

Climate Variability and Uncertainty in Flood Risk Management in Colorado: An interdisciplinary project on extremes Rebecca Morss, Doug Nychka Mary Downton, Olga Wilhelmi, Uli Schneider, Eve Gruntfest, Heidi Cullen, and others

National Center for Atmospheric Research



### Example: Fort Collins (CO), July 1997

- Over 10" of rain in 6 hours
- 5 dead, 54 injured, 200 homes destroyed





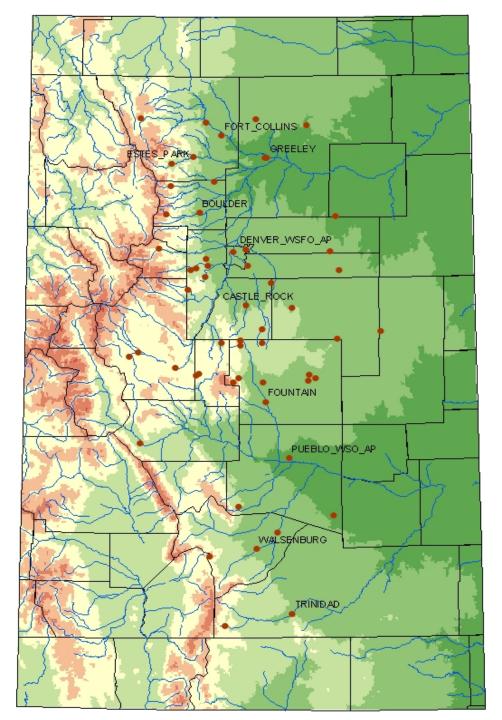
• Much of damage outside 500-year floodplain

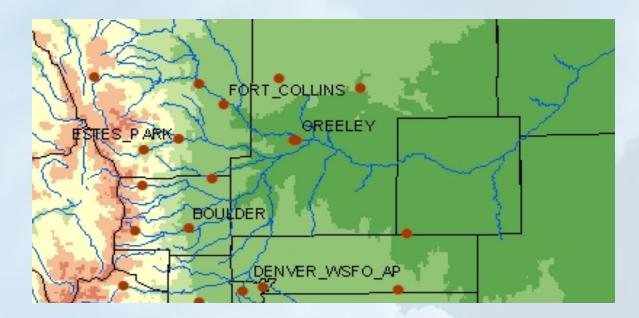
# Key Questions

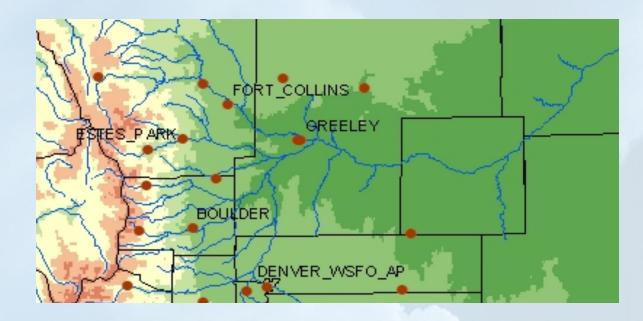


- How is scientific information about extreme flooding used in flood management decision-making?
- How does uncertainty in risk of extreme flooding interact with flood management decision-making?
- What new or improved scientific information about
  extreme floods could we provide that would benefit
  flood management?

- Focus: Colorado Front Range
- Steep topography
- Spatially and temporally variable precipitation
- Limited data
- Growing population
- ⇒ Flood management important, but flood risk uncertain



