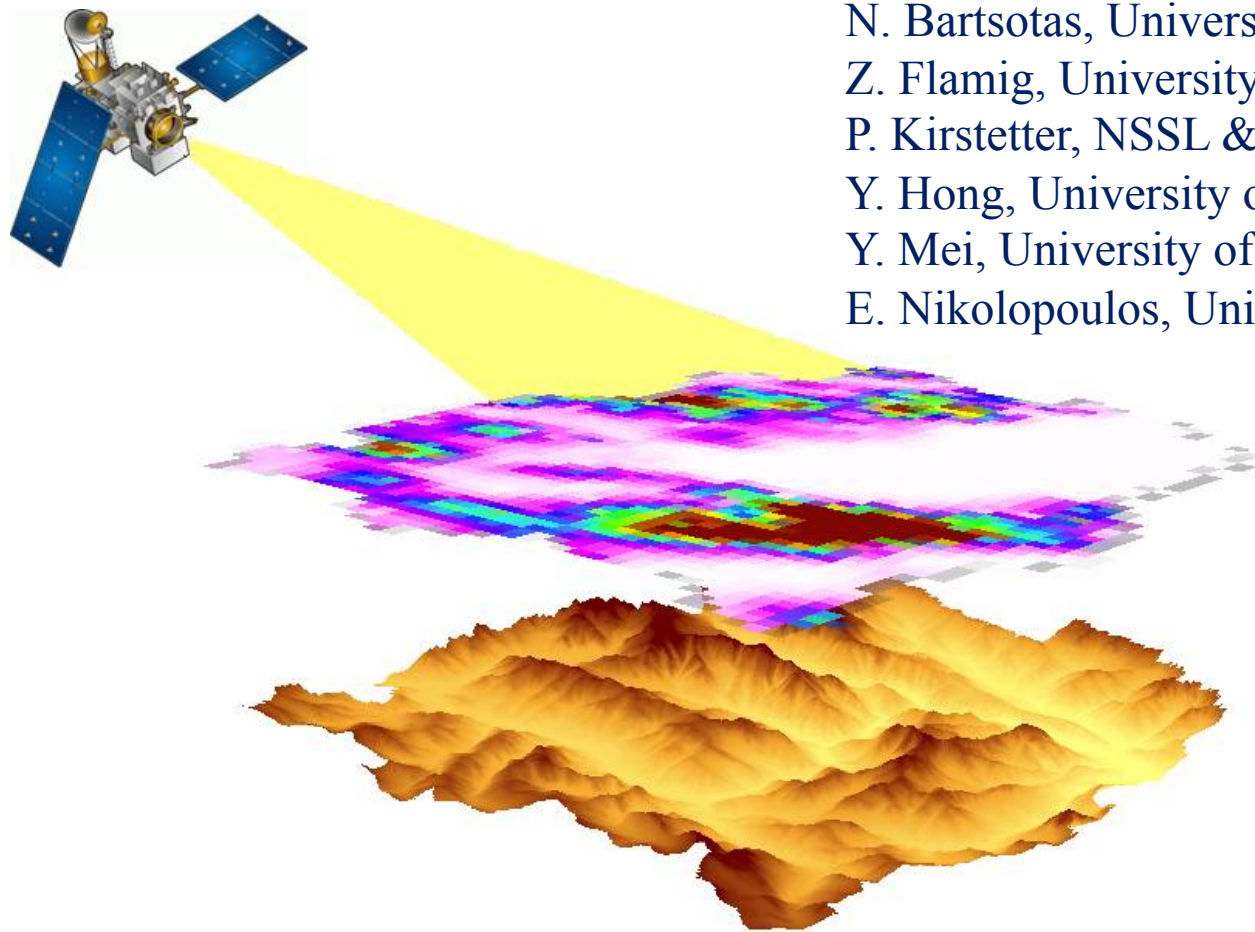


# Use of GV data to evaluate and improve uses of satellite-rainfall in hydrologic modeling of complex terrain floods

*E.N. Anagnostou, University of Connecticut and J.J. Gourley, NOAA/NSSL*



## With contributions from:

N. Bartsotas, University of Athens

Z. Flamig, University of Oklahoma

P. Kirstetter, NSSL & University of Oklahoma

Y. Hong, University of Oklahoma

Y. Mei, University of Connecticut

E. Nikolopoulos, University of Padova



P E R I A P E L O I E E E

A + I V

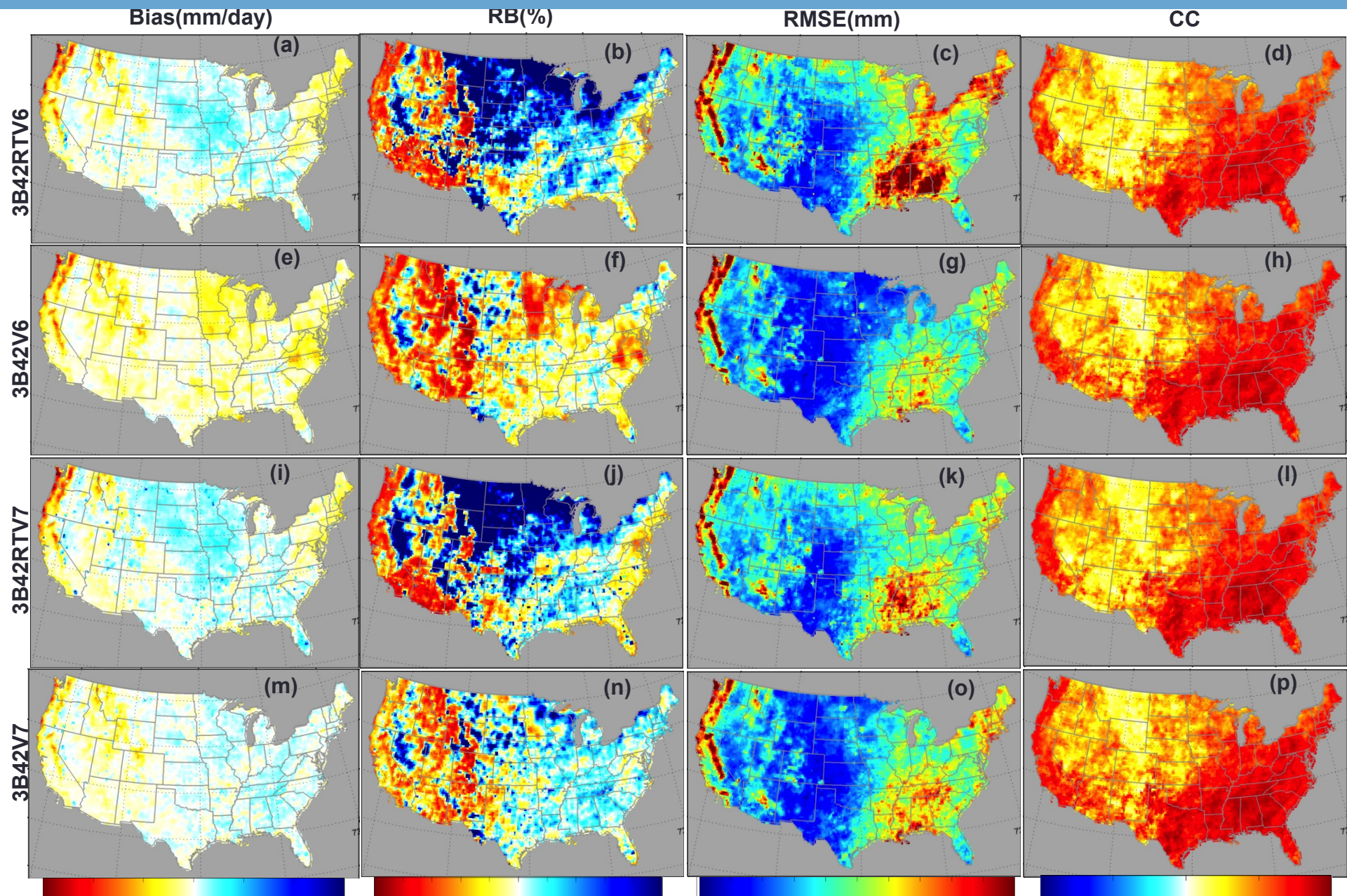
T P I D E B

T C O S

H E P O S I T I E L



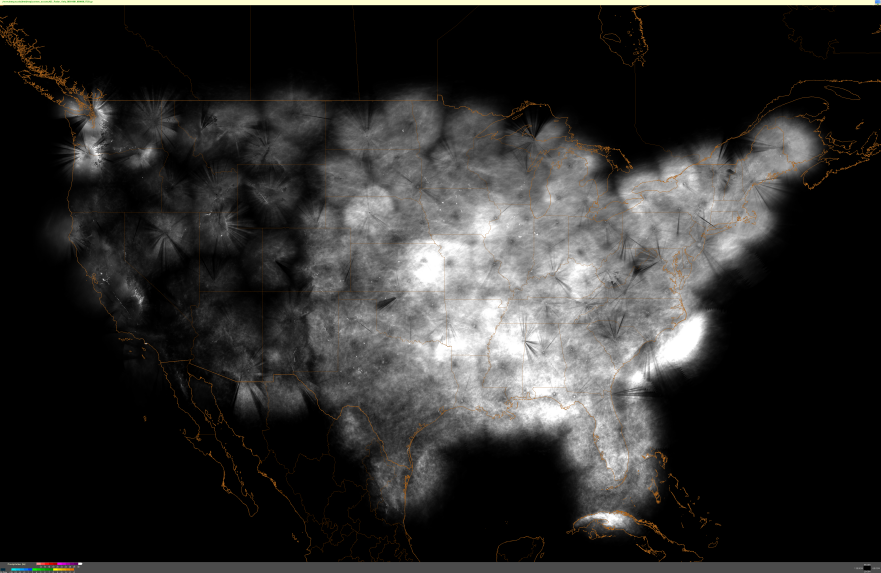
# 3-YEAR EVALUATION OF TRMM V6 AND V7 PRODUCTS OVER US



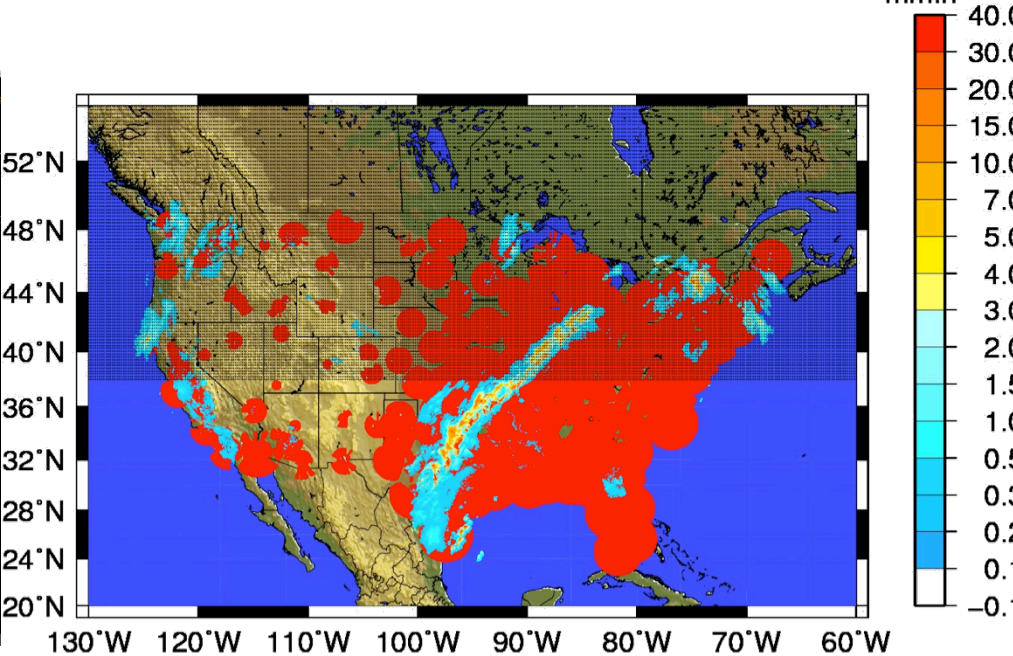


# FRAMEWORK FOR EVALUATING RAINFALL FROM LOW-EARTH ORBITING PLATFORMS

## - National Mosaic and QPE system (NMQ)



Real-time radar products have problems in similar areas



Significant post-processing needed to use ground radar network for evaluating level II rainrate products

- 1. Bias-corrected Q2 (raingauge; hourly bias applied to 5mn; quantitative quality control)
- 2. Best sampling conditions filtered with Radar Quality Index (qualitative quality control)

Kirstetter, P.E., Y. Hong, J. J. Gourley, et al. 2012: Toward a framework for systematic error modeling of spaceborne radar with NOAA/NSSL ground radar-based National Mosaic QPE. *Journal of Hydrometeorology*. doi:10.1175/JHM-D-11-0139.1

