# Hydrologic Impacts of Atmospheric Rivers in the Western U.S.

#### **Christine M. Albano**

Graduate Program in Hydrologic Sciences University of Nevada, Reno and Desert Research Institute

#### Michael D. Dettinger

U. S. Geological Survey

**Adrian Harpold** 

University of Nevada, Reno







## AR Impacts Range from Hazardous to Beneficial



- Cat 5 Primarily hazardous
- Cat 4 Mostly hazardous, also beneficial
- Cat 3 Balance of beneficial and hazardous
- Cat 2 Mostly beneficial, also hazardous
- Cat 1 Primarily beneficial
- Ralph et al., In review



## Objective: Characterize the magnitude, nature, and variability of hydrologic impacts based on the AR scale

**Over Space** 



- Atmospheric conditions and terrain affect precipitation amounts
- Site differences affect hydrology

**Over Time** 

#### MENDOCINO MOUNTAINS



- Atmospheric conditions affect precipitation amounts
- Antecedent conditions affect hydrology

## Approach: Assess probable (50<sup>th</sup> percentile) and potential extreme (90<sup>th</sup> percentile) hydrologic impacts

**Over Space** 

- Composite 50<sup>th</sup>/90<sup>th</sup> percentile hydrologic responses by AR CAT
  - Runoff
  - $\Delta$  Soil Moisture
  - $\Delta$  Snow Water Equivalent



Over Time

 Quantile regression of 50<sup>th</sup>/90<sup>th</sup> percentile hydrologic responses to AR conditions from 1980-

2013, by pixel



Storm Total IVT

### Data and Methods – Atmospheric River Event Chronology

- Rutz MERRA-2
- Winter (Oct-Apr) ARs, 1980-2013
- 3 hr  $\rightarrow$  Daily  $\rightarrow$  Event Scale

### **Composite Analysis**

• Storm Total IVT Classes

STIVT 10 <sup>7</sup> (kg/m)	AR CAT
>19.44	Cat 5 – Primarily hazardous
>12.96	Cat 4 – Mostly hazardous, also beneficial
>8.64	Cat 3 – Balance of beneficial and hazardous
>4.32	Cat 2 – Mostly beneficial, also hazardous
>2.16	Cat 1 – Primarily beneficial

### Multiple Quantile Regression

- n > 30 AR events per pixel
- Predictor Variables:
  - Storm Total IVT
  - Specific humidity-weighted
    - temperature
    - wind direction
  - Antecedent soil moisture (VIC model)

### Data and Methods – Hydrologic Responses

Variable Infiltration Capacity (VIC) Model

- 1/16<sup>th</sup> degree, Livneh et al. 2015
  - Antecedent Soil Moisture (Predictor)
  - Response Variables
    - Precipitation (model forcing)
    - Surface Runoff +  $\Delta$  Baseflow
    - $\Delta$  Soil Moisture
    - $\Delta$  Snow Water Equivalent

#### Variable Infiltration Capacity (VIC) Macroscale Hydrologic Model



## Number of AR Events by Category 1980-2013



## Median Precipitation Impact by AR Category







RUNOFF



 $\Delta$  SOIL H<sub>2</sub>O

PRECIP



NTILE 90<sup>th</sup> PERCENTILE





## **Quantile Regression Model Fits**

**Precipitation** =STIVT + WIND DIRECTION

 $\Delta Soil Moisture$ = STIVT + WIND+TEMP+ANTECEDENT SOIL

**Runoff** =STIVT + WIND+TEMP+ ANTECEDENT SOIL

 $\Delta$  **SWE** =STIVT + WIND+TEMP



0.9 0.8 0.7 0.6 -0.5 -0.4 -0.3 -0.2 -0.1 0

 $\mathsf{R}^1$ 

90<sup>th</sup>

Runoff

Storm Total IVT







## Runoff Response to Antecedent Conditions and IVT Varies among Sites



## Summary

- Similar AR conditions (STIVT) overhead result in different impacts over space
  - Larger response in coastal and mountainous terrain
  - Magnitude and nature of hydrologic response differs among sites
- Wide variation in impacts over time at a given location and STIVT
  - Upper limits have closer connection and are more sensitive to STIVT than central tendencies
  - STIVT has almost no association with precipitation in some places
  - Role of other drivers varies among locations





Storm Total IVT

## Implications

- Results provide a spatially explicit interpretation of the AR scale that can be used for communicating potential impacts
- Quantile regression may be a useful tool for linking maximum potential hydrologic impacts to AR characteristics
  - Response 'ceiling' varies linearly with STIVT in many locations