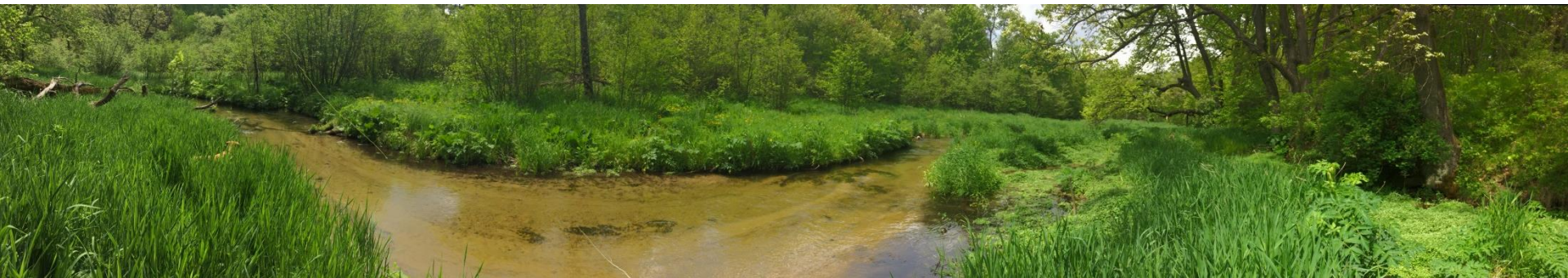
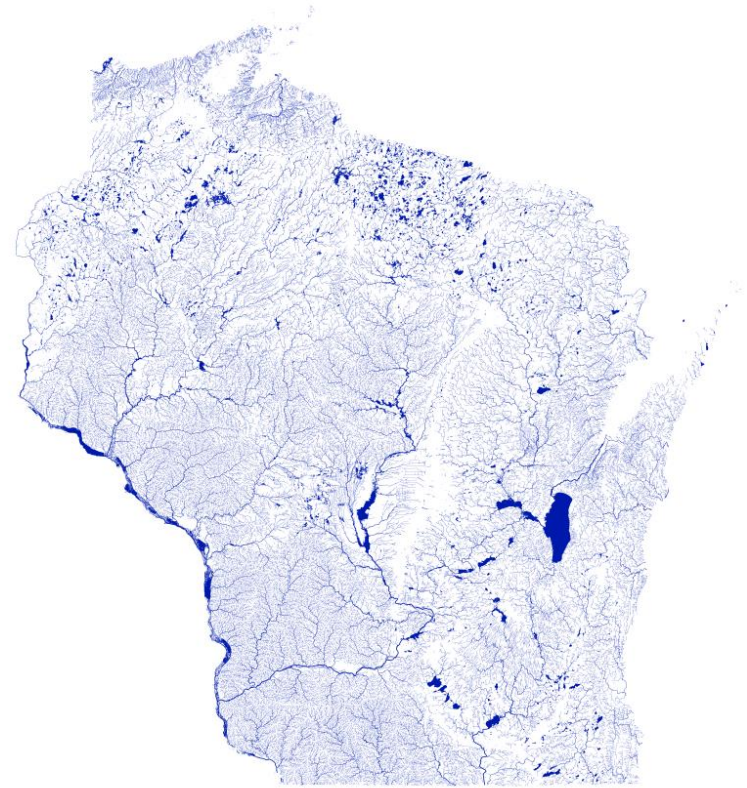




Understanding and Forecasting Variation on Water Levels

Bob Smail
Wisconsin DNR
Water Use Section

AWRA - WI
Appleton, WI
March, 8 2018



Lake Michigan





2012

Long Lake Waushara Co, WI

WI Center for
Investigative Journalism



2017

Lost Land Lake
Sawyer Co, WI



Monday July 11, 2016

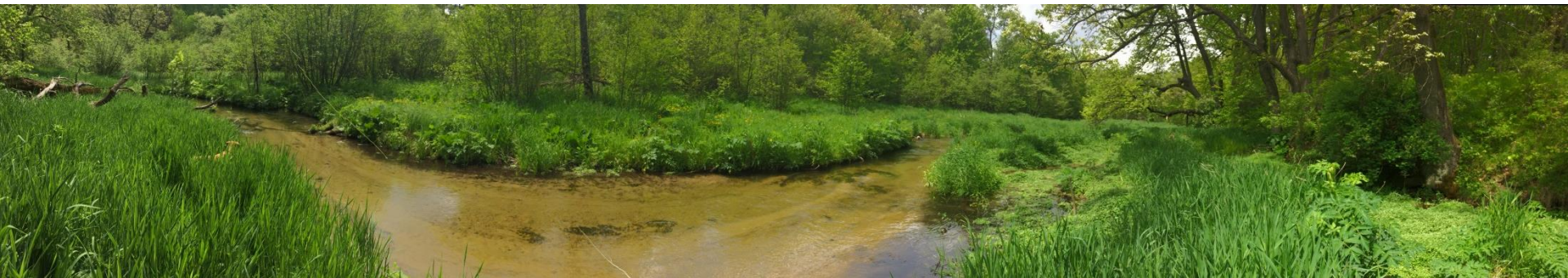


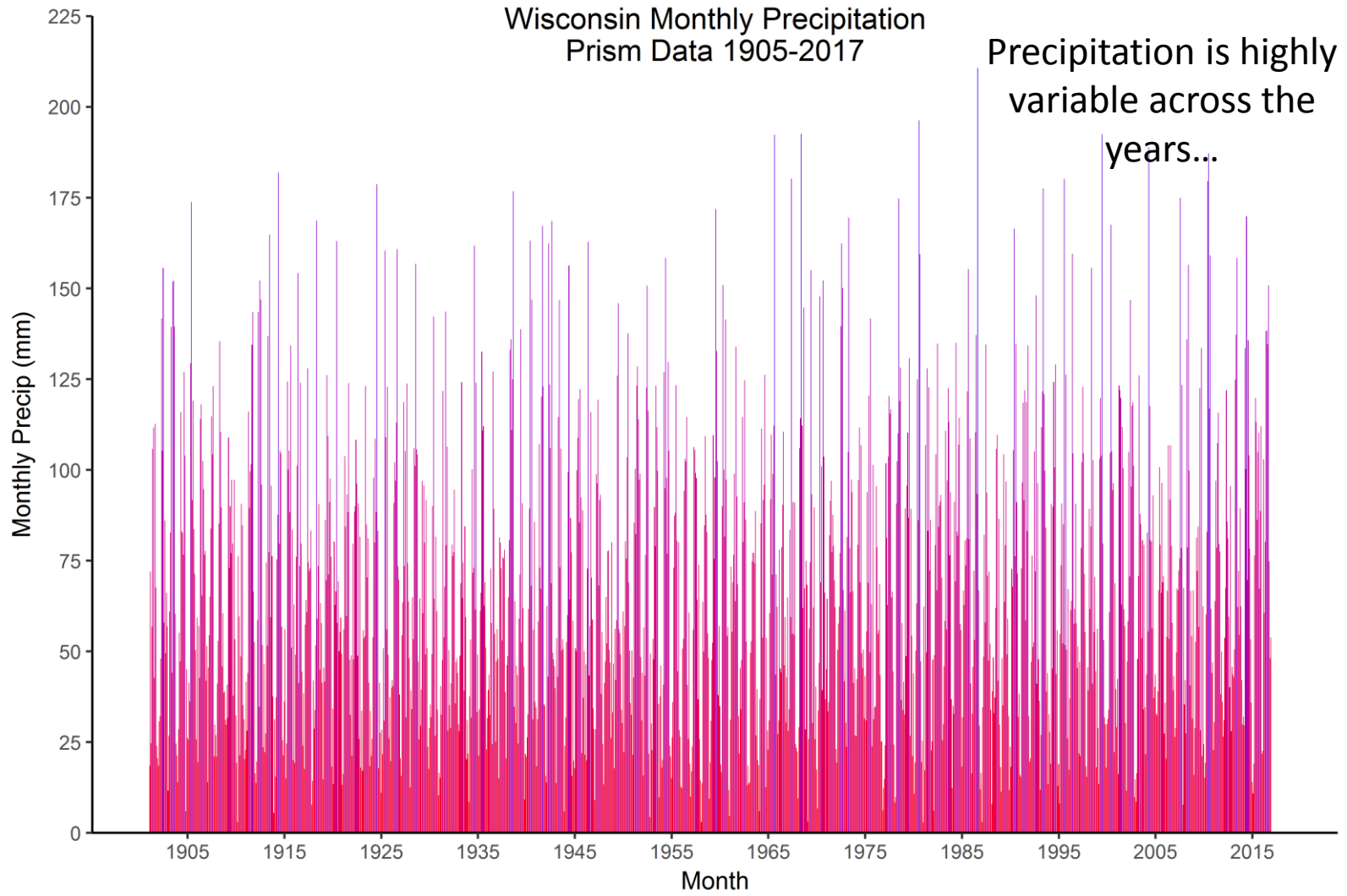
Thursday July 14, 2016



Understanding and Forecasting Variation on Water Levels

- ✓ Interpreting Precipitation
- Drivers of Water Level variation
- Forecasting Future Water Levels