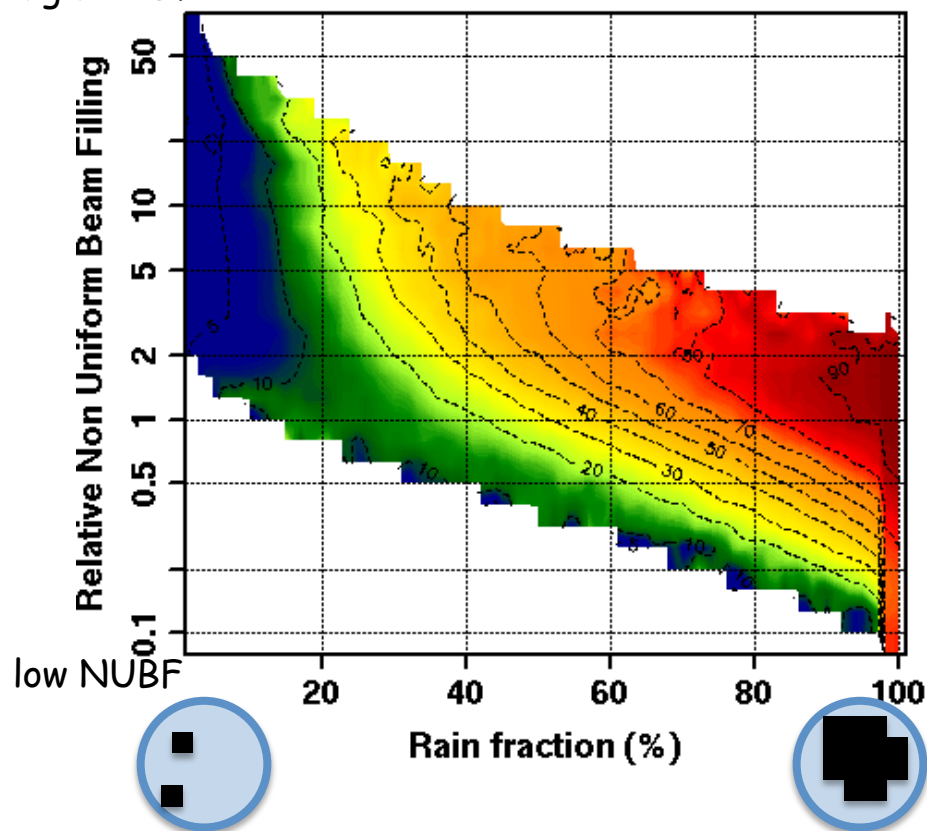


# WHAT HAS BEEN LEARNED FROM TMI/PR EVALUATIONS?

PR/2A25: influence of the NUBF, the rain type and rain rate

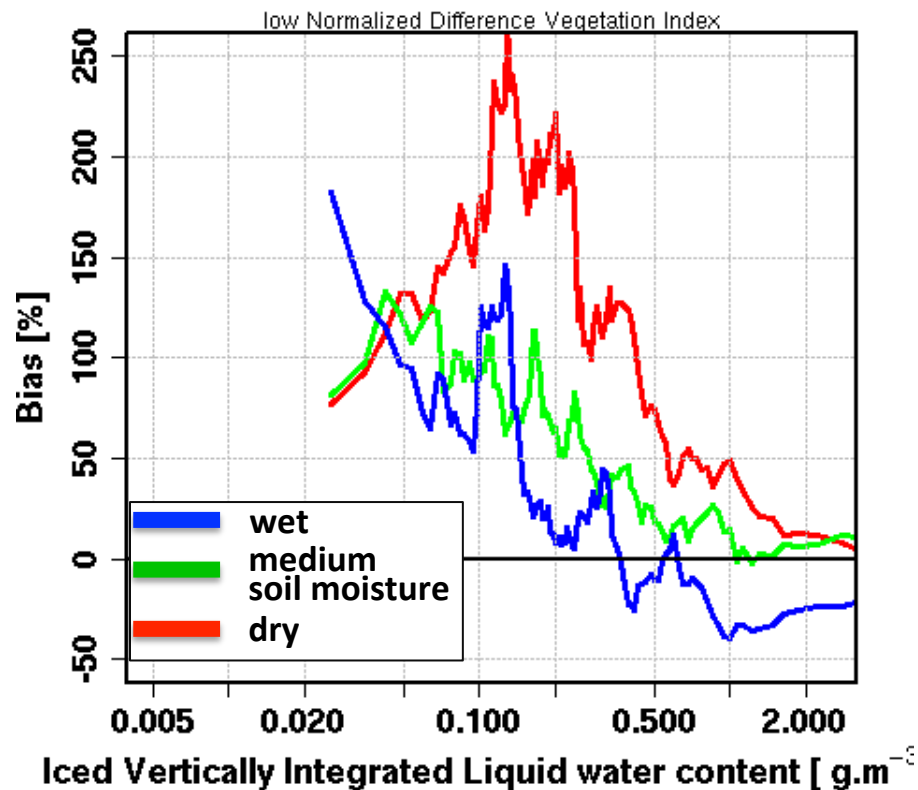
## Probability of Detection

high NUBF



low NUBF

TMI/2A12: influence of the surface (soil moisture), the vegetation and the vertical structure of rainfall (iced VIL)





What are the error characteristics of PMW and combined high-resolution satellite products over complex terrain (mountainous) areas?

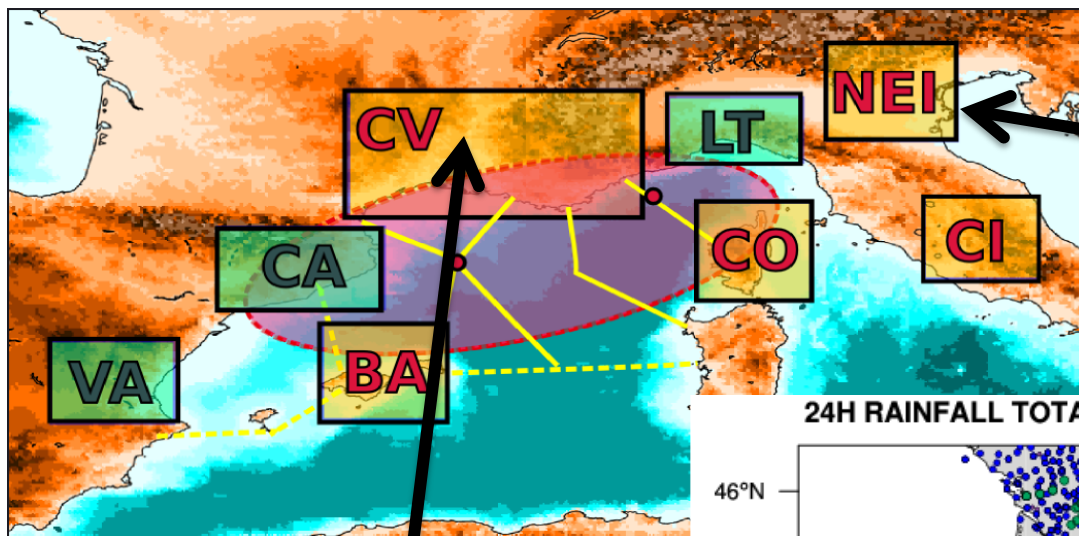
- ✓ *What is the error dependence on rainfall magnitude, vertical structure, or elevation?*
- ✓ *What are the effects of rainfall estimation uncertainty (bias, random error) and product resolution on flood simulation uncertainty; how those error effects vary across basin scales and event magnitude*

What improvements are obtainable in satellite rainfall estimates to improve uses in flood modeling?