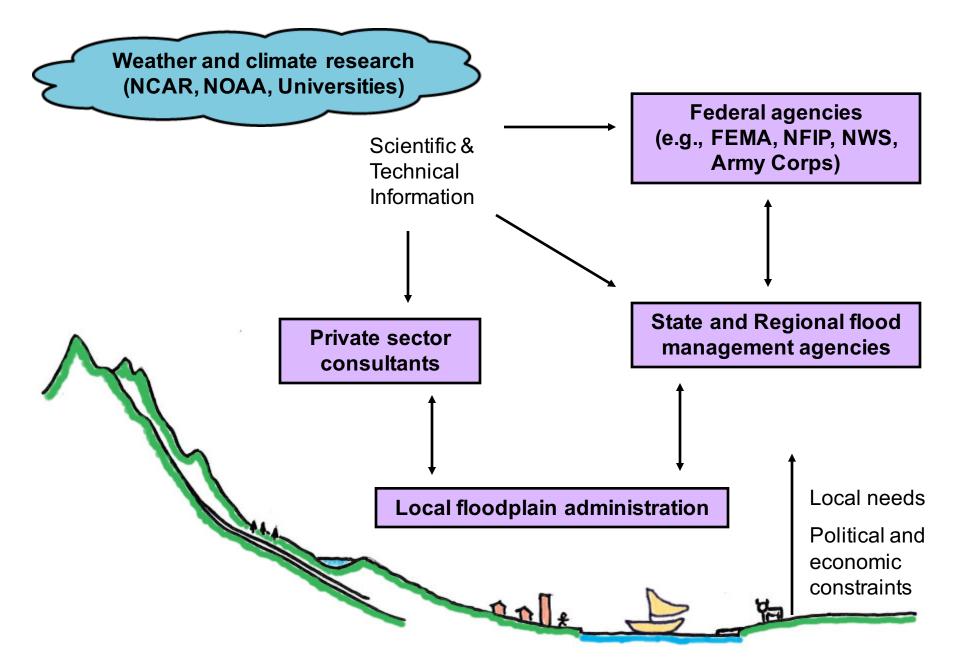


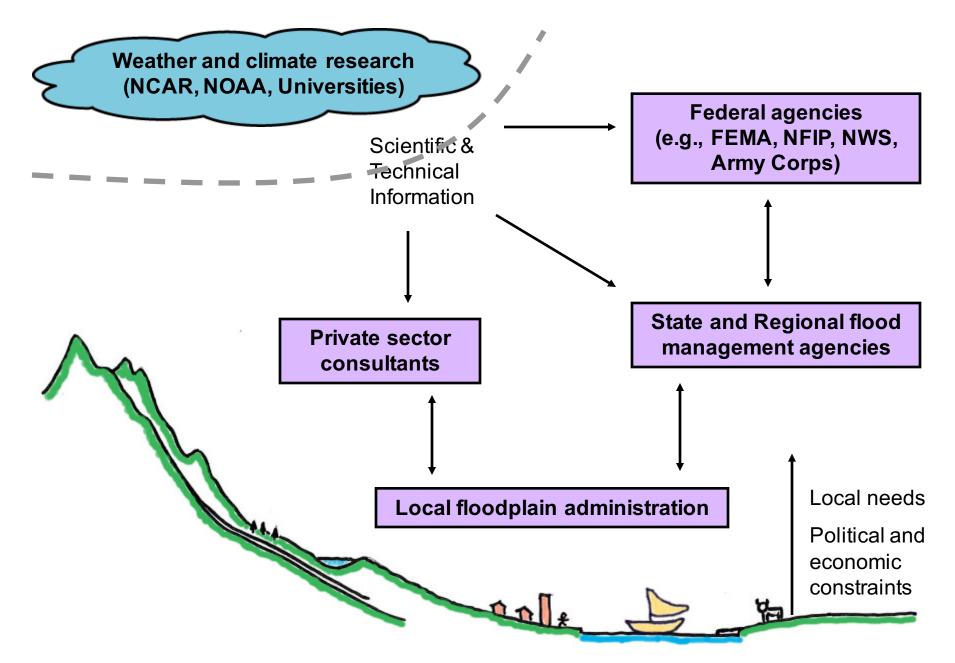


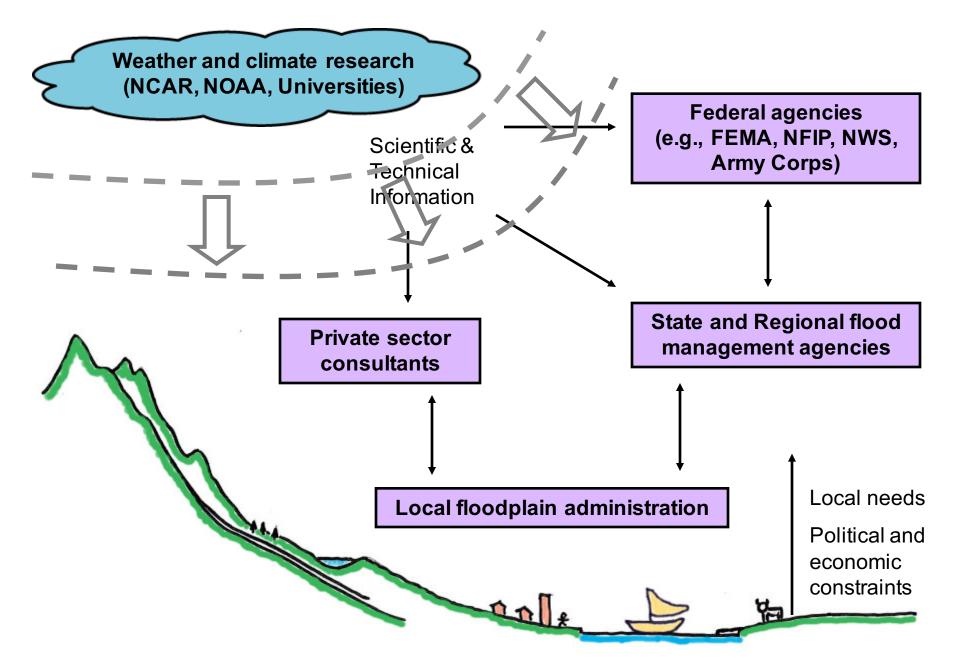
- Meteorology/climatology
- Hydrology

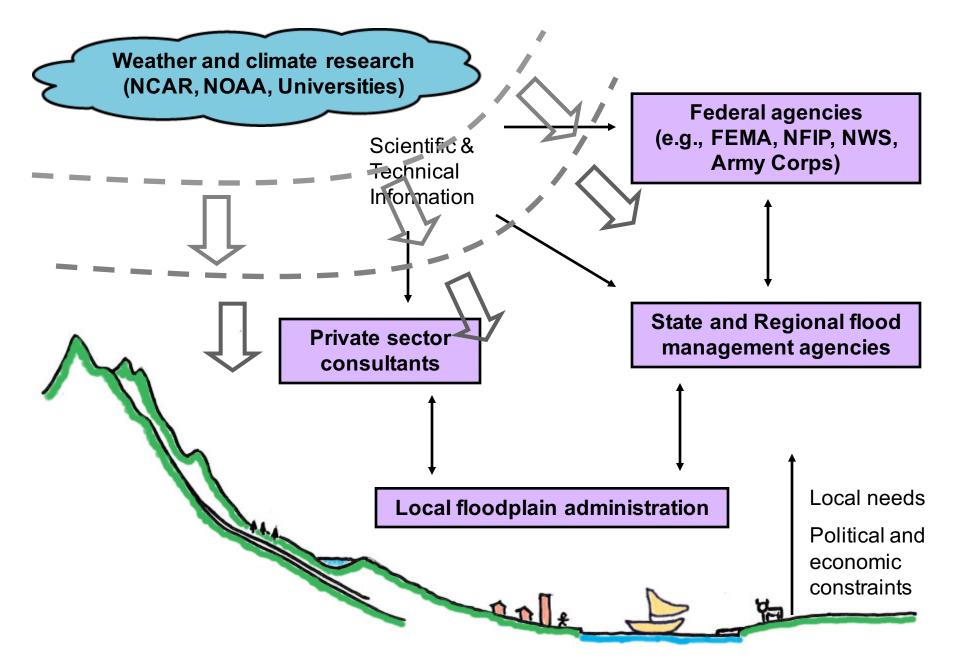
• Statistics

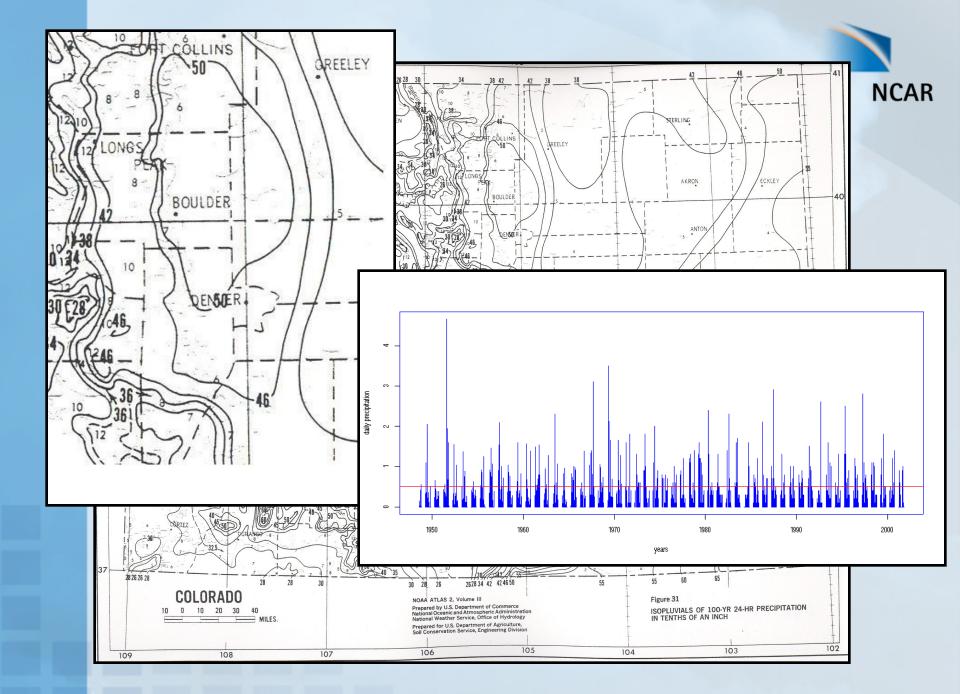
- Engineering
- Decision-making





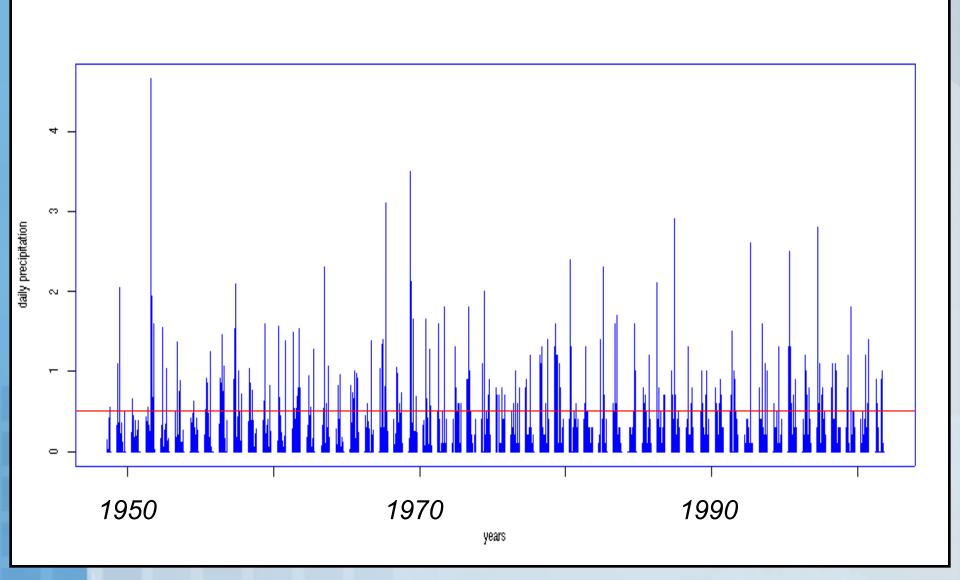