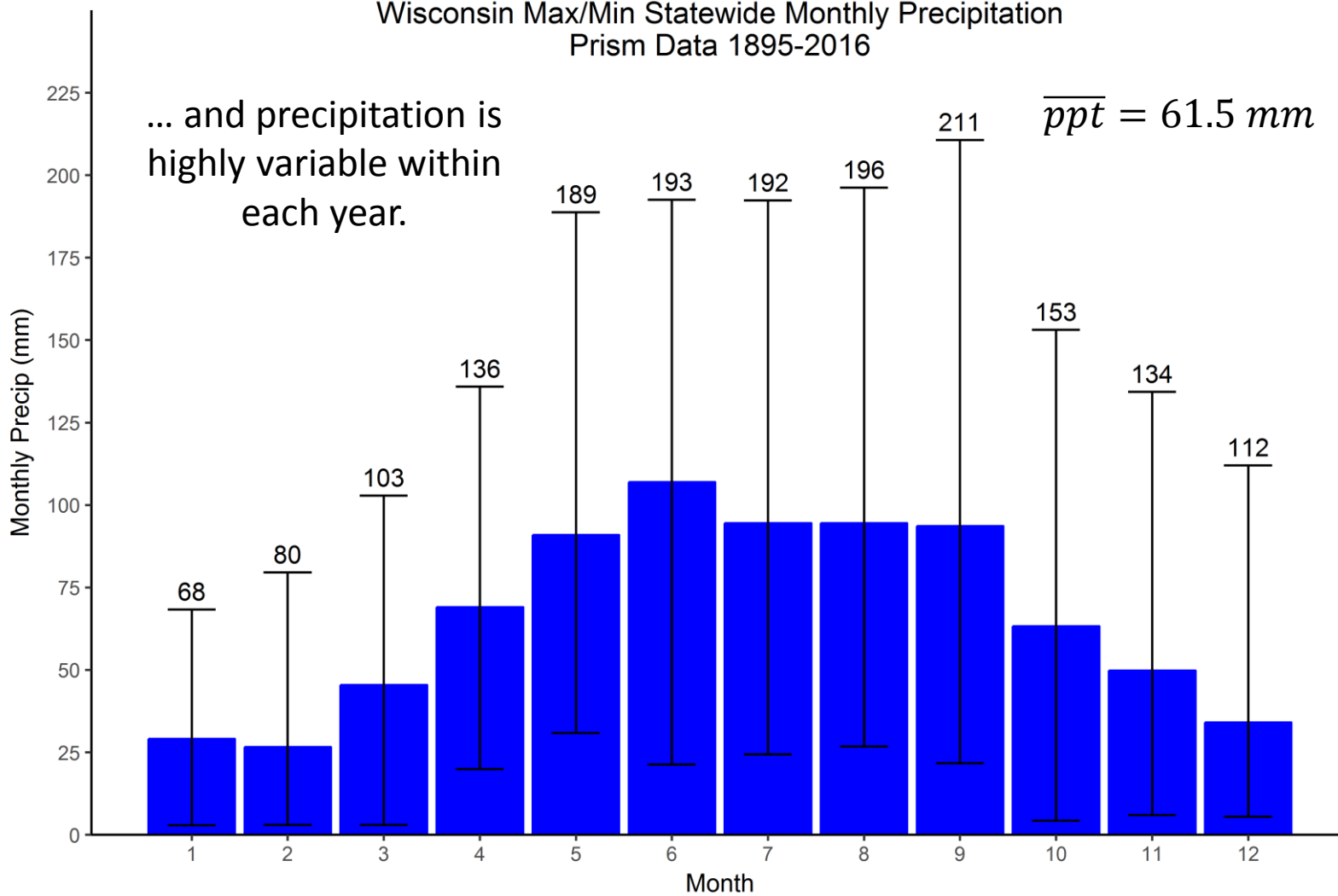




Wisconsin Max/Min Statewide Monthly Precipitation
Prism Data 1895-2016

... and precipitation is highly variable within each year.

$$\overline{ppt} = 61.5 \text{ mm}$$





Understanding and Forecasting Variation on Water Levels

- We know that variability in precipitation is strongly related to changes in water levels.
- Hydrological models are best answer but are very time intensive, costly and usually limited to smaller areas.
- Can precipitation data be interpreted to serve as the expected water level given local weather?

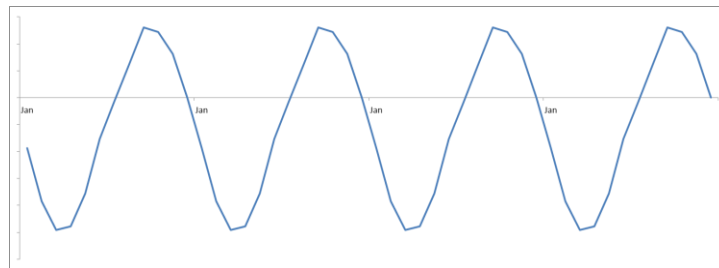
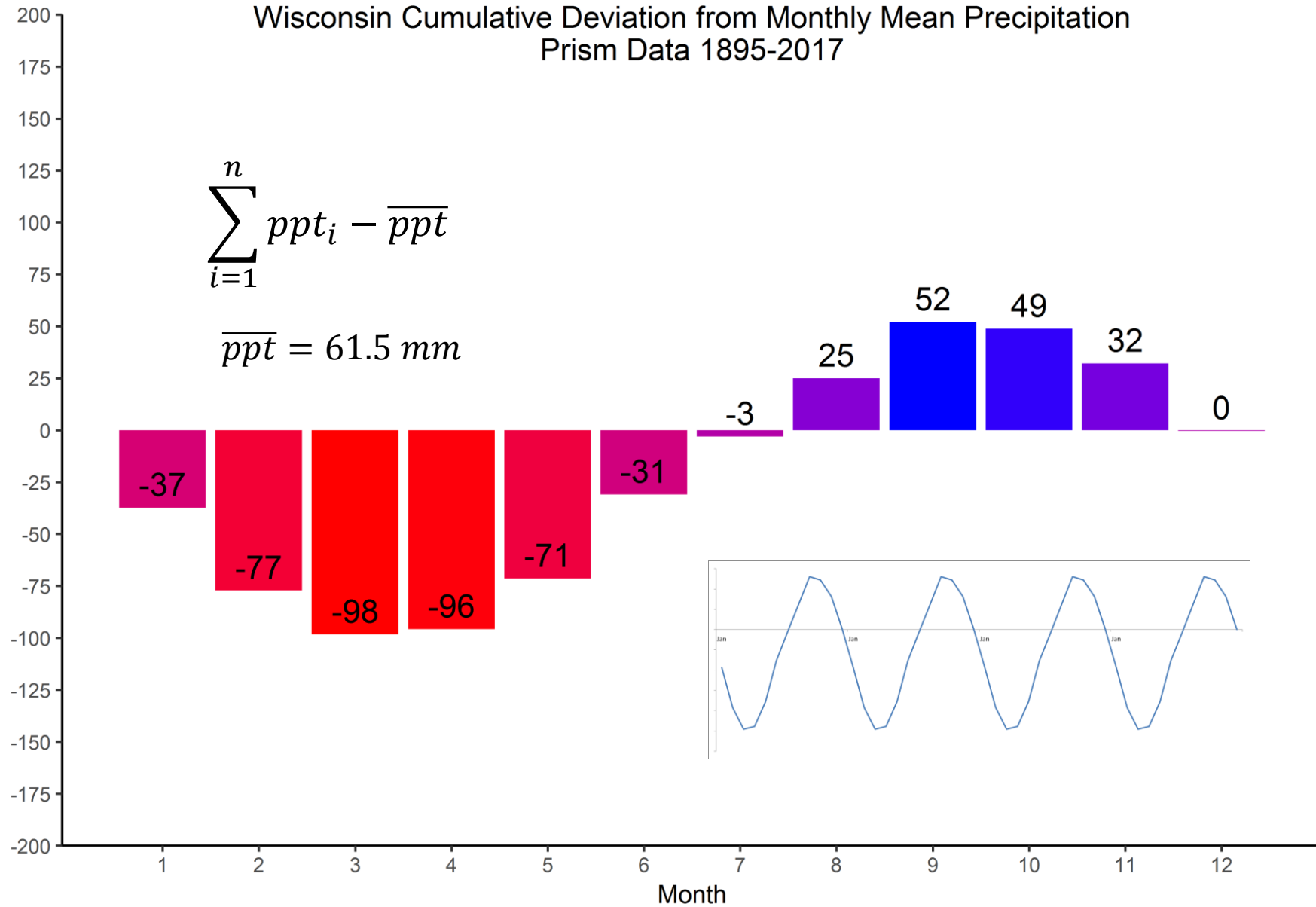