- ✓ *How can we achieve those error corrections in absence of ground reference data?*
- ✓ *How can errors in satellite retrievals be accounted for in combination with uncertainties in hydrologic model simulations?*

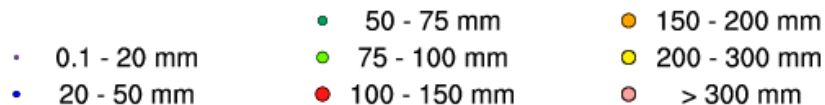
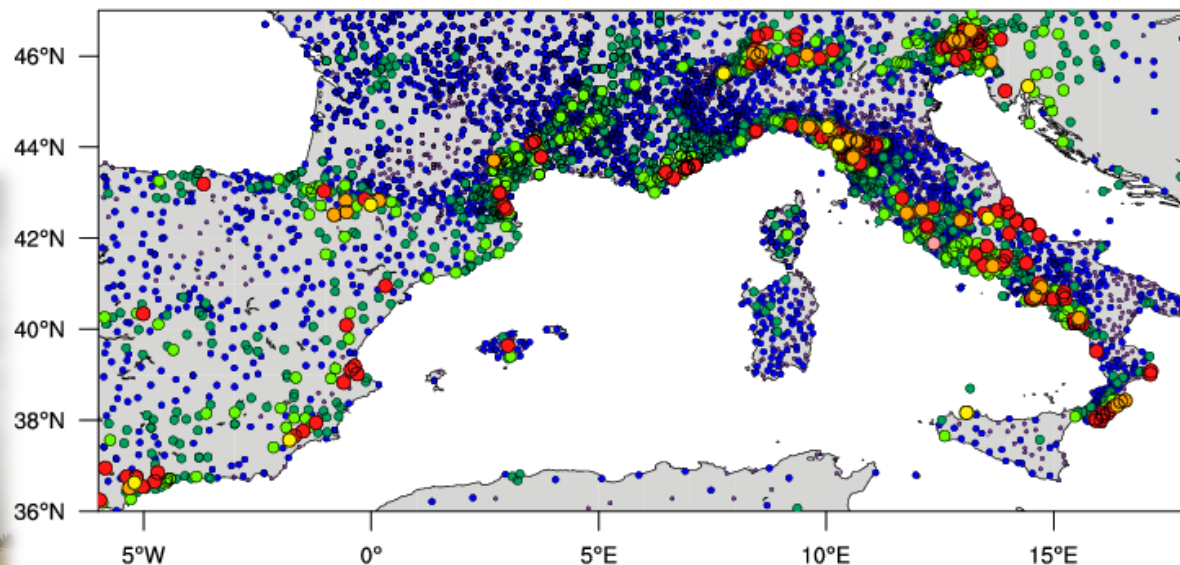


# STUDY AREAS – REPRESENTING HEAVY PRECIPITATION EVENTS (HPEs) AND COMPLEX TERRAIN FLOODS IN HyMEX



XPOL

24H RAINFALL TOTALS (mm) - Maximum at each station over 5 Sep.-6 Nov. 2012



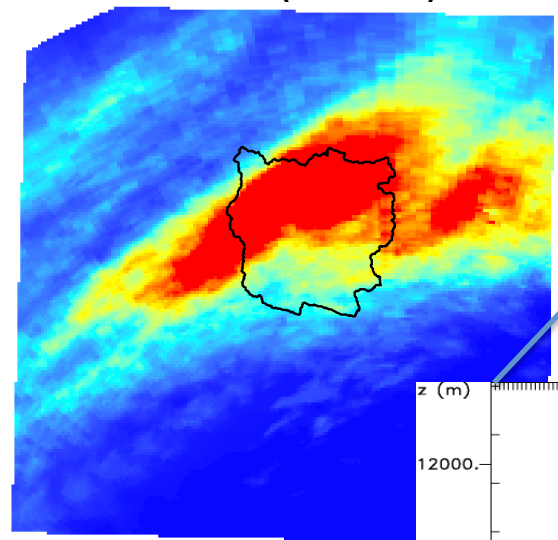
NOXP



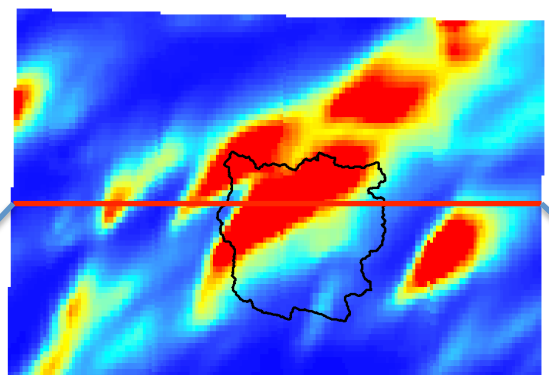
# COMPLEX TERRAIN RAINFALL ERROR ANALYSIS - NEI OBSERVATORY

## Radar and NWP based error analysis of HPE (>500 yr flood) – Fella 2003

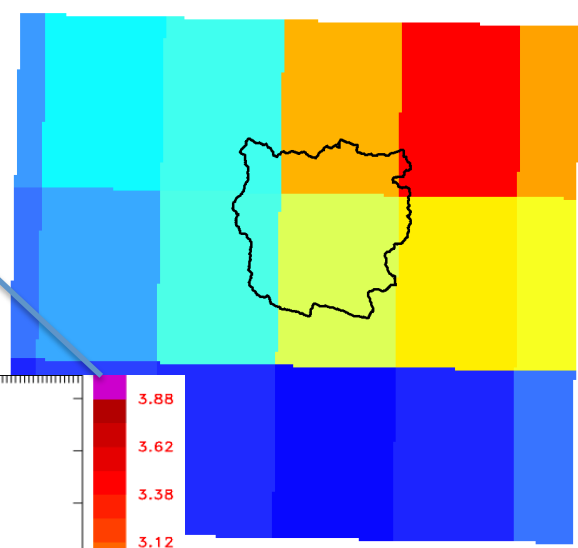
Radar (0.5 km)



NWP\* (1 km)

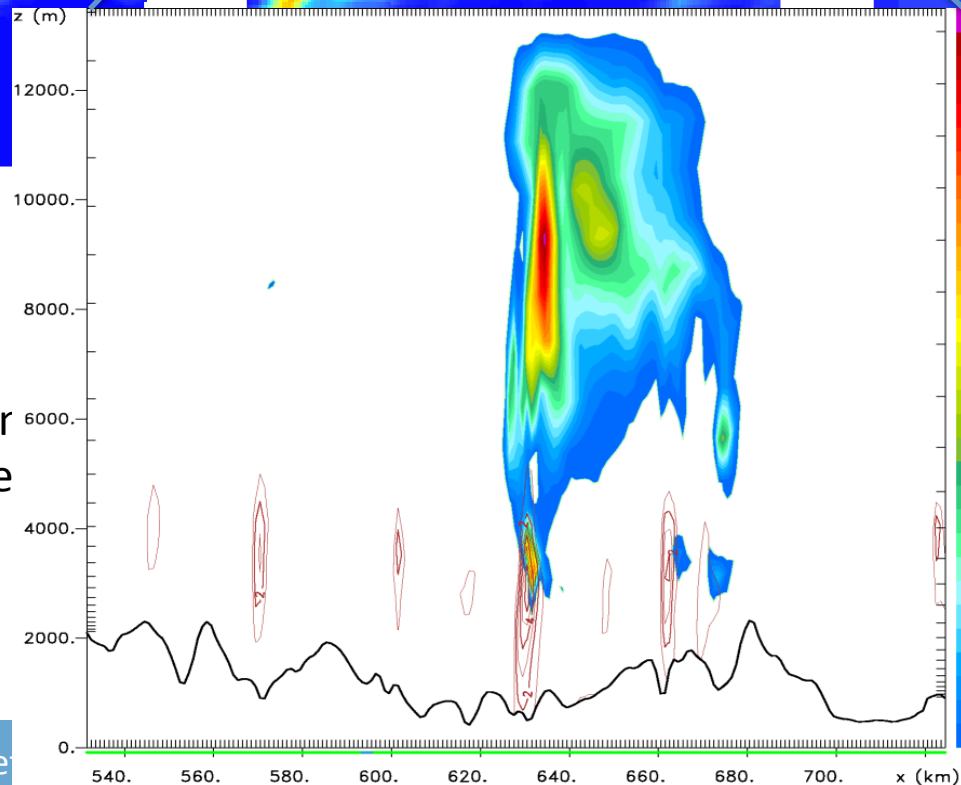


3B42V7 (25 km)



