



# PRECIPITATION DISTRIBUTIONS ARE NOT NORMAL; CAN WE MAKE THEM LOOK THE PART?

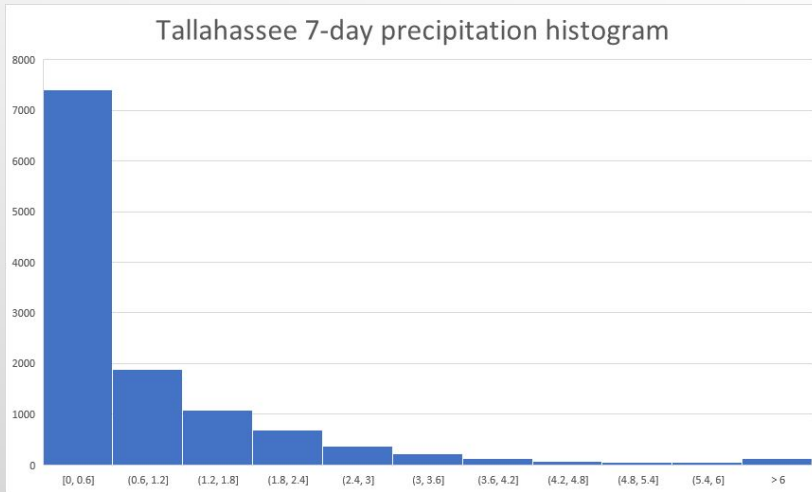
MICHAEL GOSS, EMERSON LAJOIE, CORY BAGGETT, ERICA BURROWS

WITH ADDITIONAL INPUT FROM: JOHNNA INFANTI, LAURA CIASTO, MATT ROSENCRANS, DAN  
COLLINS, LEIGH ZHANG

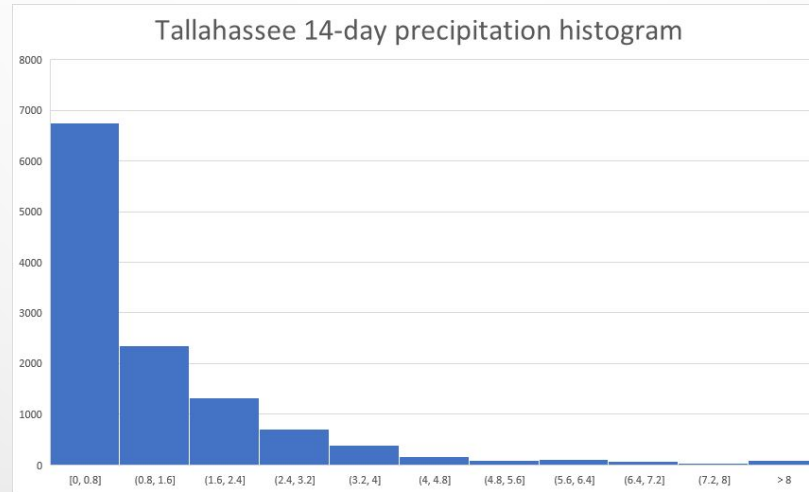


# THE PROBLEM

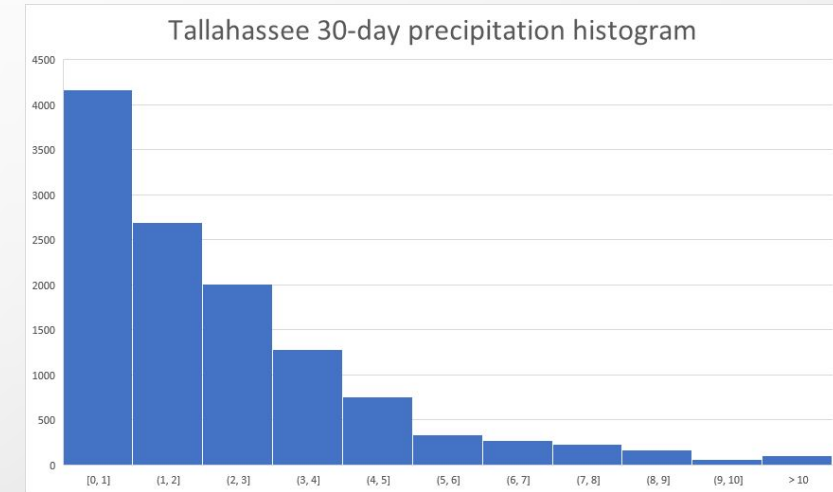
Tallahassee 7-day precipitation histogram