Wisconsin Cumulative Deviation from Monthly Mean Precipitation Prism Data 1895-2017

$$\sum_{i=1}^n ppt_i - \overline{ppt}$$

$$\overline{ppt} = 61.5 \text{ mm}$$

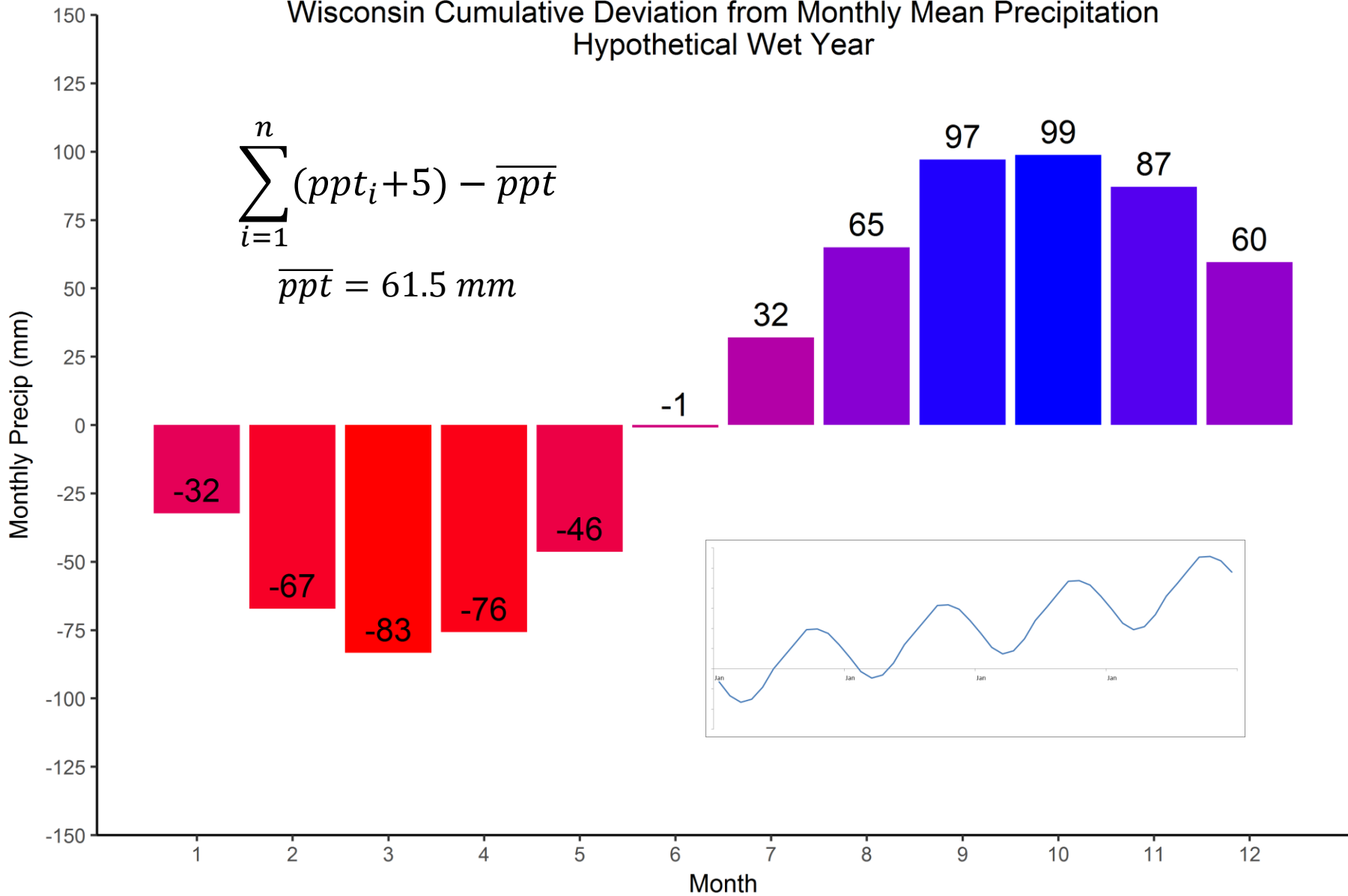
Monthly Precip (mm)



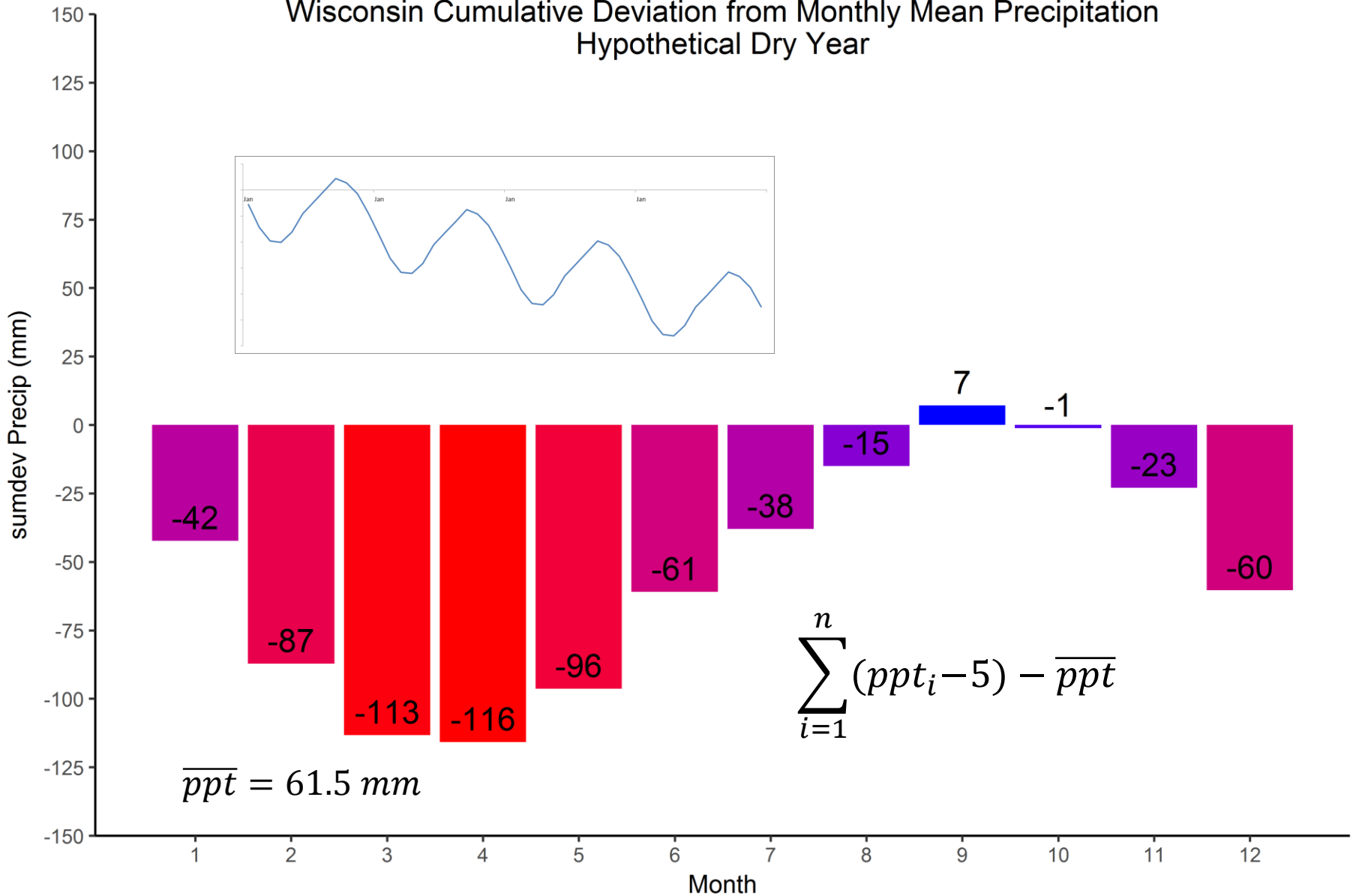
Wisconsin Cumulative Deviation from Monthly Mean Precipitation Hypothetical Wet Year