Radar [mm]  
443 mm  
Low : 0.064252

\* High-resolution treatment of pre 11, 873-892)

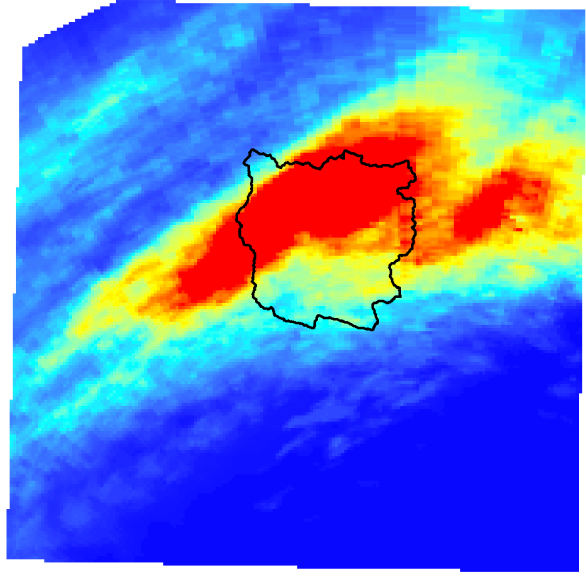


th explicit . Phys.,

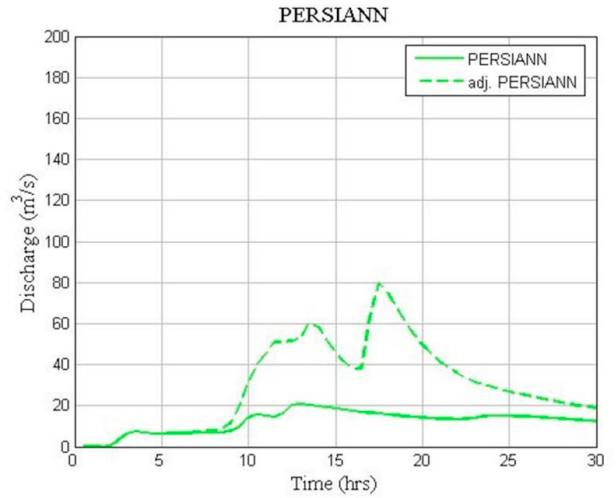
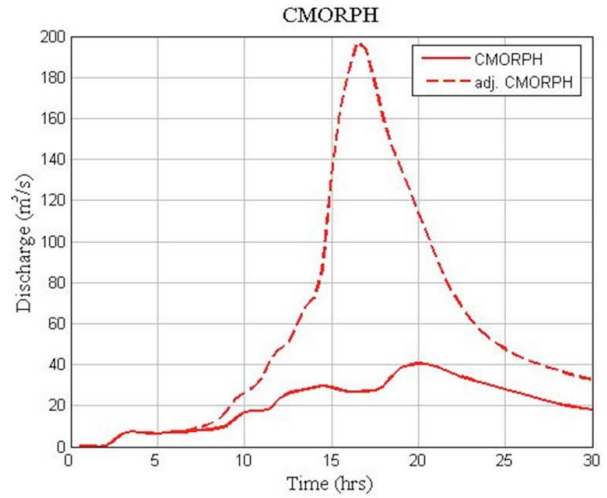
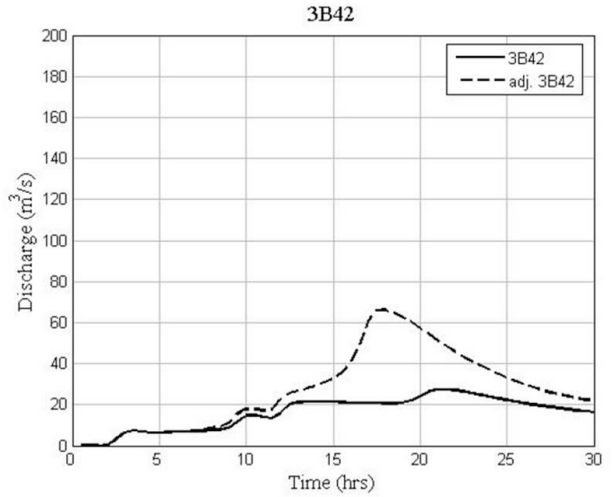
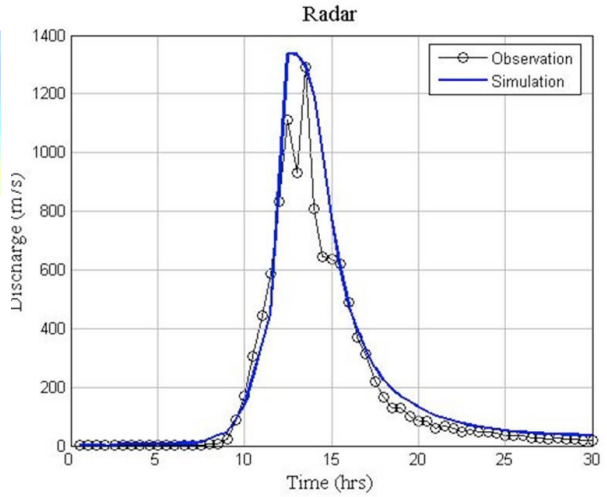