Tallahassee 14-day precipitation histogram



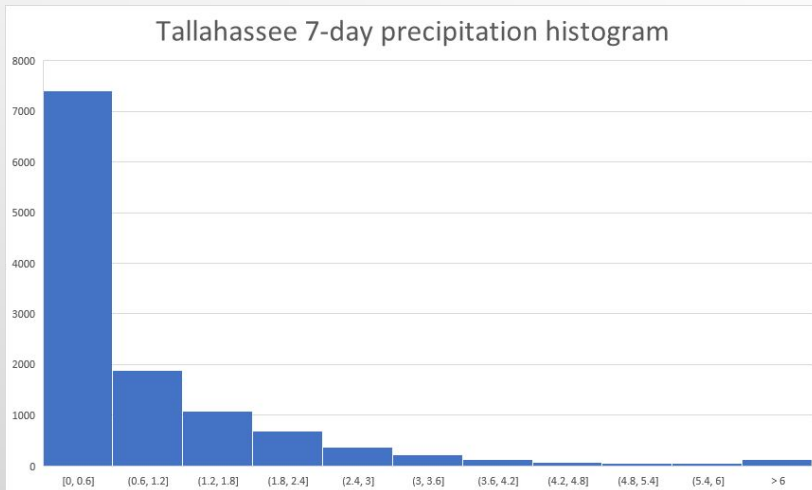
Tallahassee 30-day precipitation histogram



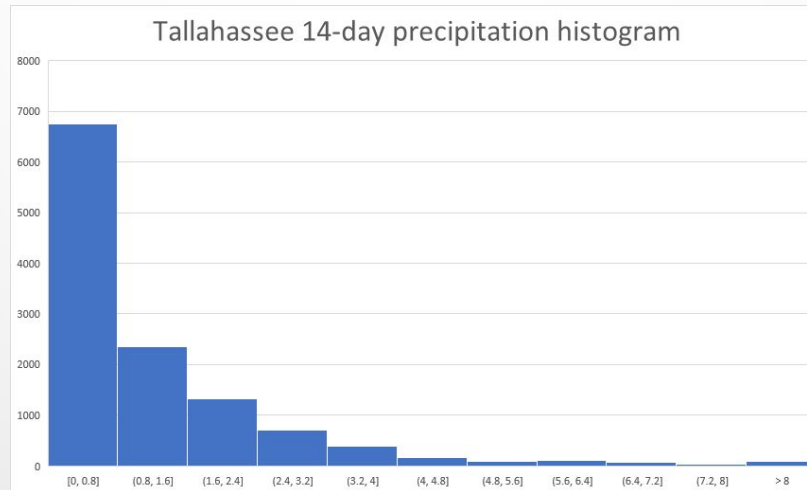
- Precipitation distributions are rarely close to normal/Gaussian, especially for:
  - Shorter integration periods
  - Drier regions

# THE PROBLEM

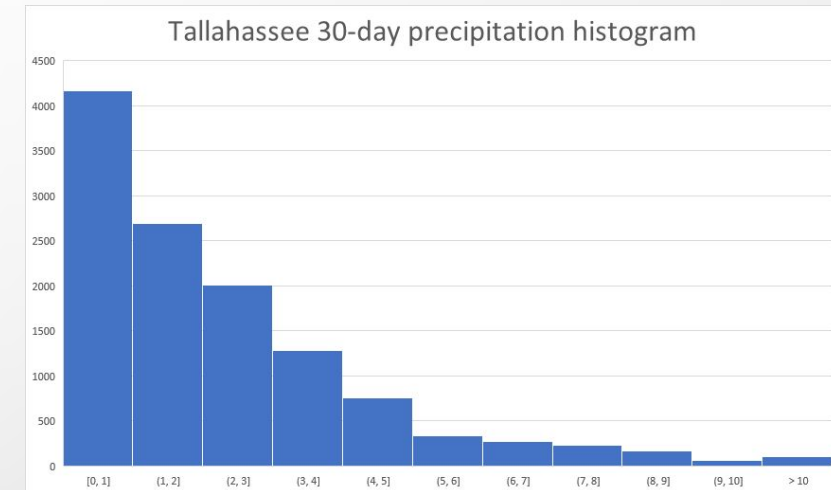
Tallahassee 7-day precipitation histogram



