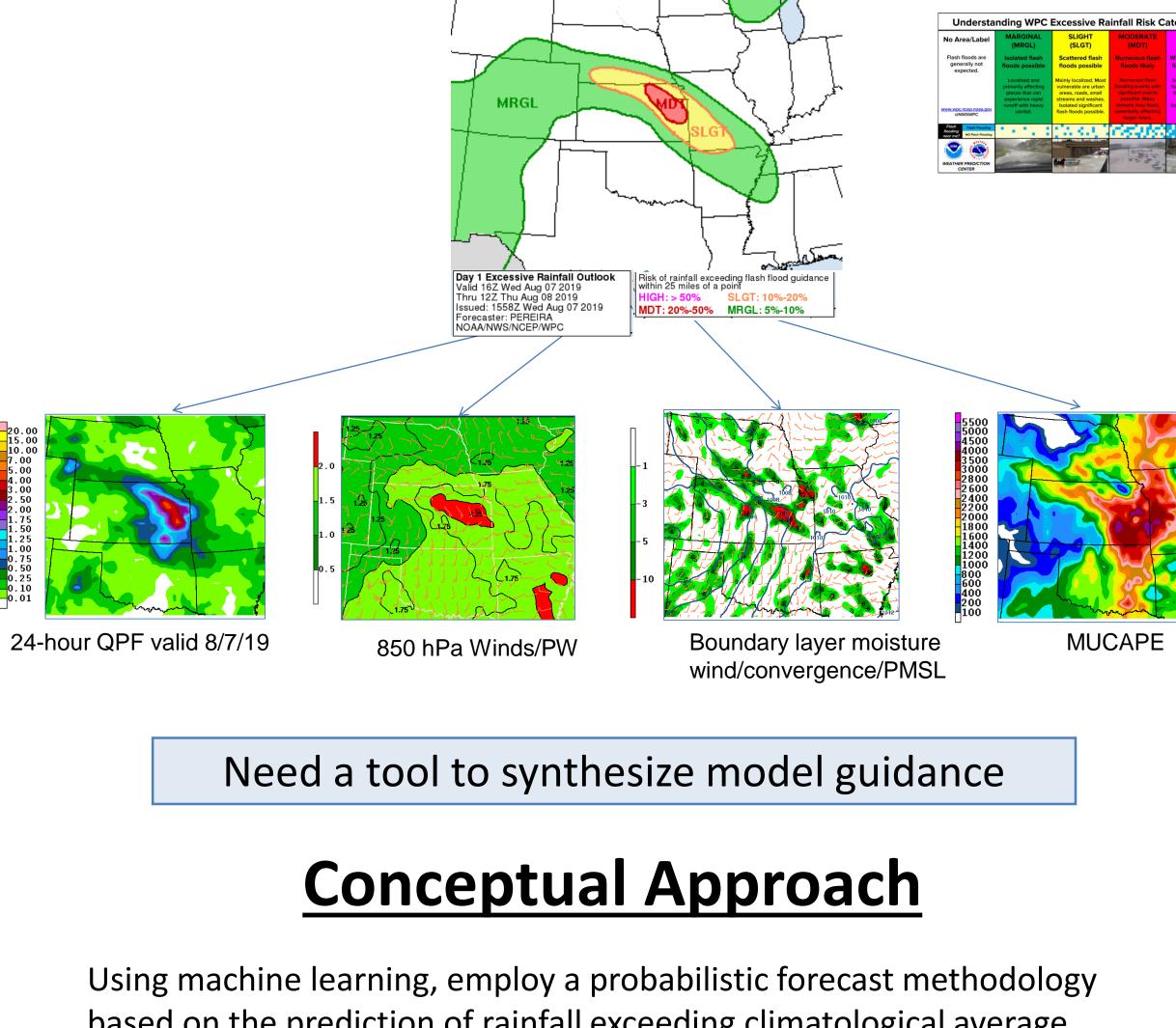


Forecast Challenge

WPC's Excessive Rainfall Outlooks (EROs) show the potential for rainfall to produce flash flooding over the CONUS for the Day 1-3 period. Forecasters review copious data to create the forecast.