### **Boulder Daily Precipitation**

NCAR

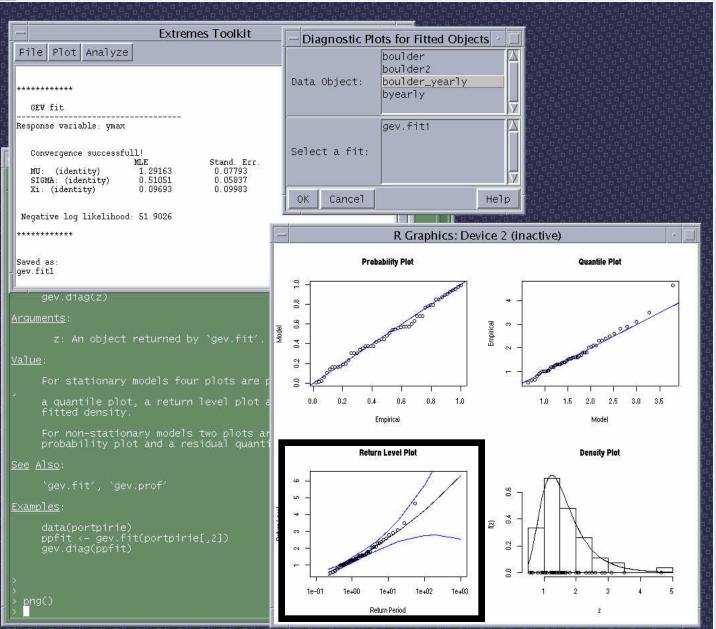


Extremes Analysis of Boulder Daily Precipitation



- *Generalized extreme value distribution* can model the annual maxima
- *Generalized Pareto distribution* is an equivalent model for daily data