Tallahassee 14-day precipitation histogram

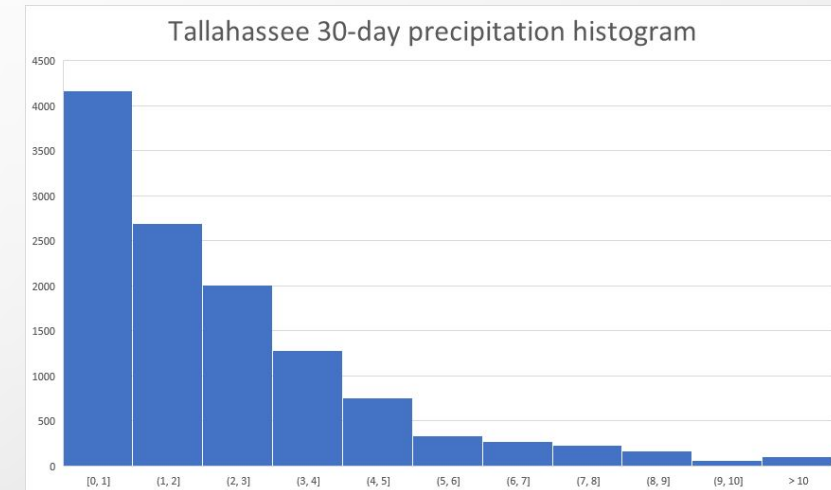
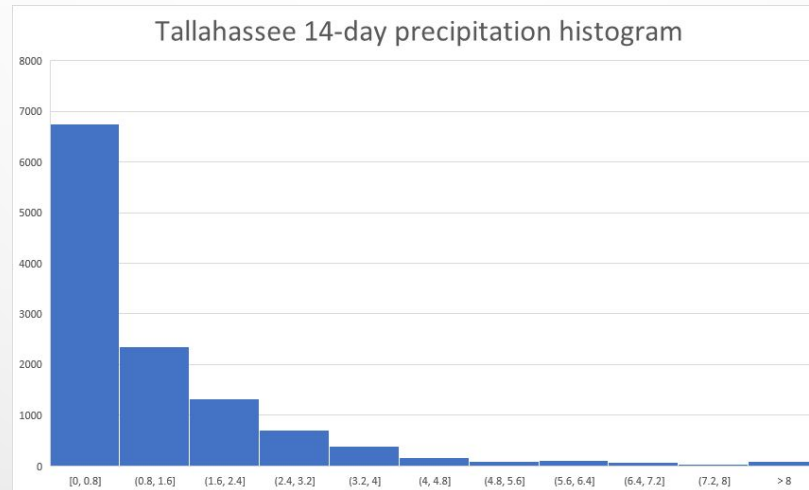
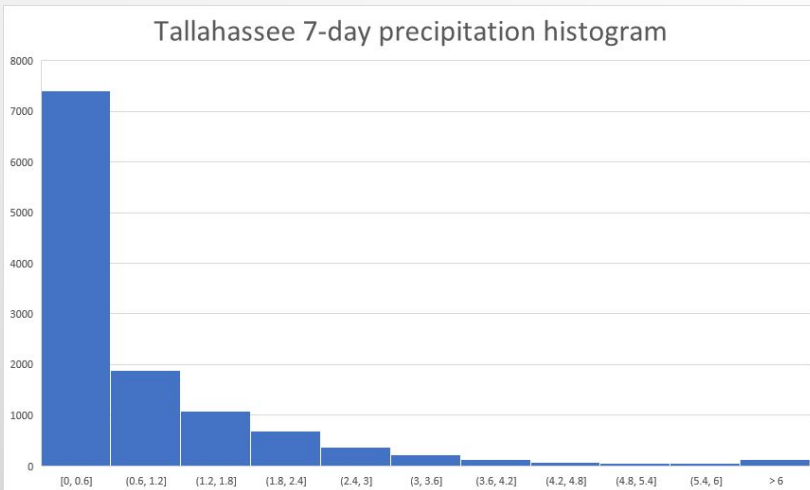


Tallahassee 30-day precipitation histogram



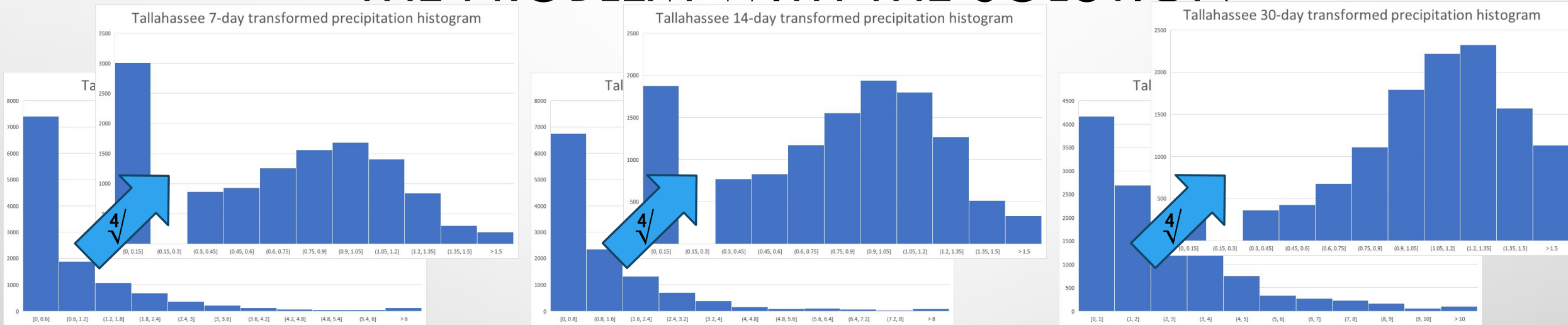
- Normally distributed data helps improve:
  - Reliability; especially in the context of two-category/three-category forecasting
  - Proper fitting in models/regressions that expect Gaussian data (i.e. some MLR code we use at CPC)