$$\sum_{i=1}^n (ppt_i + 5) - \overline{ppt}$$

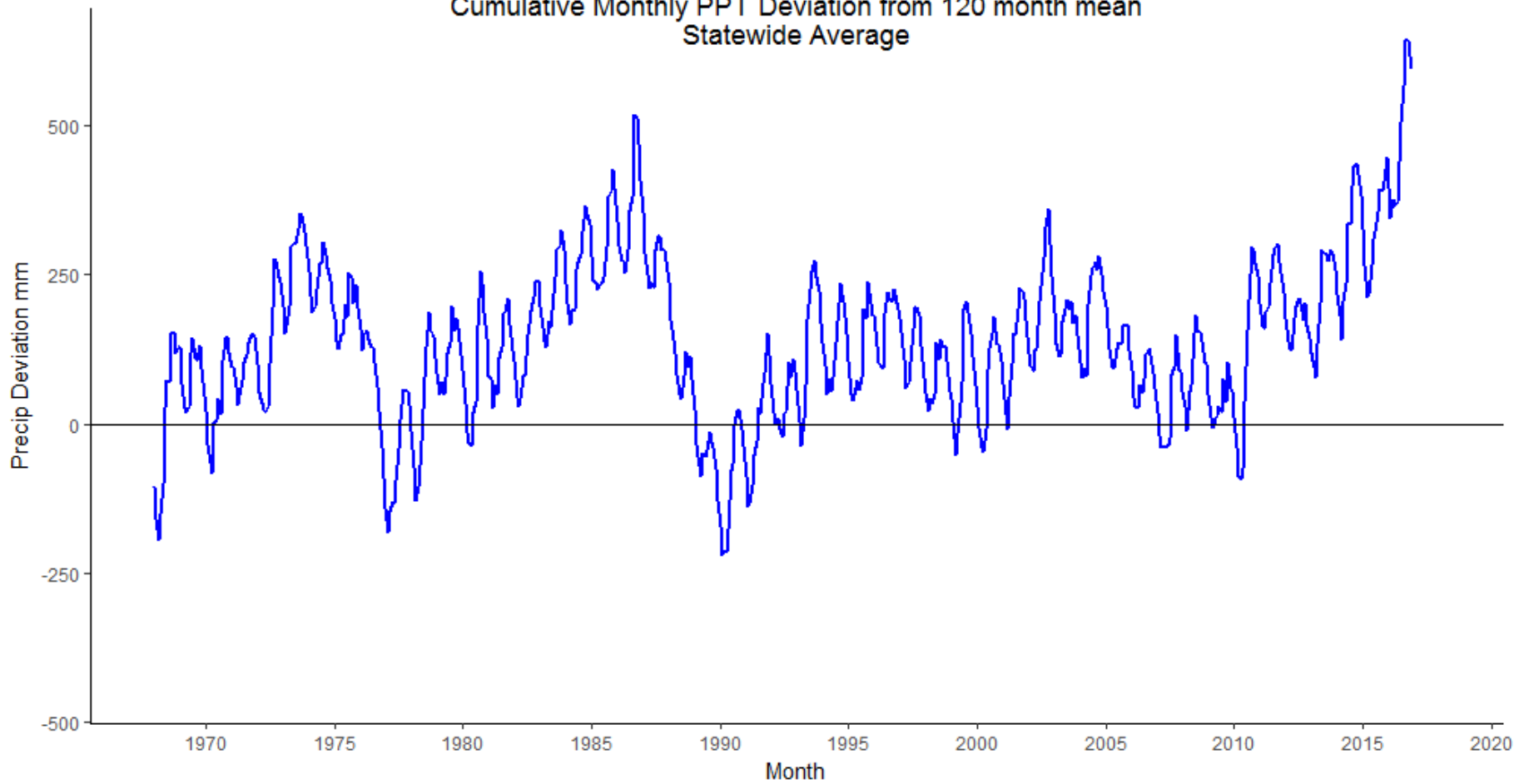
$\overline{ppt} = 61.5 \text{ mm}$



Wisconsin Cumulative Deviation from Monthly Mean Precipitation Hypothetical Dry Year