# HYDROLOGIC ERROR PROPAGATION (FELLA 2003)

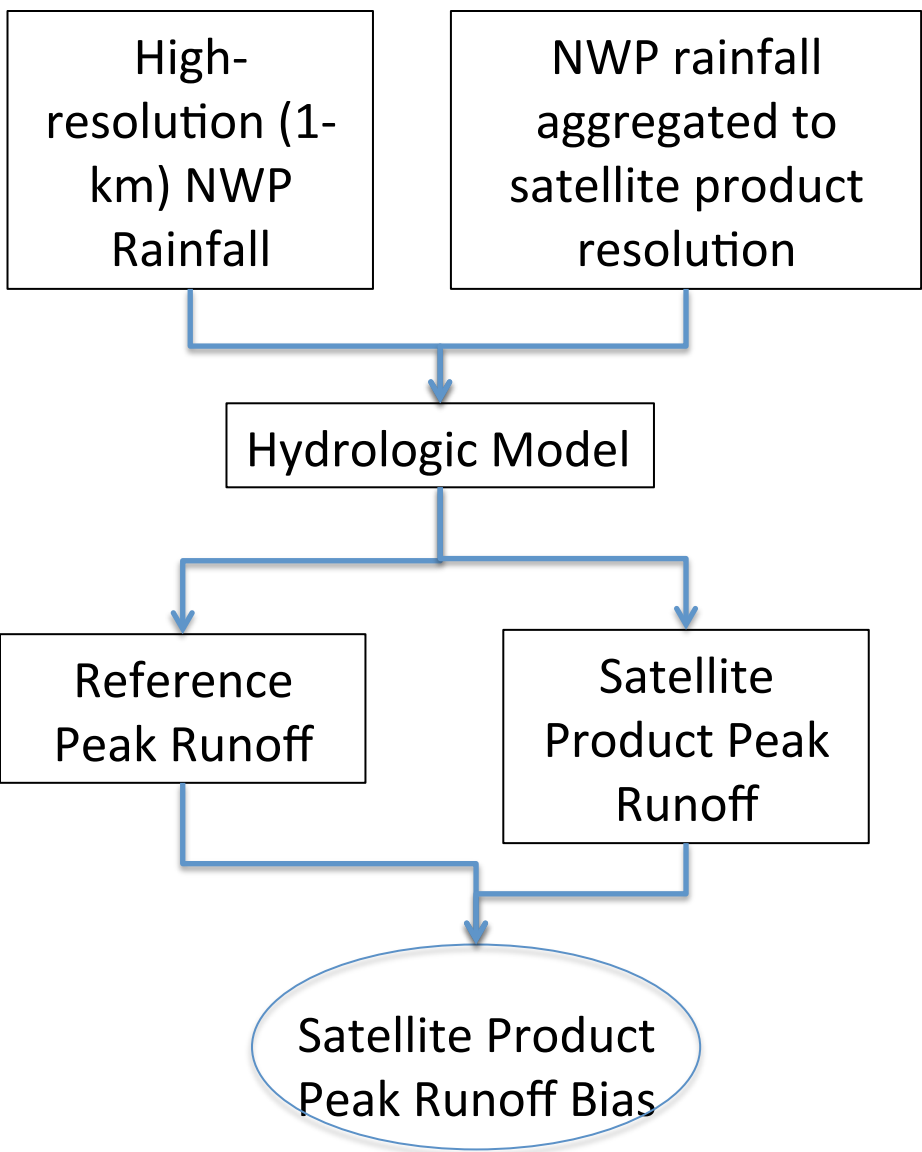
## Fella 2003



Radar [mm]  
 High : 449.83  
 Low : 0.064252



# NWP-BASED ANALYSIS OF RAINFALL-RUNOFF ERROR PROPAGATION



## RESOLUTION EFFECT

	Bias in Peak Runoff	
	Dry cond.	Wet cond.
PERSIANN (4km)	0.88	0.98
CMORPH (8km)	0.73	0.92
3B42 (25km)	0.22	0.75

## SATELLITE RAINFALL BIAS

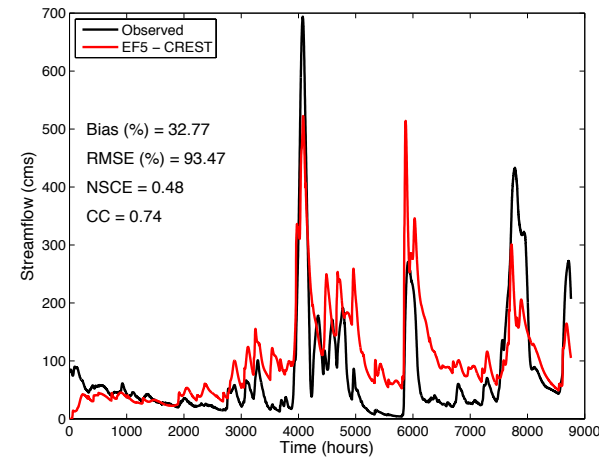
	Original data	NWP adjustment
3B42	0.34	0.70
CMORPH	0.46	0.70
PERSIANN	0.14	0.70



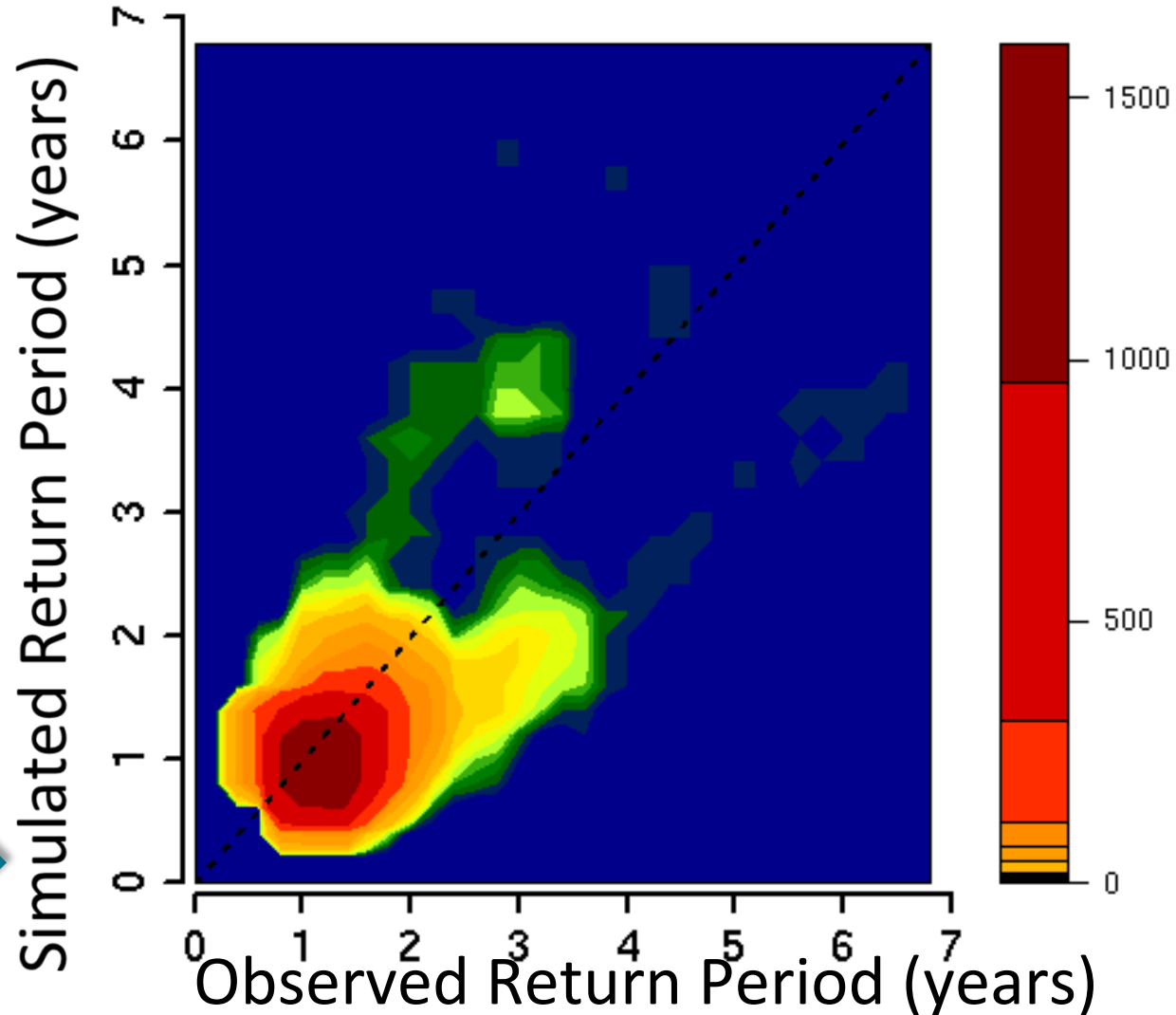
# STEPS FOR DEVELOPING A HYDROLOGIC FRAMEWORK FOR PROBABILISTIC DIAGNOSTICS (AND PROGNOSTICS)