# THE SOLUTION?



- One method of addressing this issue is to use one of many standard functions to transform the distribution
- Commonly, CPC uses 4<sup>th</sup>-root, square-root, or logarithmic transformations with precipitation data
  - Truly, probably a gamma distribution is “best”

# THE PROBLEM WITH THE SOLUTION



- The “best” transform function for one situation may be quite poor for another
  - Transformations can be quite sensitive to integration period, grid point, and season

## WHAT WE WANT

- Transformations that give more normally distributed precipitation data as a function of:
  - Integration period
  - Grid point
  - Season
- “Best” transformation parameters that don’t take too long to derive or calculate

## WHAT WE DID

- Box-Cox transformation
  - One parameter
  - Parameter ( $\lambda$ ) values of 0,  $1/4$ ,  $1/2$  correspond with already-used log, 4<sup>th</sup>-root, and square-root transforms

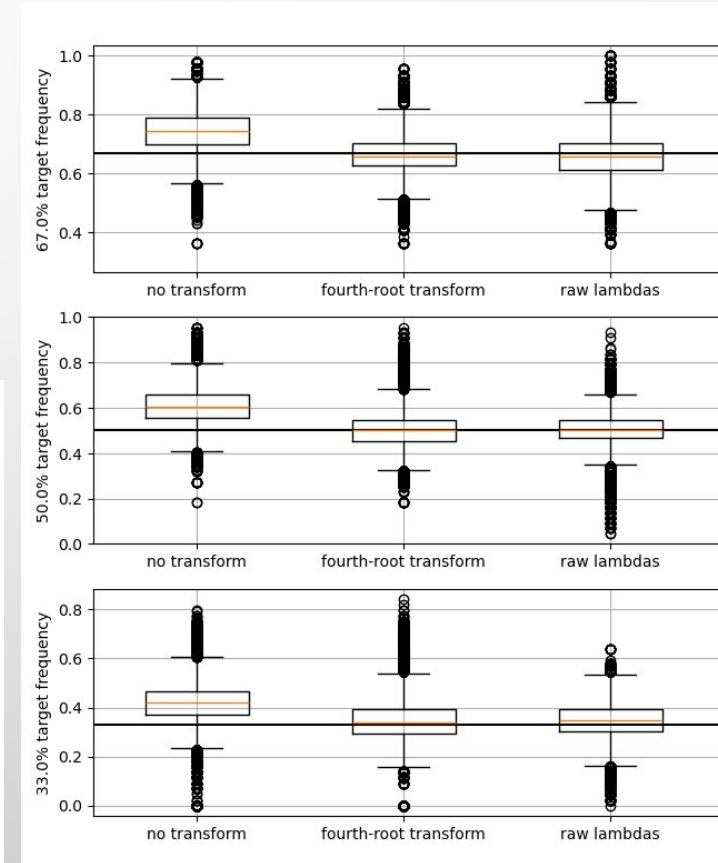
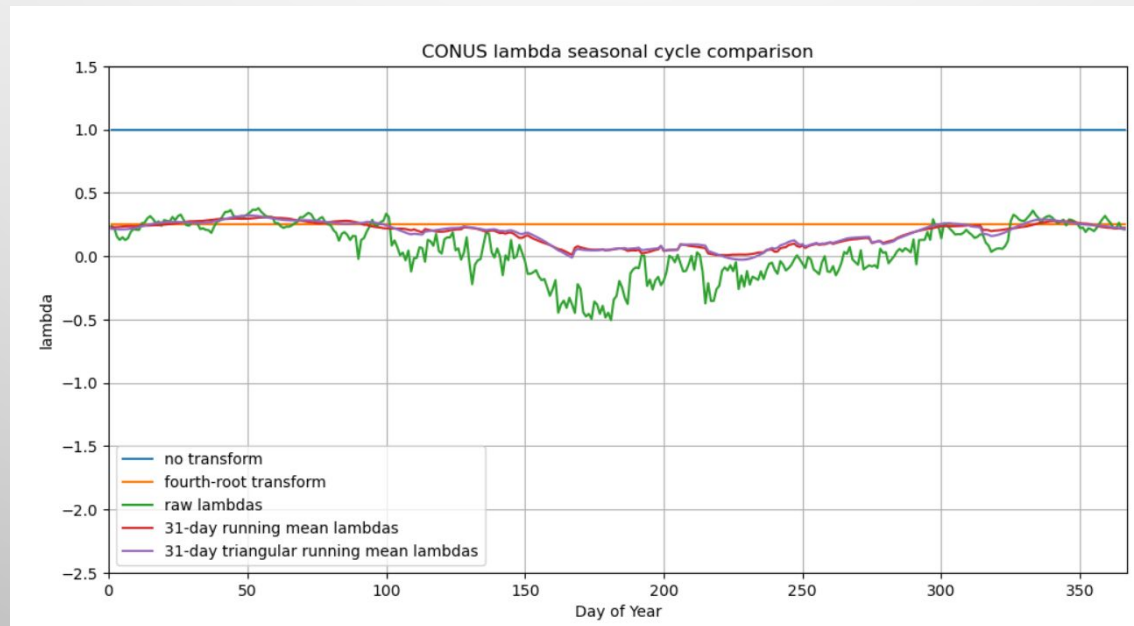
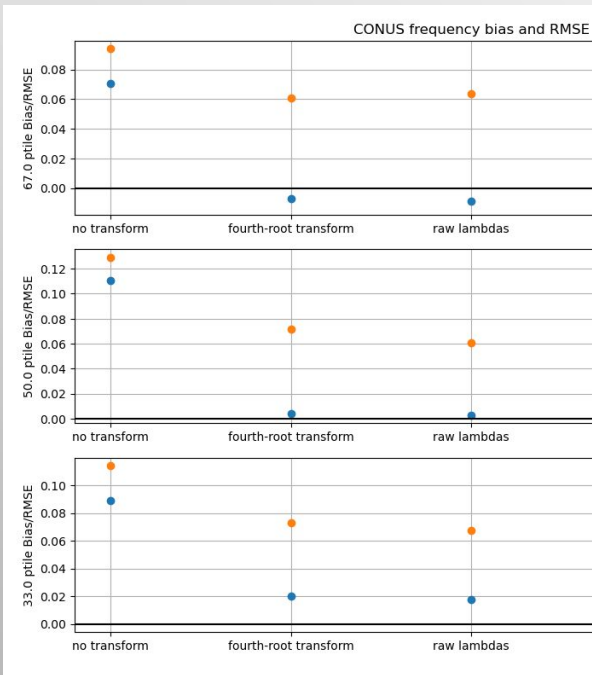
$$\psi(y, \lambda) = \begin{cases} \frac{(y+1)^\lambda - 1}{\lambda} & y \geq 0 \text{ and } \lambda \neq 0, \\ \log(y+1) & y \geq 0 \text{ and } \lambda = 0, \\ -\frac{(-y+1)^{2-\lambda} - 1}{2-\lambda} & y < 0 \text{ and } \lambda \neq 2, \\ -\log(-y+1) & y < 0, \lambda = 2. \end{cases}$$

Yeo and Johnson (2000); Also Wilks' textbook

- $\lambda$  values that minimize the  $d_\lambda$  statistic (Hinkley, 1977)
  - Essentially, the normalized difference between the mean and the median