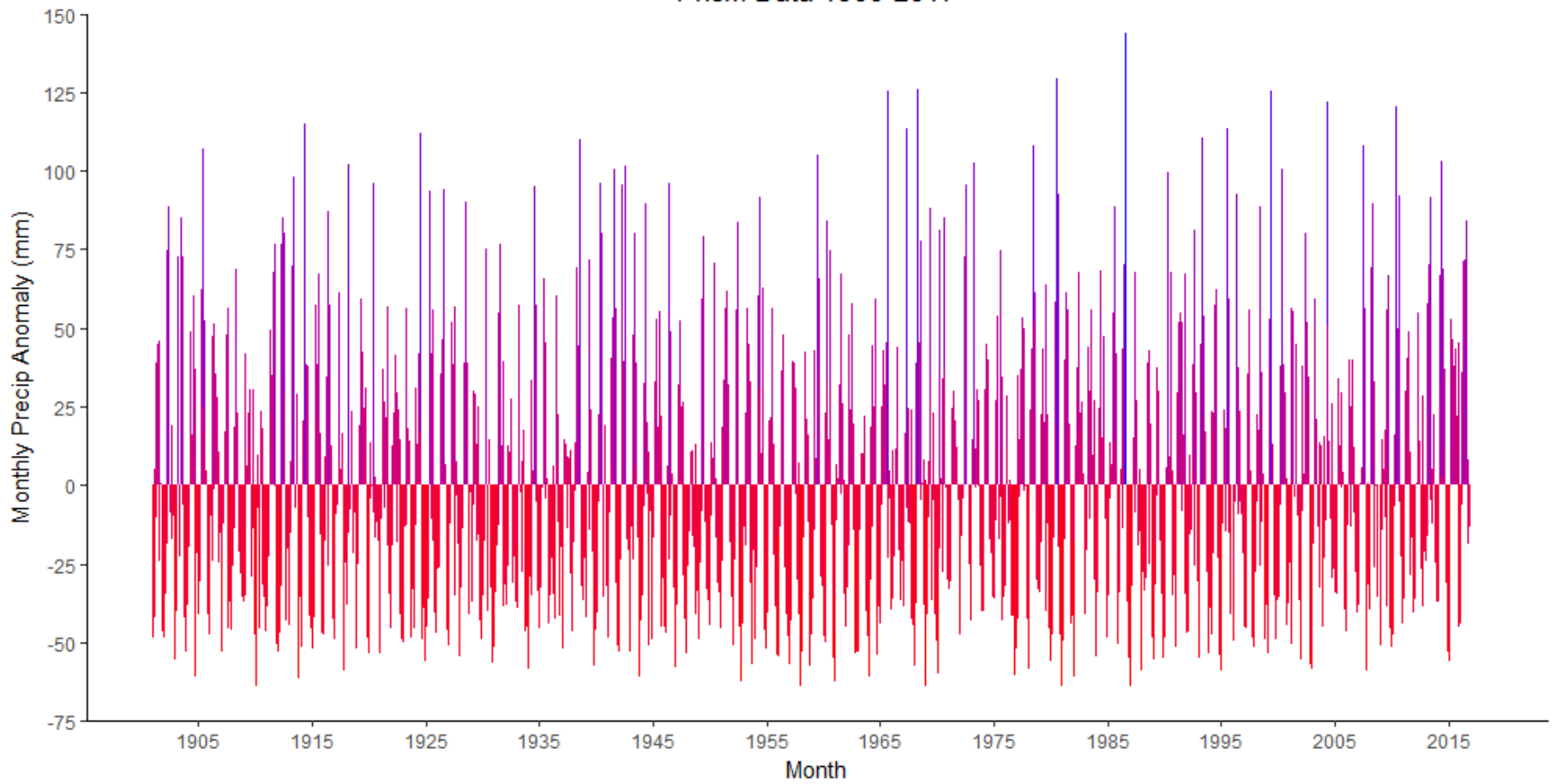


Cumulative Monthly PPT Deviation from 120 month mean
Statewide Average

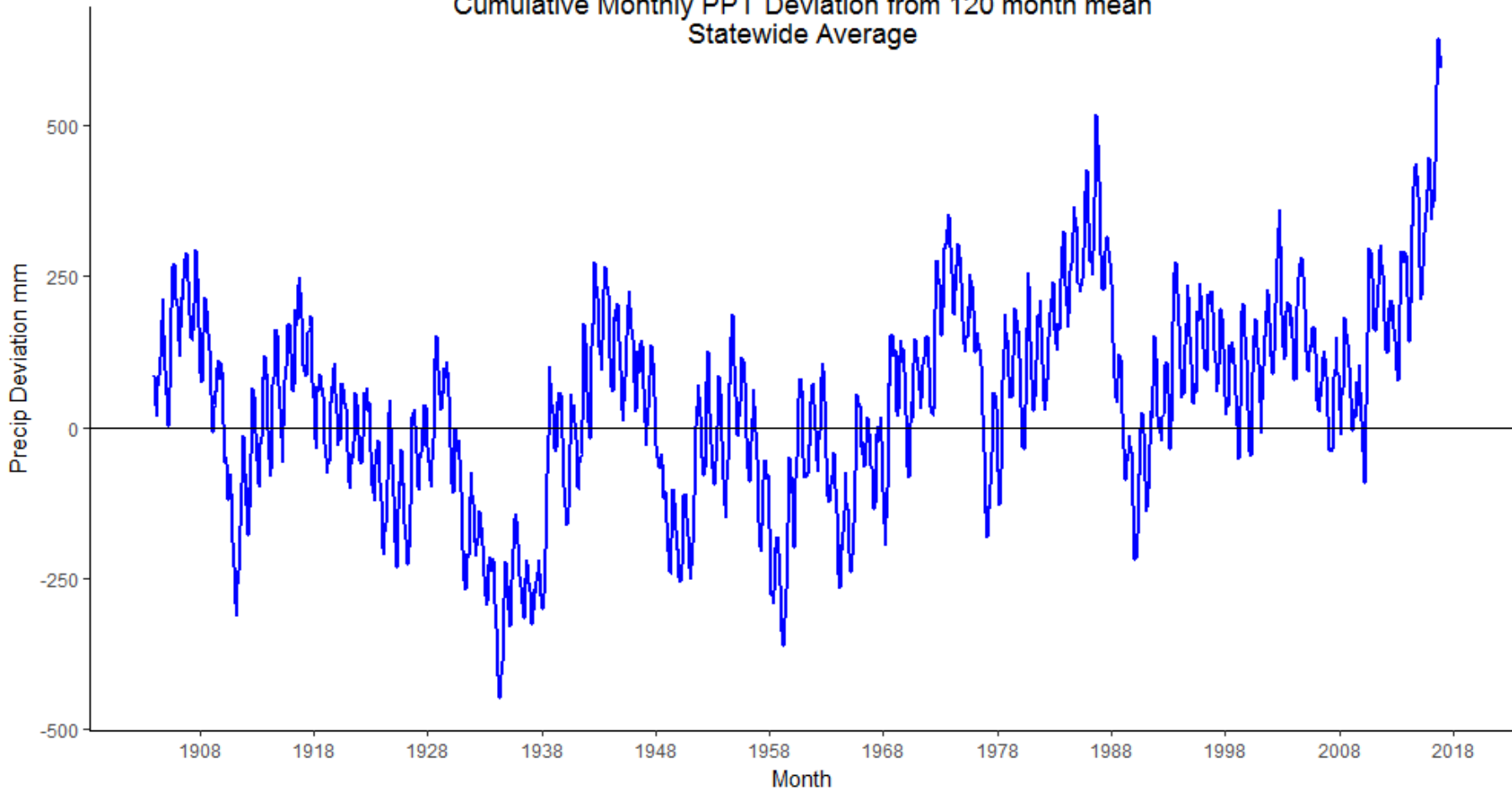


$$\sum_{i=1}^n (ppt_i) - \overline{ppt}_{MA120}$$

Wisconsin Monthly Precipitation Deviation from Monthly Mean Prism Data 1900-2017

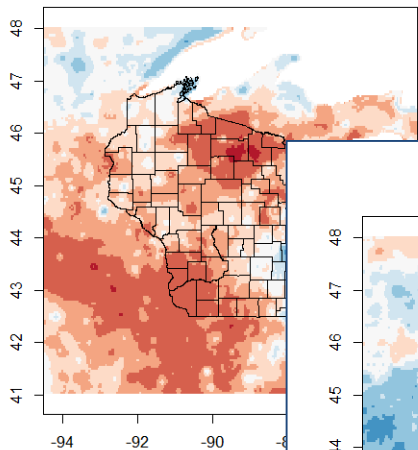


Cumulative Monthly PPT Deviation from 120 month mean
Statewide Average

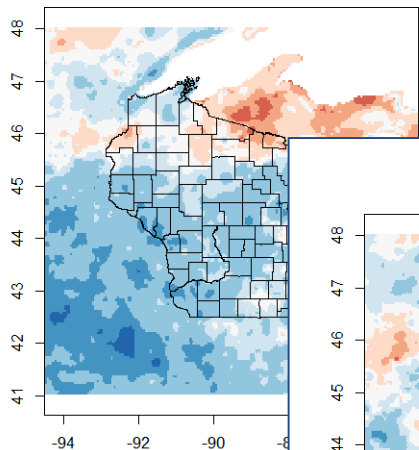


$$\sum_{i=1}^n (ppt_i) - \overline{ppt}_{MA120}$$

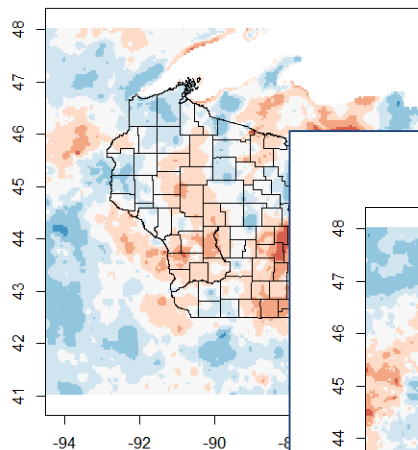
1989-12



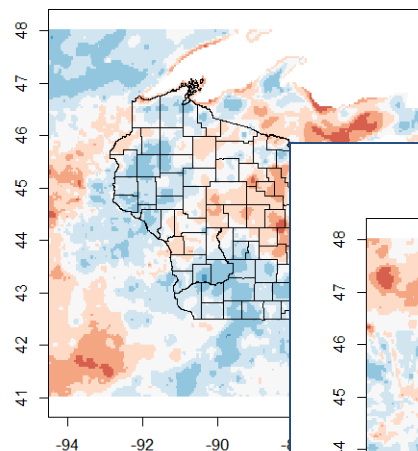
1993-08



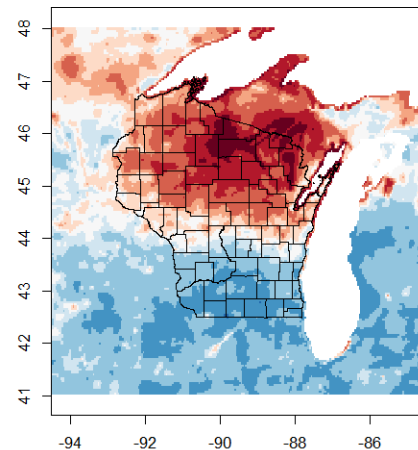