- Both are characterized by scale, shape and location parameters
- A common summary is the *return time*: *e.g.*, the 100 year return level has probability .01 of occurring in a given year

### Analysis Using Extremes Toolkit



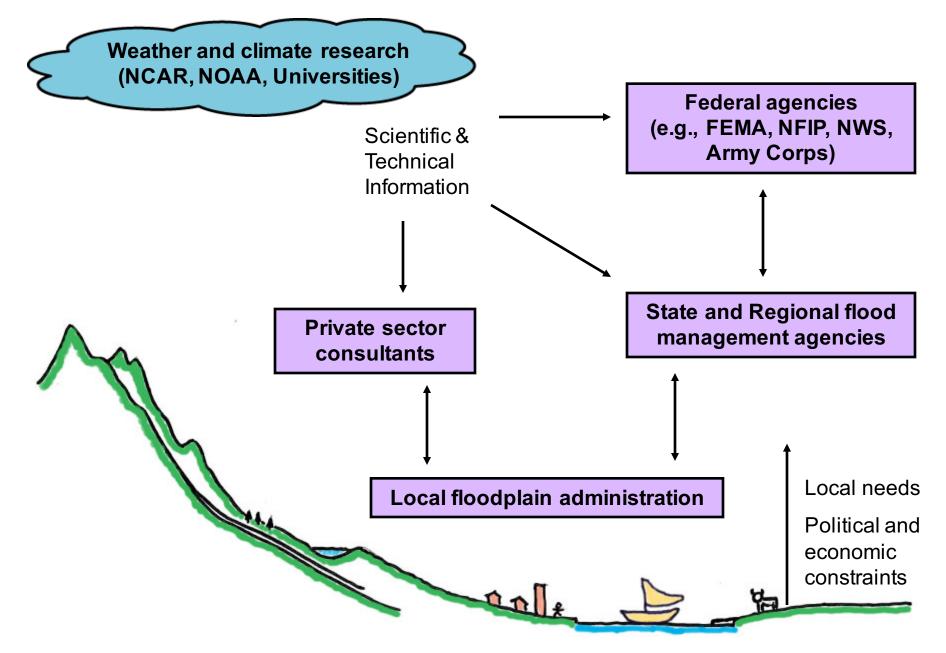
NCAR

### 100 Year Return Level for Boulder NCAR 15 무 Return Level S 0 0 0.1 100 