# WHAT WE FOUND

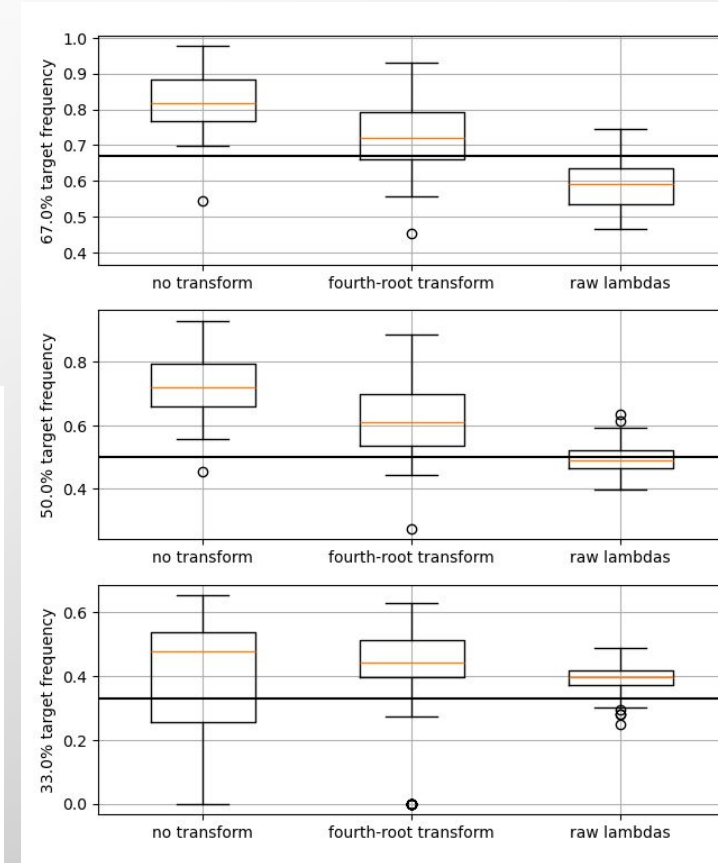
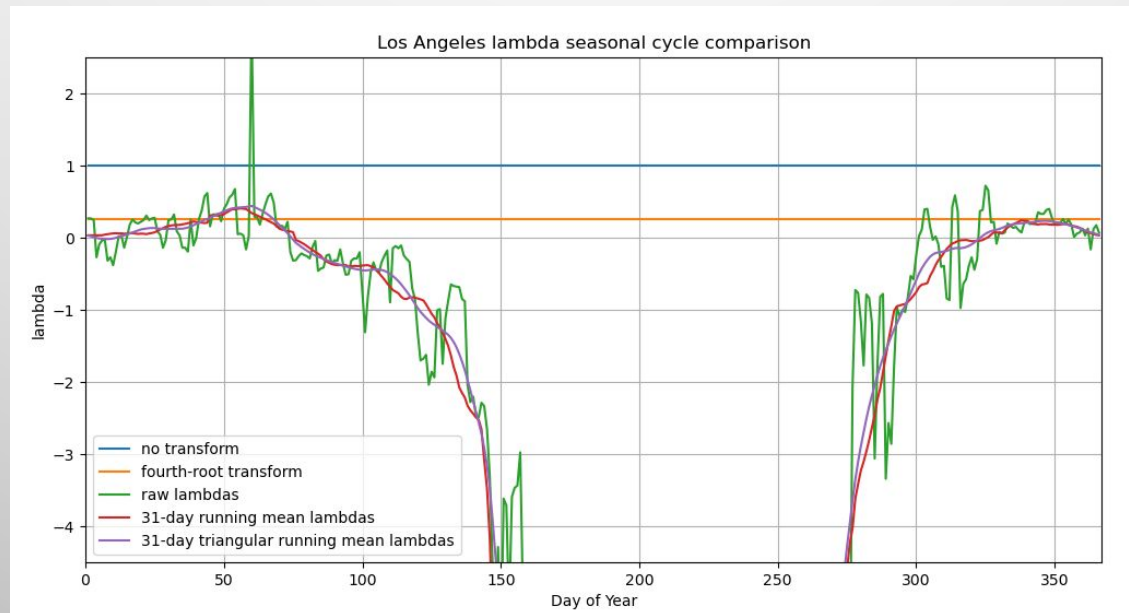
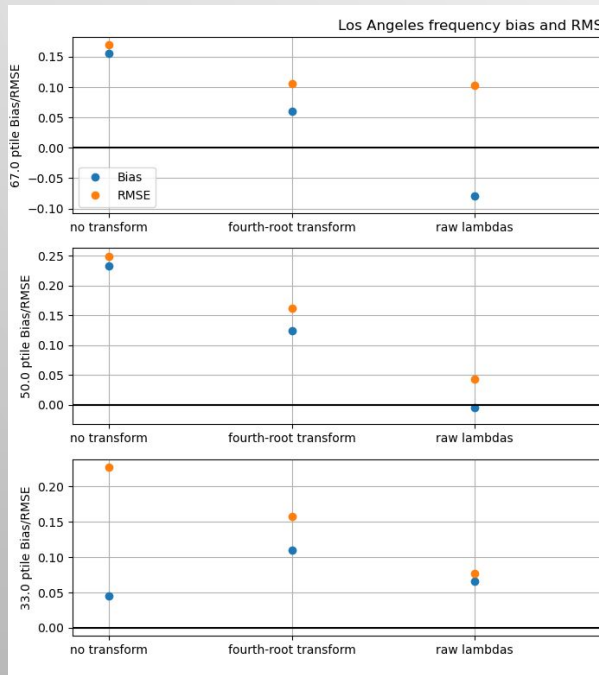
- For CONUS as a whole, the 4<sup>th</sup>-root transformation is indeed a pretty good approximation for 14-day precipitation