1997-05



2002-05



2010-05

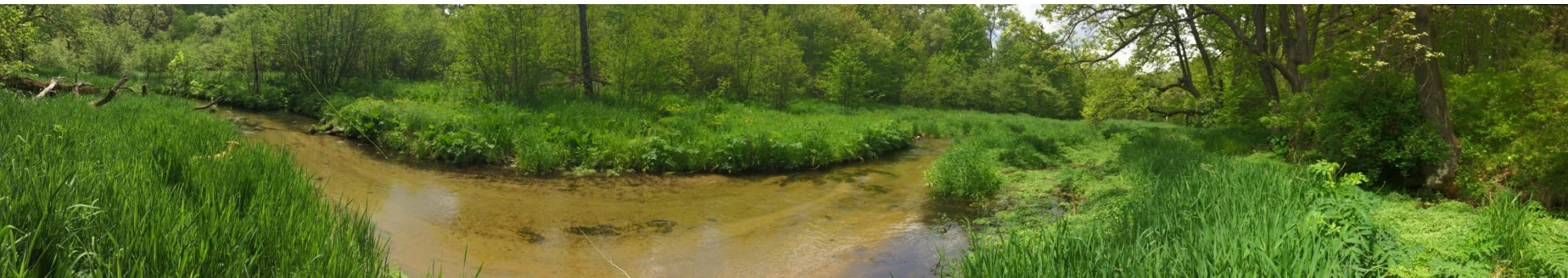


Z-Score

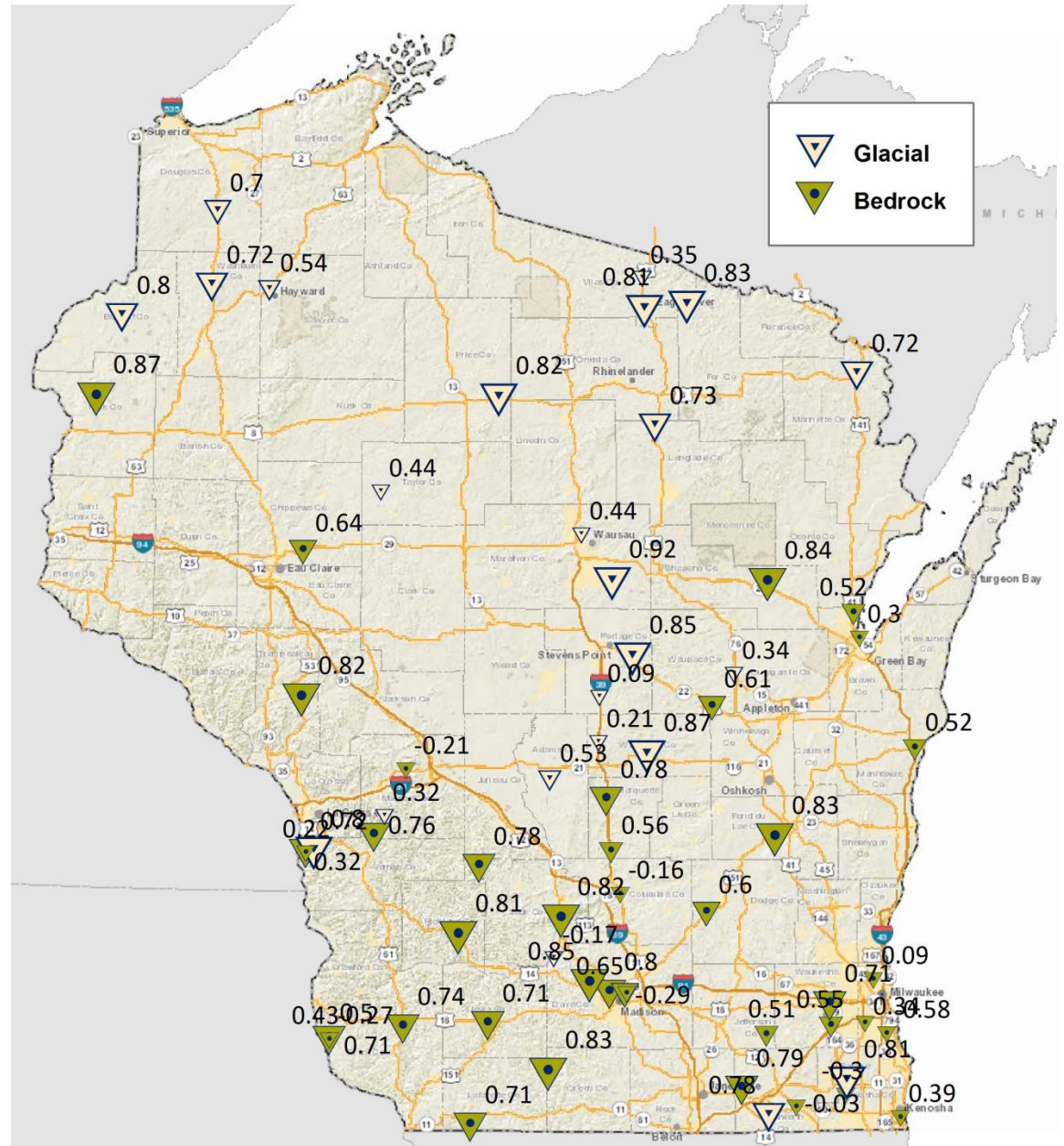


Understanding and Forecasting Variation on Water Levels

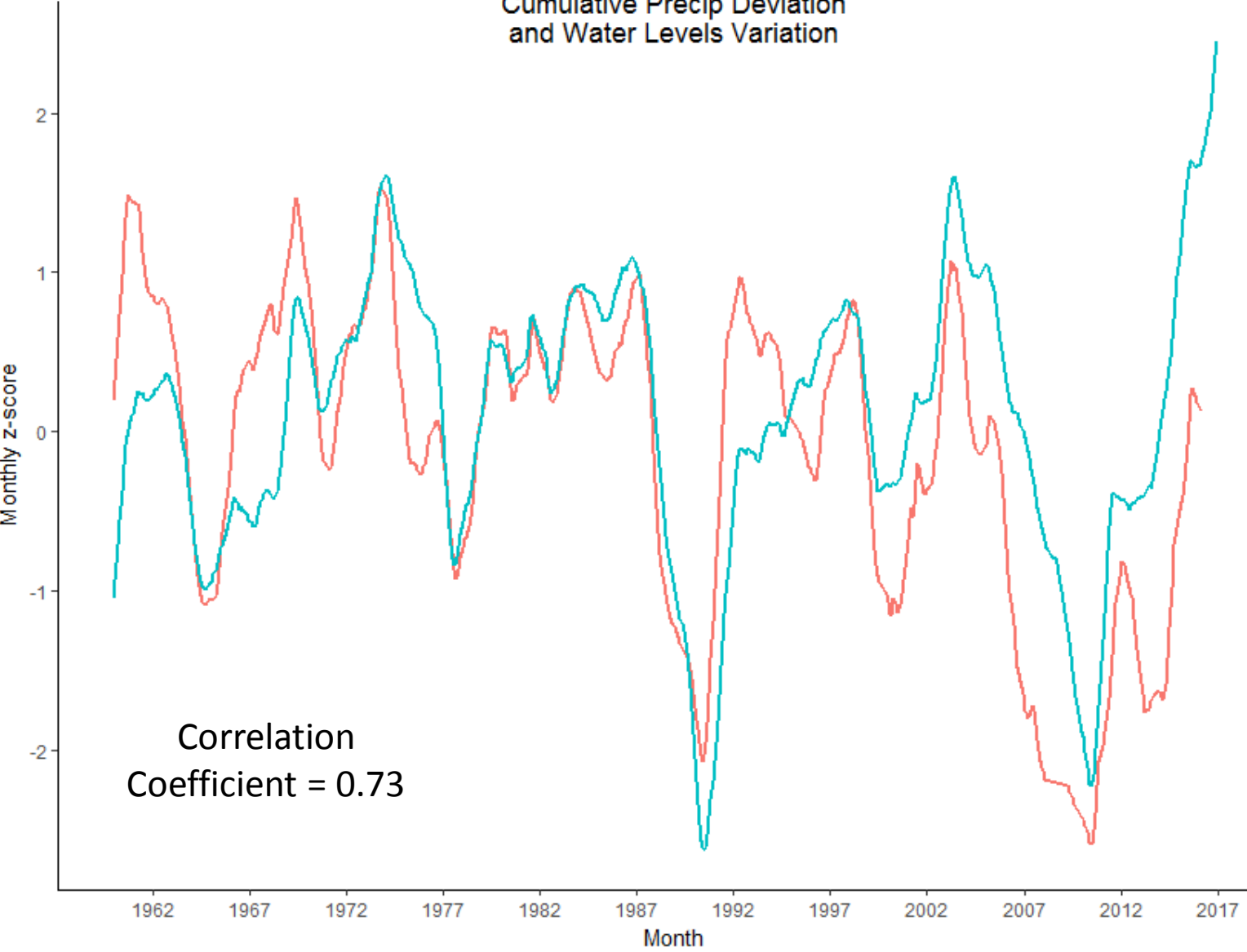
- Interpreting Precipitation
- ✓ Drivers of Water Level variation
- Forecasting Future Water Levels



- 68 USGS Gage Sites with data from 1987 to 2017
- Standardized and smoothed to rolling 12 month average
- Compared to cumulative deviation from mean at location lat/long
- 70% had pearson correlation > 0.50
- 50% had pearson correlation coef $> .70$
- No difference in glacial vs. bedrock formations