10 1000 Return Period



# Combining Geostatistics and Extremes

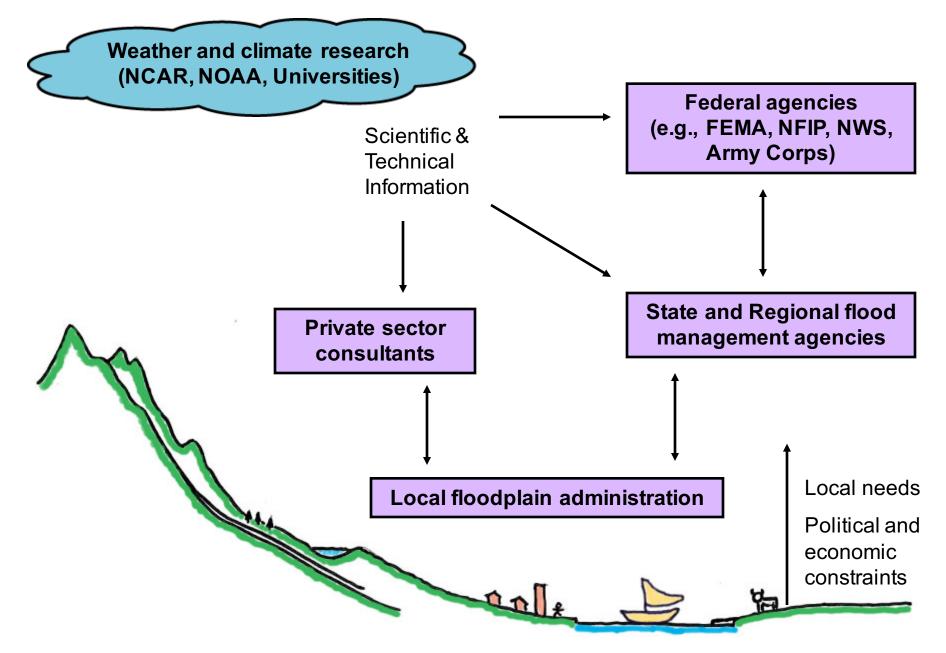
- Use (Bayesian) spatial statistical models to extrapolate the extremal analysis to unobserved locations.
- Incorporate covariate information such as elevation, aspect.
- Use ensembles of maps to quantify uncertainty.