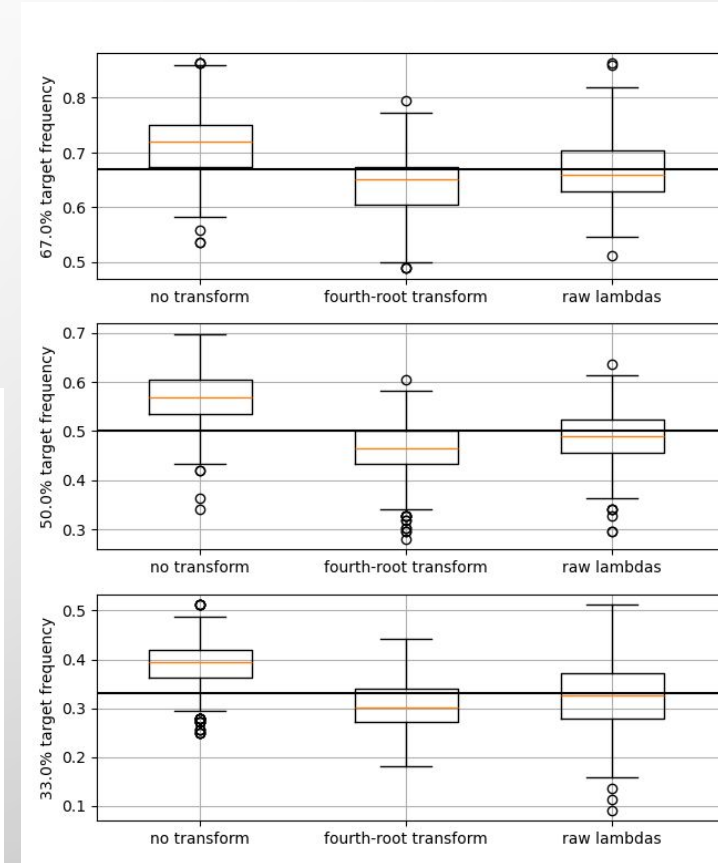
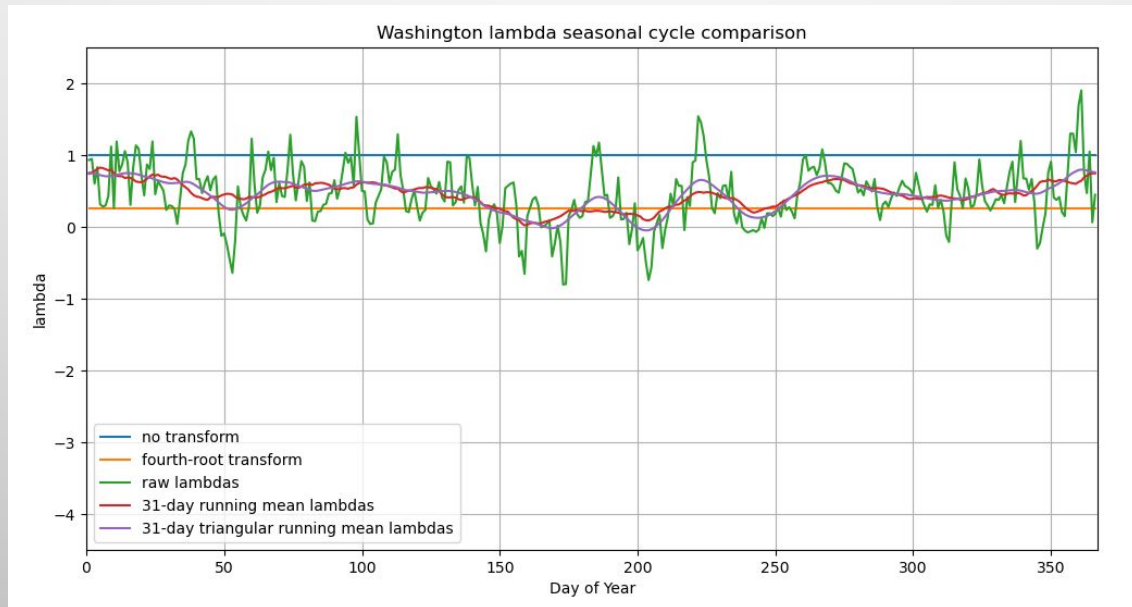
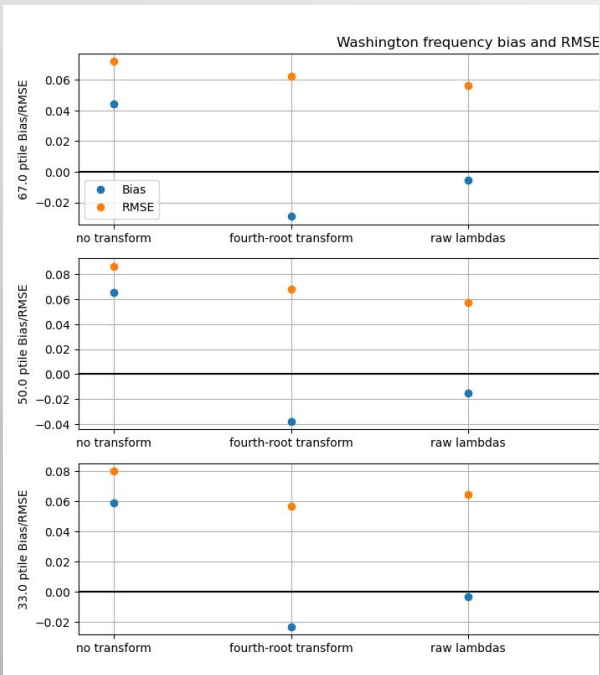
# WHAT WE FOUND