1 Year Average Monthly Z Score
Cumulative Precip Deviation
and Water Levels Variation



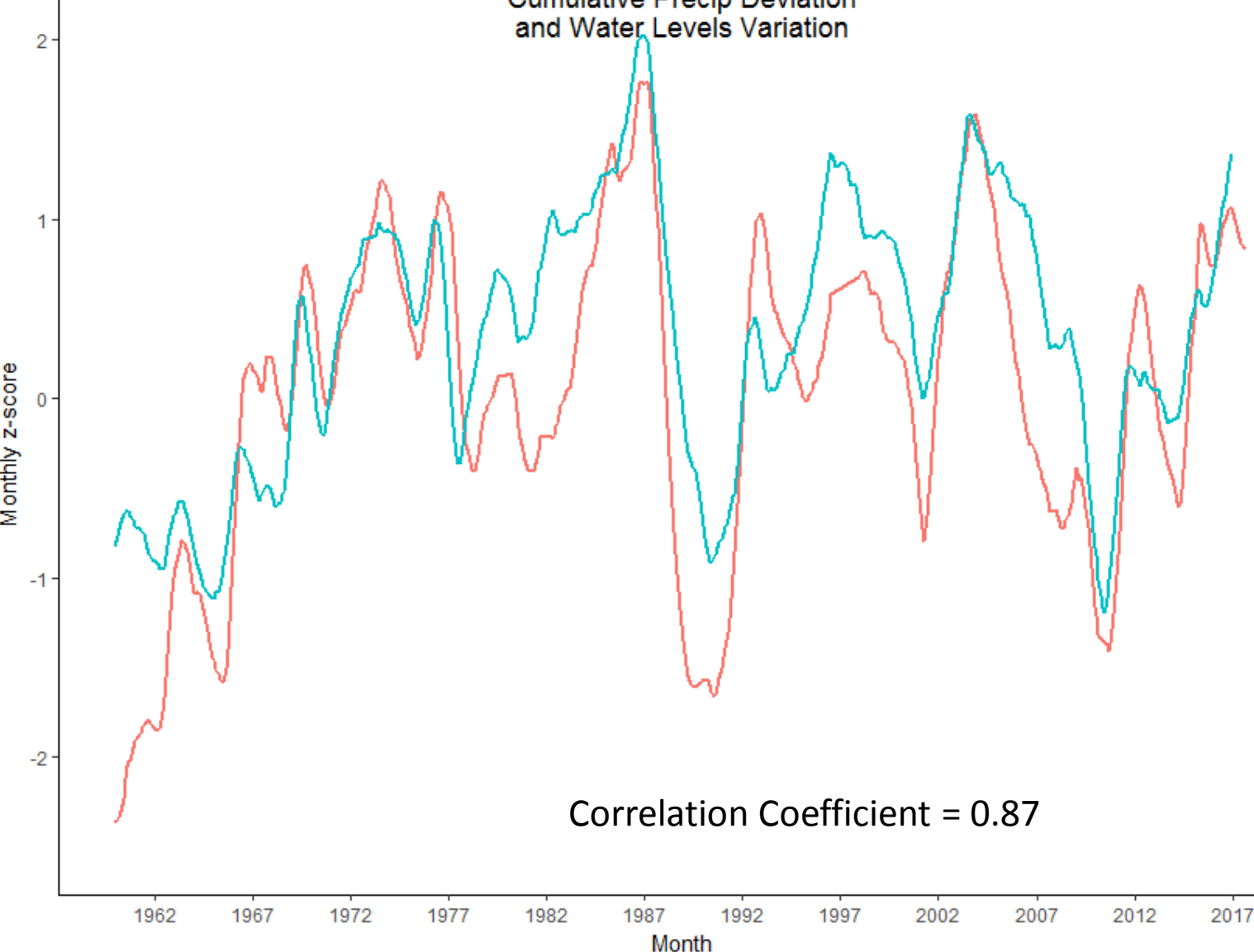
Observation Type

- 453720089215401
- Precip



Oneida County

1 Year Average Monthly Z Score
Cumulative Precip Deviation
and Water Levels Variation



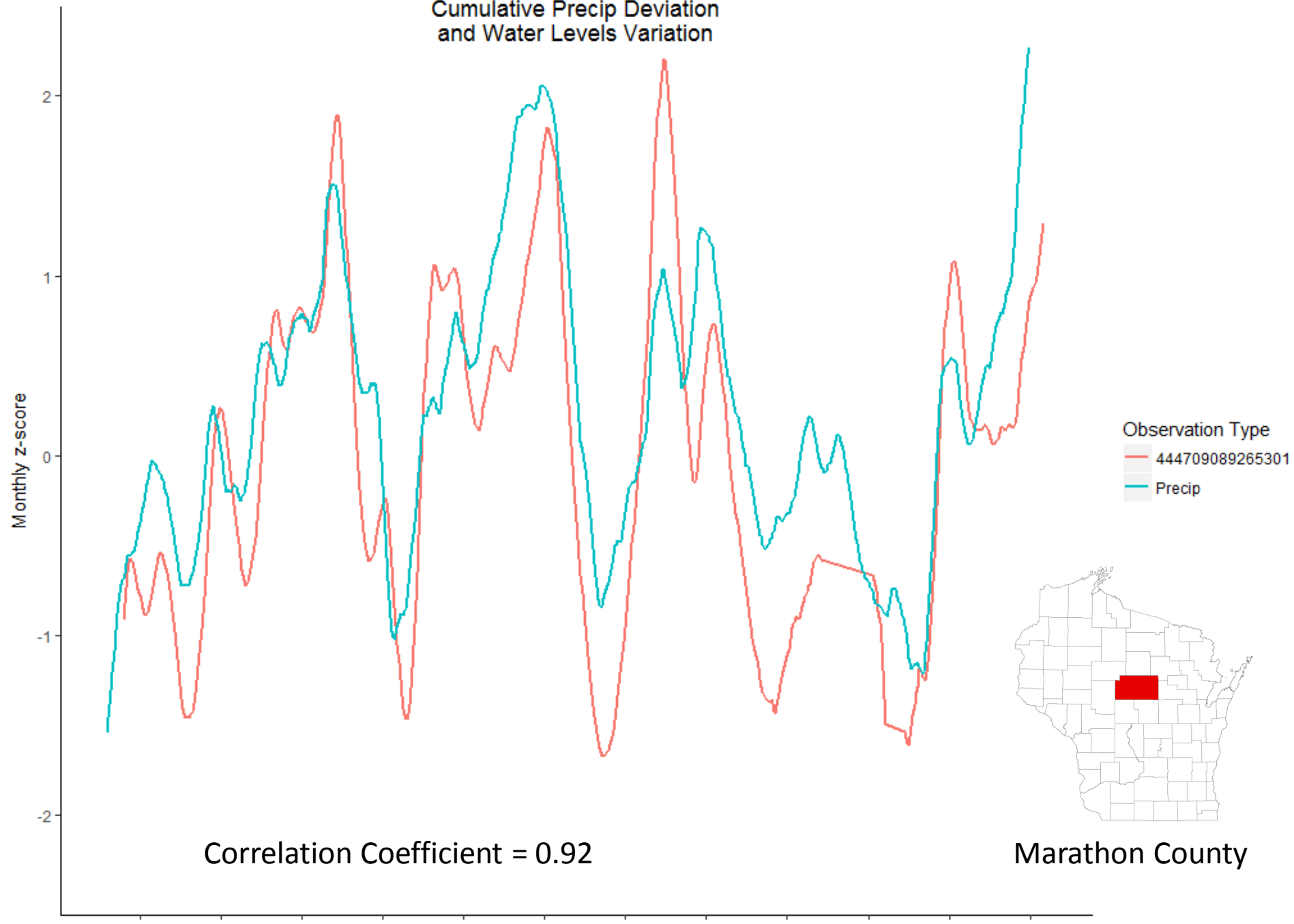
Observation Type

- 453013092314601
- Precip

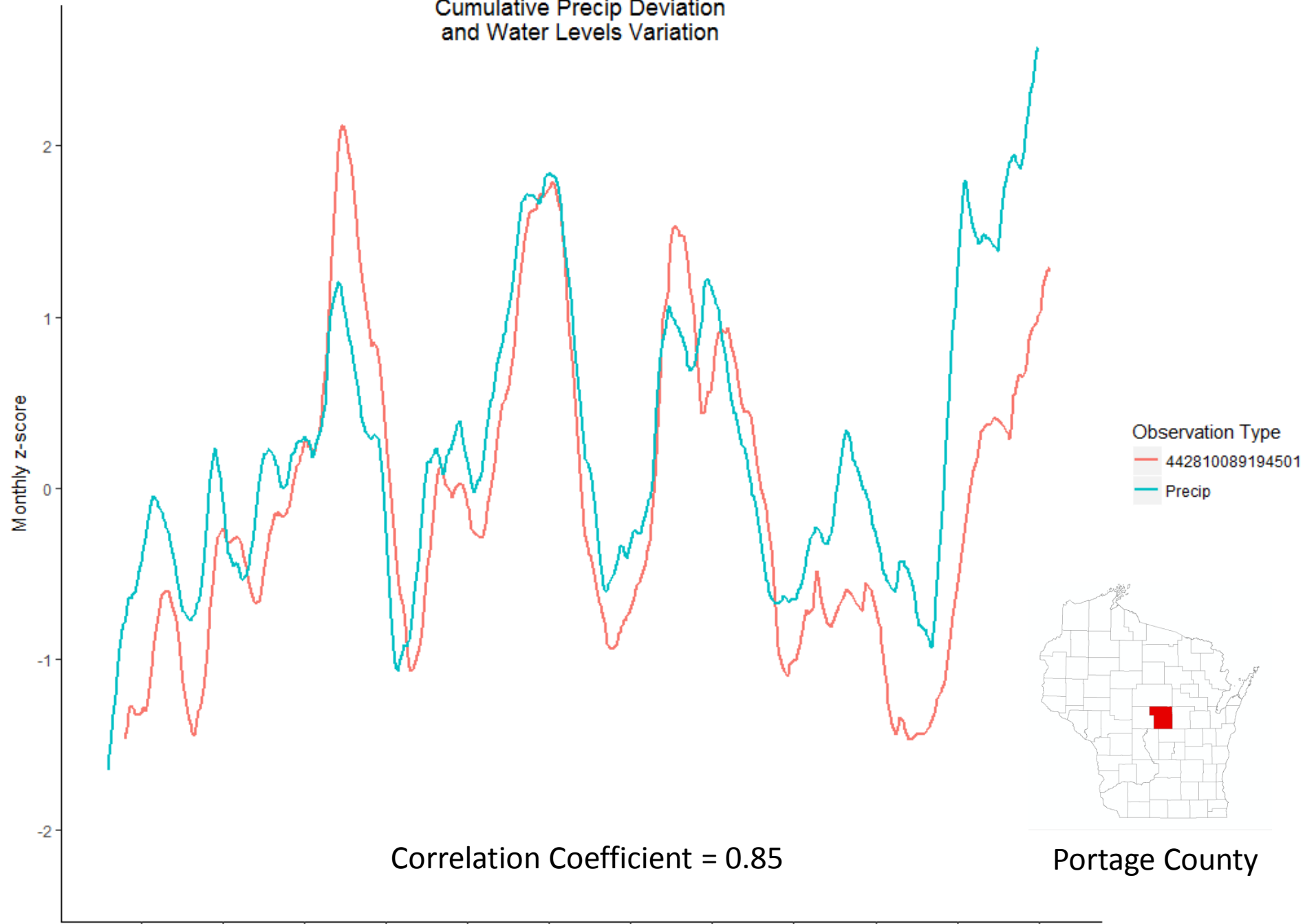


Polk County

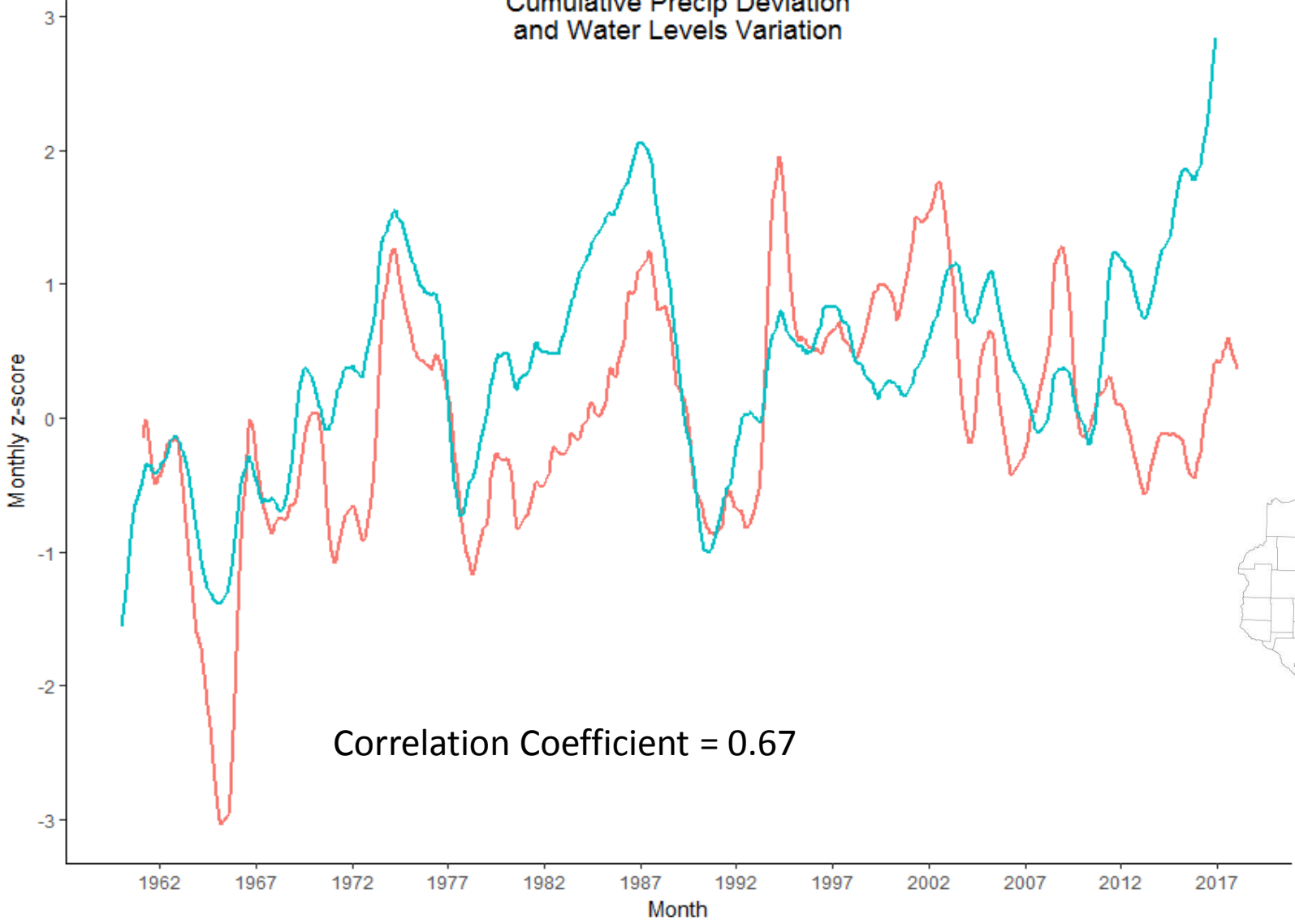