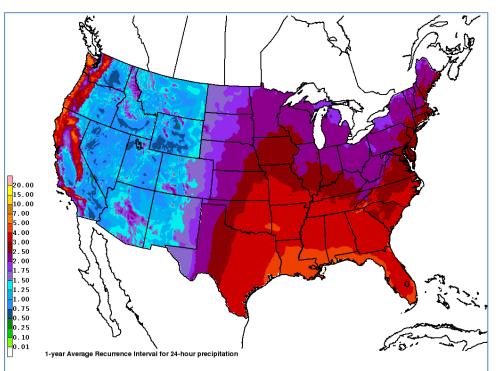


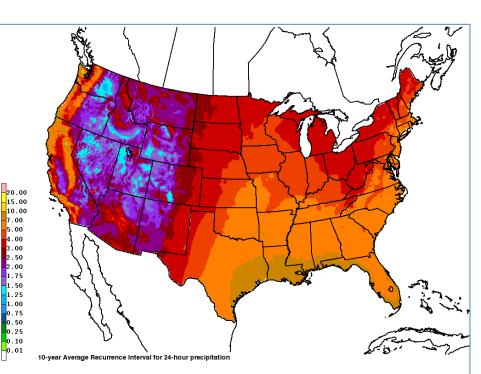
based on the prediction of rainfall exceeding climatological average recurrence intervals (ARIs).

ARIs:

- Better corresponds to actual impacts than fixed thresholds
- Do not bias toward climatologically wetter regions

24-hour precipitation average recurrence intervals over the CONUS





Project Goals

- Derive a forecast technique that can:
 - a. Combine QPF and multiple relevant atmospheric ingredients into a probabilistic forecast
 - b. Correct for numerical model timing and displacement biases
 - c. Run within WPC's computational framework
- 2. Create an operational "recommender or "first guess" for WPC's ERO

References

Herman, G.R. and R.S. Schumacher (2018): Money Doesn't Grow on Trees, but Forecasts Do: Forecasting Extreme Precipitation with Random Forests. *Monthly Weather Review*, 146, 1571-1600.

Herman, G.R. and R.S. Schumacher (2018b): 'Dendrology' in Numerical Weather Prediction: What Random Forests and Logistic Regression Tell Us About Forecasting Extreme Precipitation. Monthly Weather Review, 146, 1785-1812.

Use of a Machine Learning Algorithm in the Prediction of **Extreme Rainfall Events**

Mark Klein

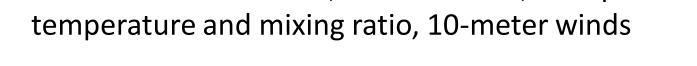
NOAA/NWS/Weather Prediction Center, College Park, MD

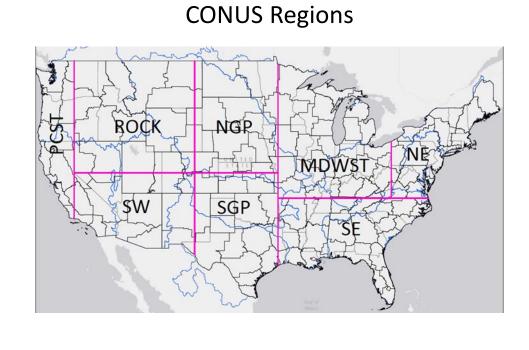
Machine Learning Technique

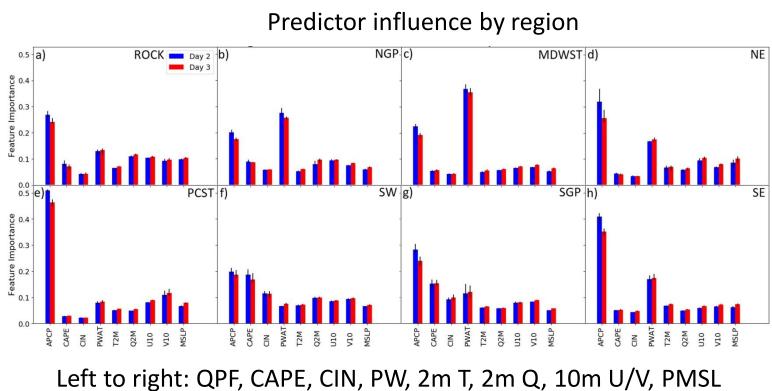
Developed at Colorado State University and evaluated during three years of WPC's Flash Flood and Intense Rainfall experiments, a prediction system trained using Random Forests predicts the probability that a 1- or 2-year ARI will be exceeded within 40km of a point during a 24-hour period (in line with WPC's ERO spatial and temporal definition).