- However, the best  $\lambda$  varies substantially by integration period (not shown), grid point, and season



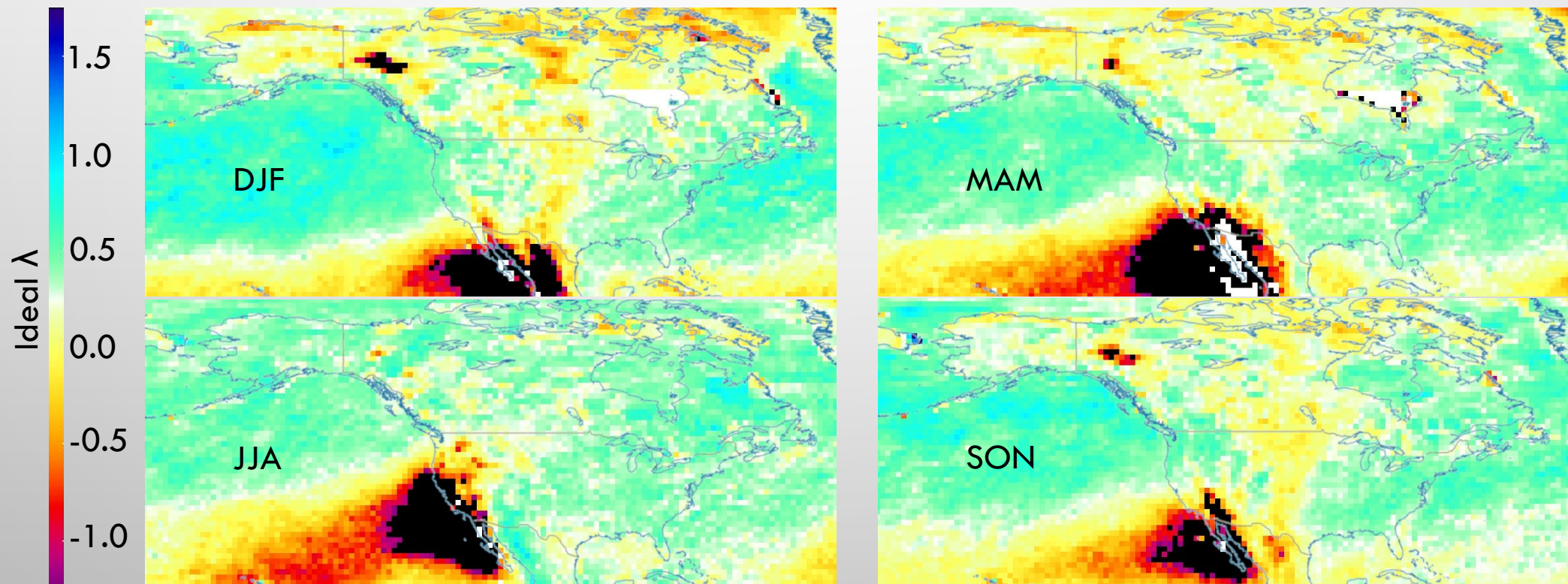
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## WHAT WE'RE DOING

- Now that we have  $\lambda$  as a function of integration period, grid point, and day of year (and python functions to transform and inverse-transform using a map of lambdas!)...
  - Emerson is currently testing some MLR code with the newly transformed precipitation data to test whether or not categorical forecast skill can be improved
  - Preliminary results are equivocal, but more testing needed!

# THANK YOU!

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