1. Aggregate reference rainfall product up to the space-time resolution of the satellite product
2. Run hydrologic model with a-priori parameters for lengthy period using gridded precip record (~ 1 decade for NEXRAD)
3. Convert time series of streamflow to return period
4. Compare observed & simulated time series and adjust time series by maximizing correlation coefficient (to correct for routing errors)
5. Fit GAMLSS error model
6. Evaluate diagnostic skill of model + precip
7. Provide probabilistic predictions of flooding via Bayes' theorem

- Scatter/Density plots, made from time series



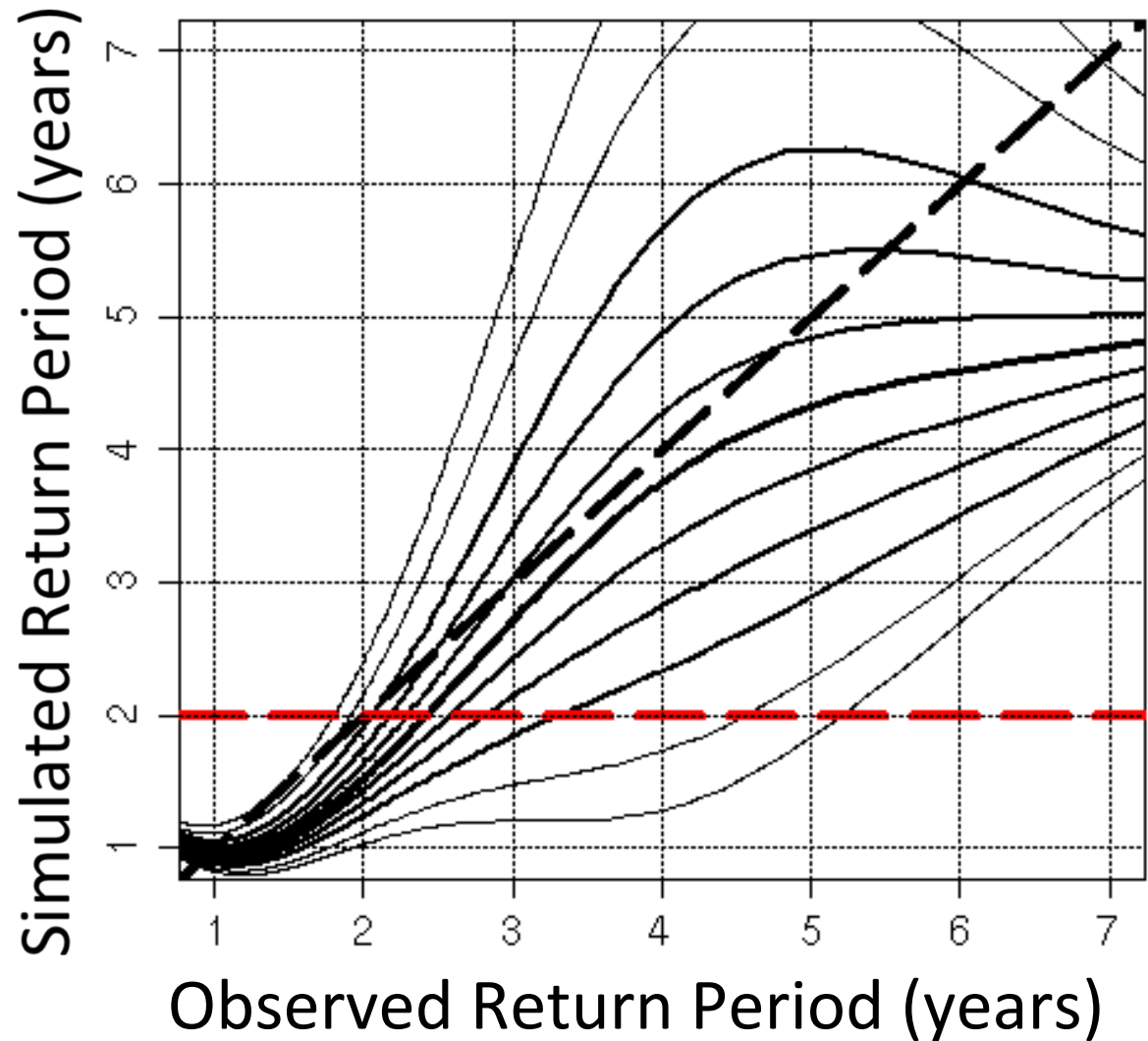
Normalize discharge (obs and sim) by converting to return periods