Random Forest Prediction system

- An ensemble of 1000 decision trees, each with a deterministic outcome aggregated to produce a forecast probability
- Trained for eight climatologically distinct CONUS regions using nearly 11 years of 00Z GEFS Reforecast initializations from January 2003 – August 2013
- Verification data: Stage IV QPE, CCPA, and flash flood Local Storm Reports (LSRs) Predictors include: QPF, CAPE and CIN, Precipitable Water, Mean Sea Level Pressure, 2-meter







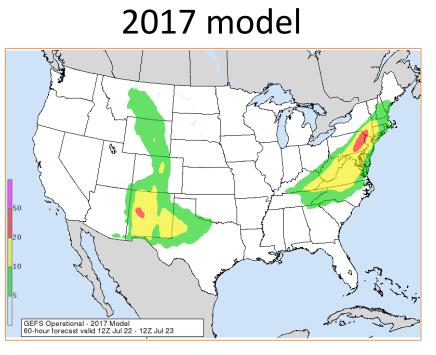
Predictor Influence

- QPF most predictive of ARI exceedances for *most* areas of the CONUS
- CAPE and QPF nearly equally predictive over the Southwest (monsoonal convection)
- Precipitable water more predictive than QPF over the Northern Plains and Midwestern regions

Forecast Products

Several versions of the model are running operationally at WPC using the GEFS Reforecast (GEFS/R) and 00Z/12Z runs of the operational GEFS (GEFS/O). Yearly changes based on feedback from FFaIR experiments.

Day 2 GEFS 24-hour ERO first guess valid 12Z July 23, 2019



Trained on exceedance of 1-year ARI

Spatial coverage generally too

Probabilities low-biased except

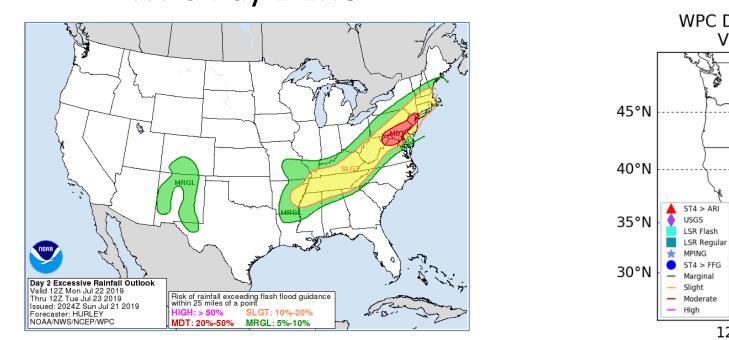
• Verification: Stage IV only

in Southwest

2017 Version

• FFaIR Feedback

low

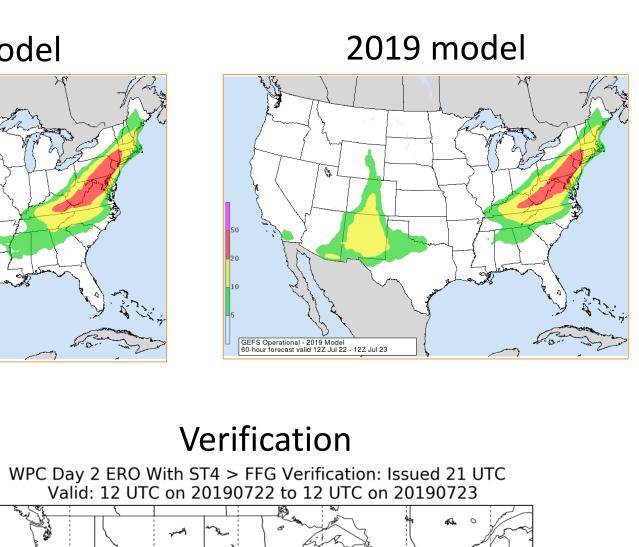


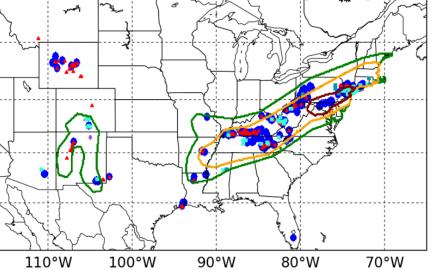
2018 Version

- Trained on exceedance of 1-yea
- and/or flash flood LSRs • Verification: Stage IV, CCPA, LSR
- FFaIR Feedback
- Spatial coverage too broad
- High bias, especially in central/northern Plains a Rockies

WPC Day 2 ERO

- 2018 model





2019 Version

flash flood LSRs

models

• FFaIR Feedback

• Trained on exceedance of 1- or 2-

• Verification: CCPA and LSRs

year ARI (ROCK/SW/NGP) and/or

Overall better areal coverage

and bias than 2017/2018

ar ARI	
Rs	
d	
nd	