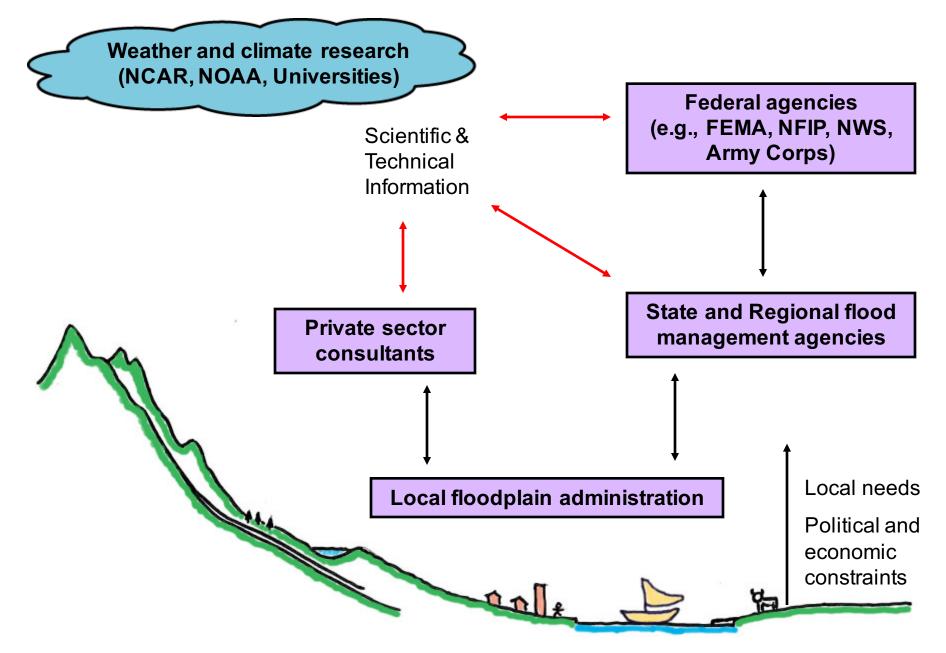


Scientific information, uncertainty, and decision-making: Lessons learned



- Decision-makers are diverse and interconnected
- Political, technical, and contextual constraints often limit decision-makers' ability and motivation to use new methods and information
- Scientists and decision-makers often perceive of and respond to uncertainty differently
- Assumptions often act as barriers to more useable scientific information





### Summary



- Important to understand, and place information in, decision context
  - For example, quantification of uncertainty is increasingly important in risk-based analysis
- Important to approach producing scientific information from decision-maker point of view, not vice versa
  - Put decision-making at center, iterate between specific flood management practitioners and scientists to co-produce useful information

## Key Questions & Answers



• How is scientific information about extreme flooding used in flood management decision-making?

In a more limited fashion than scientists might initially expect, given potential relevance of information

• How does uncertainty in risk of extreme flooding interact with flood management decision-making?

In complex ways, depends on the specific decision-maker

• What new or improved scientific information about extreme floods could we provide to benefit flood management?

New statistical approaches for precipitation extremes atlas, as a start

## Future Work



- Continue developing statistical methods for an improved extreme precipitation atlas in collaboration with users of the information
- Integrate specific features of flood risk management into more general models of uncertainty and decisionmaking
- Apply/extend methods developed and lessons learned to other projects on extremes, uncertainty, and decision-making