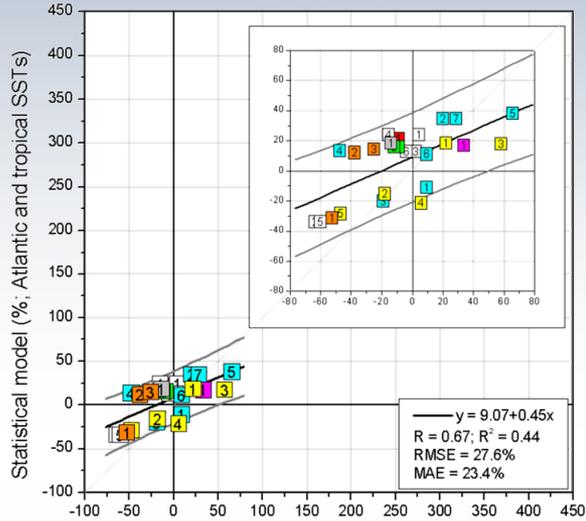


Villarini, G., J.A. Smith, A.A. Ntelekos, and U. Schwarz, Annual maximum and peaks-over-threshold analyses of daily rainfall accumulations for Austria, *Journal of Geophysical Research*, 2010 (submitted).

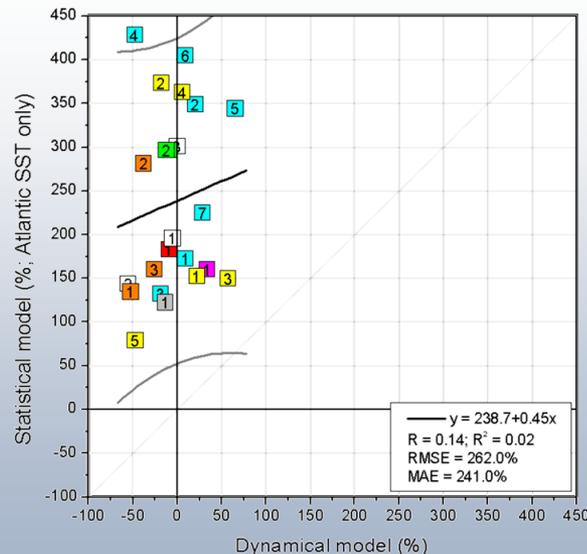
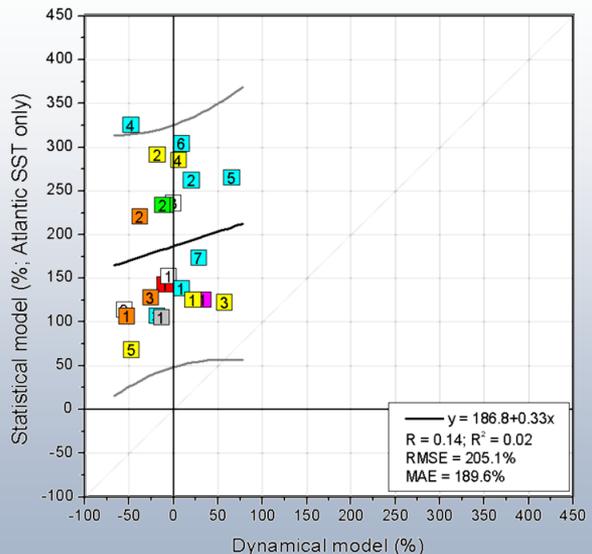
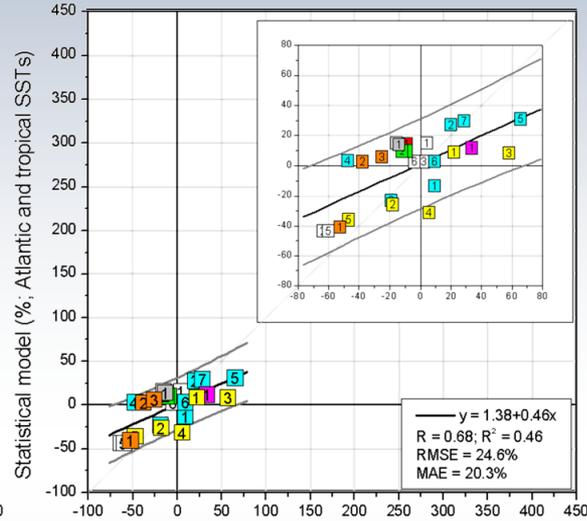
Tropical Storms: Climate Projections

Percentage Change in Tropical Storm Frequency

ERSSTv3b



HadISSTv1



- | | |
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| <ul style="list-style-type: none"> ■ Knutson et al. (2008) 1. GFDL CM2.1 Zhao et al. (2009) 1. GFDL CM2.1 (Reynolds) 2. UKMet HadCM3 (Reynolds) 3. Max Planck-ECHAM5 (Reynolds) 4. GFDL CM2.1 (HadISST) 5. UKMet HadCM3 (HadISST) 6. Max Planck-ECHAM5 (HadISST) ■ Emanuel et al. (2008) 1. NCAR CCSM3.0 2. CNRM-CM3 3. CSIRO-Mk3.0 4. Max Planck-ECHAM5 5. GFDL CM2.0 6. MIROC-Medres 7. MRI CGCM2.3.2 | <ul style="list-style-type: none"> ■ Bengtsson et al. (2007) 1. CNRM with T213 AGCM 2. CNRM with T959 AGCM ■ Oouchi et al. (2005) 1. MRI CGCM2.3 Gualdi et al. (2009) 1. INGV-ECHAM4 2xCO₂ ■ Bender et al. (2010) 1. UKMet HadCM3 2. Max Planck-ECHAM5 3. MRI CGCM2.3.2 ■ Sugi et al. (2009) 1. MRI CGCM2.3.2 (20 km) 2. MIROC-Hires (20 km) 3. MRI CGCM2.3.2 (60 km) 4. MIROC-Hires (60 km) 5. CSIRO-Mk3.0 (60 km) |
|--|---|

Villarini, G., G.A. Vecchi, T.R. Knutson, M. Zhao, and J.A. Smith, North Atlantic tropical storm frequency response to anthropogenic forcing: Projections and sources of uncertainty, *Journal of Climate*, 2010 (submitted)