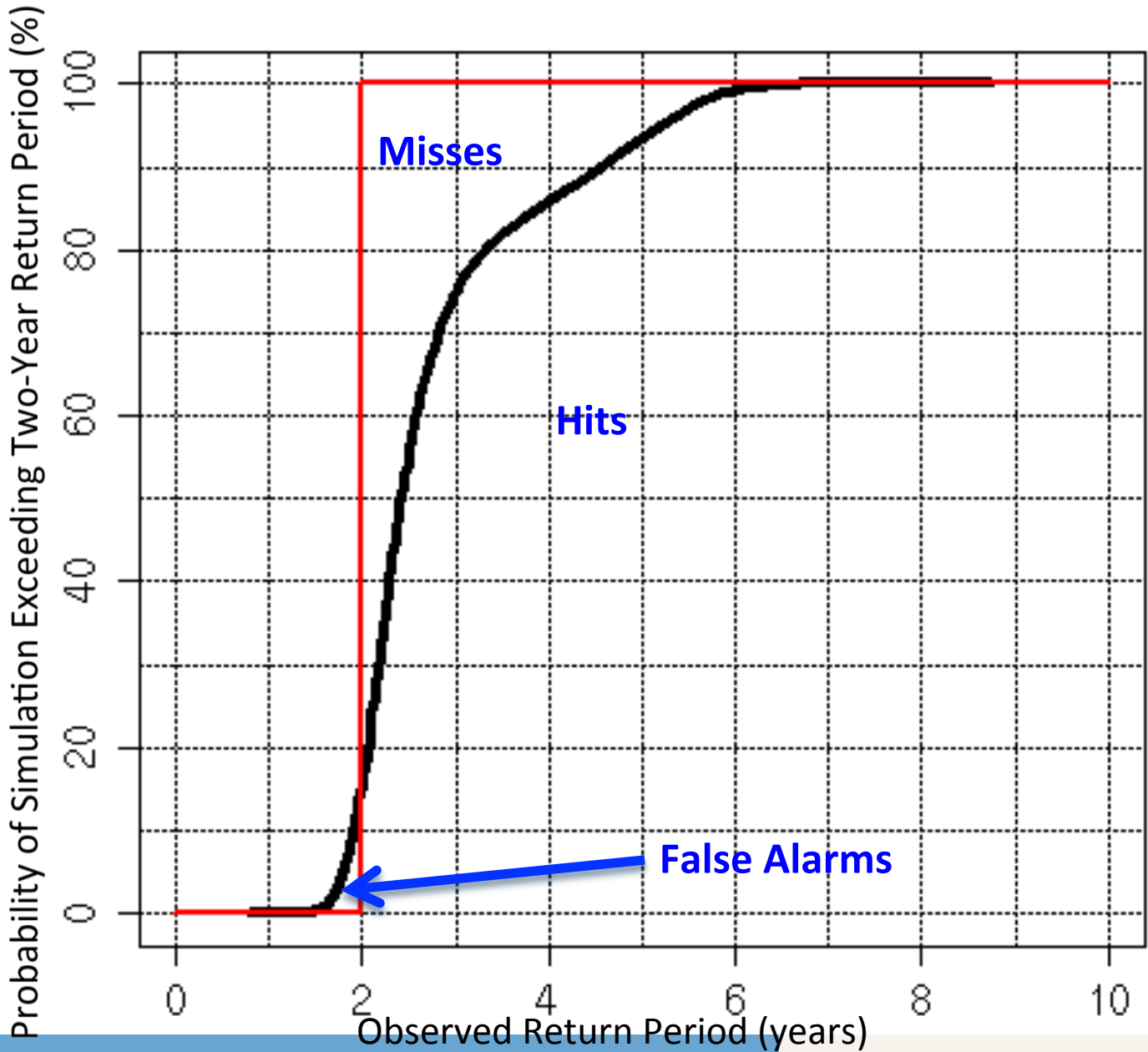


# FIT AN ERROR MODEL USING THE OBSERVED AND SIMULATED DATA

- Generalized Additive Model for Location, Scale and Shape (GAMLSS)
- Plot shows the quantiles computed from the error model (thickest is the median)



# EXAMPLE DIAGNOSTIC PLOT FOR A 2-YR RETURN PERIOD

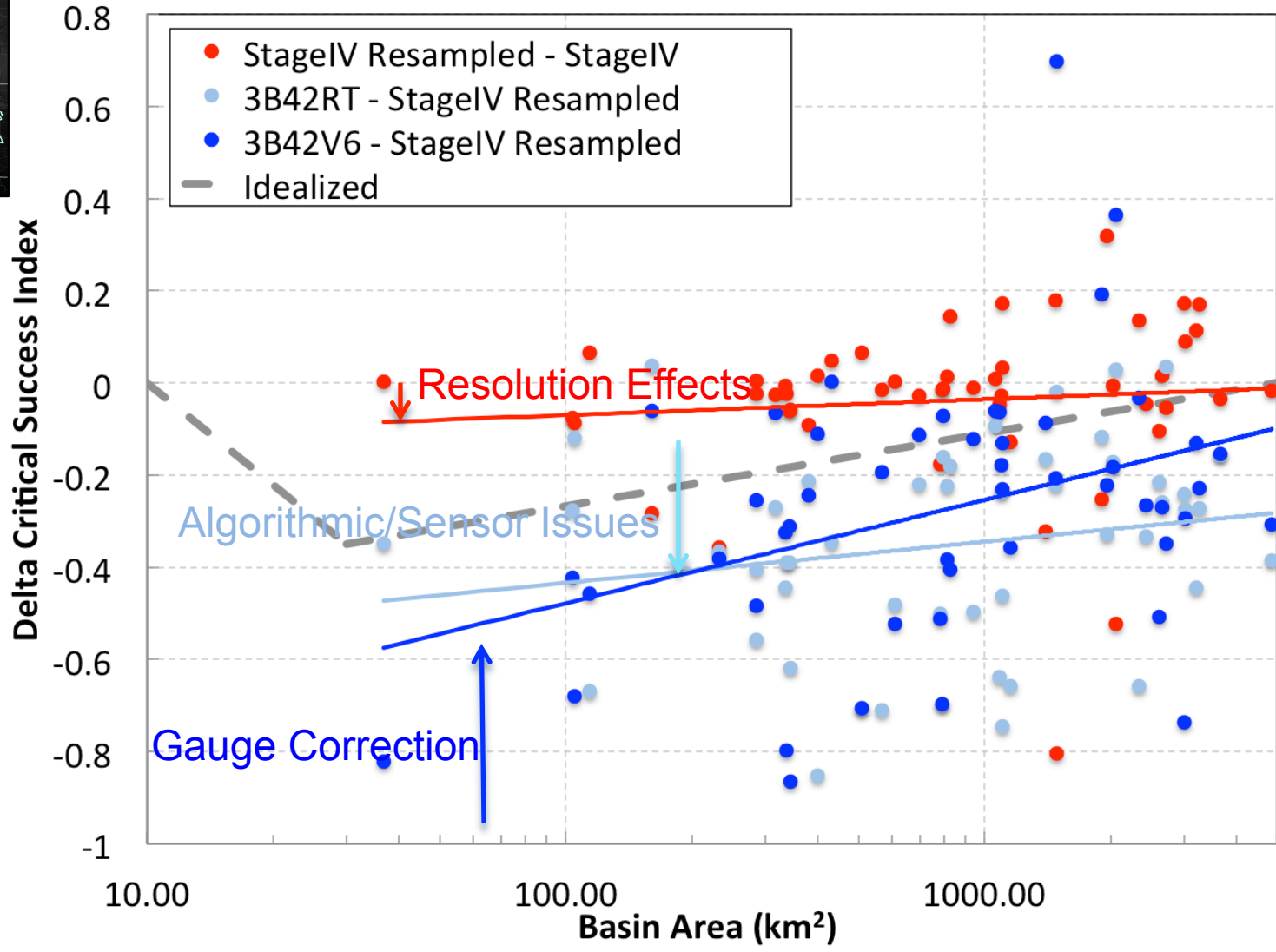
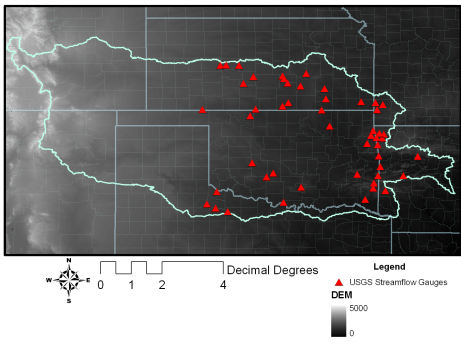


Red Line = Perfect

$$CSI = Hits / (Hits + Misses + False Alarms)$$



# COMPARE SIMULATIONS FROM 2002-2010 FOR ALL GAUGED BASINS IN THE ARKANSAS-RED RIVER BASIN



- Flood simulations for the complex terrain floods [e.g., Fella (2003) event] show some correlation with reference, but strong underestimation (5-10 times) of peak runoff; mean-field bias correction could not resolve the underestimation due to resolution effects
- Use of fine-scale (sub-pixel) ground-based radar (where available) and high-resolution NWP analysis can quantify satellite retrieval error characteristics and resolution effects when using in hydrologic applications
- **Satellite precipitation products will never be error free;** application-oriented frameworks must account for uncertainty in the retrievals, resolution effects, and how they combine with physical basin characteristics, scales and model structure