1 Year Average Monthly Z Score Cumulative Precip Deviation and Water Levels Variation



1 Year Average Monthly Z Score Cumulative Precip Deviation and Water Levels Variation



1 Year Average Monthly Z Score
Cumulative Precip Deviation
and Water Levels Variation



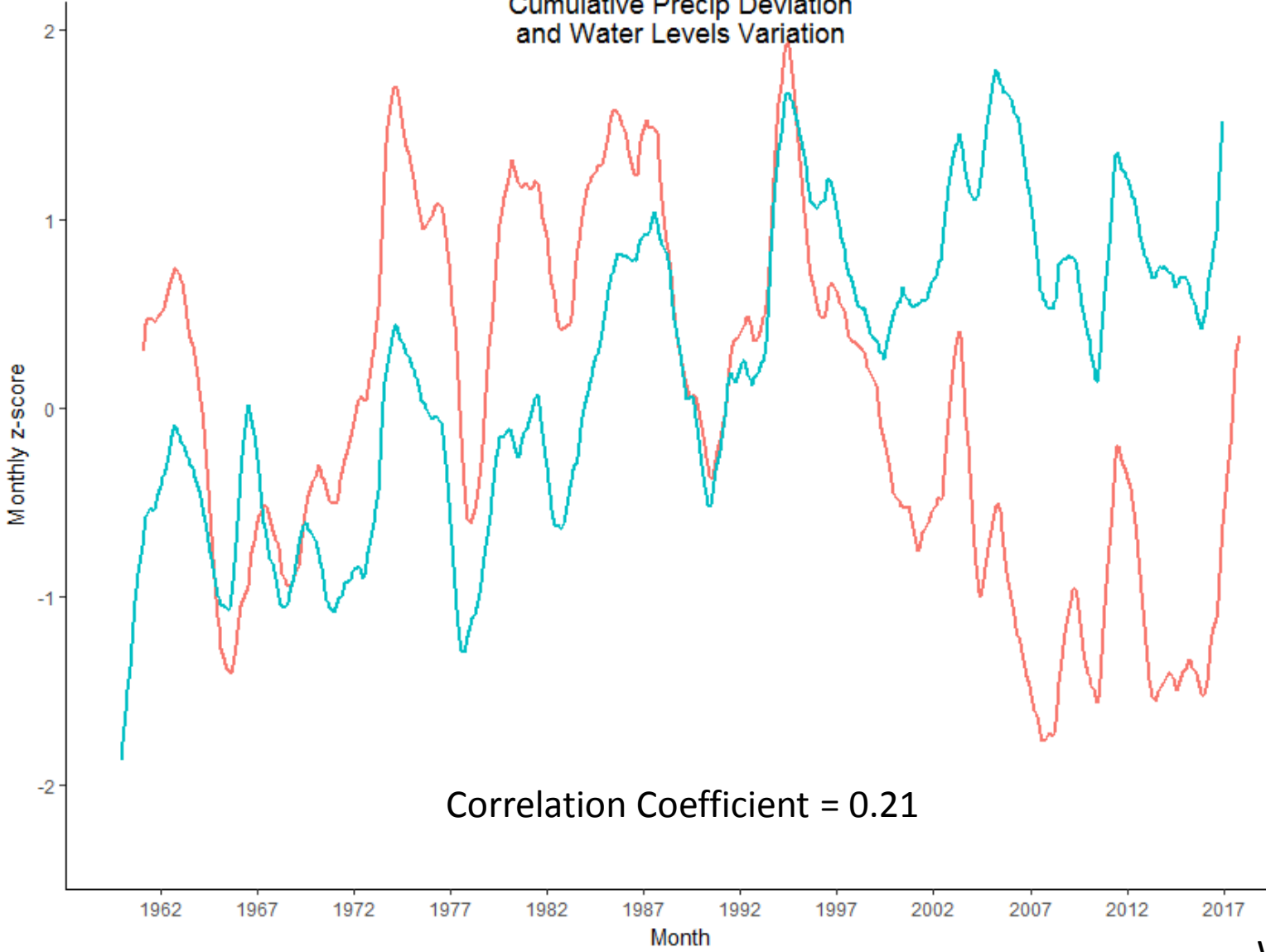
Correlation Coefficient = 0.67

Observation Type
05404500
Basin



Devils Lake
Sauk County

1 Year Average Monthly Z Score
Cumulative Precip Deviation
and Water Levels Variation



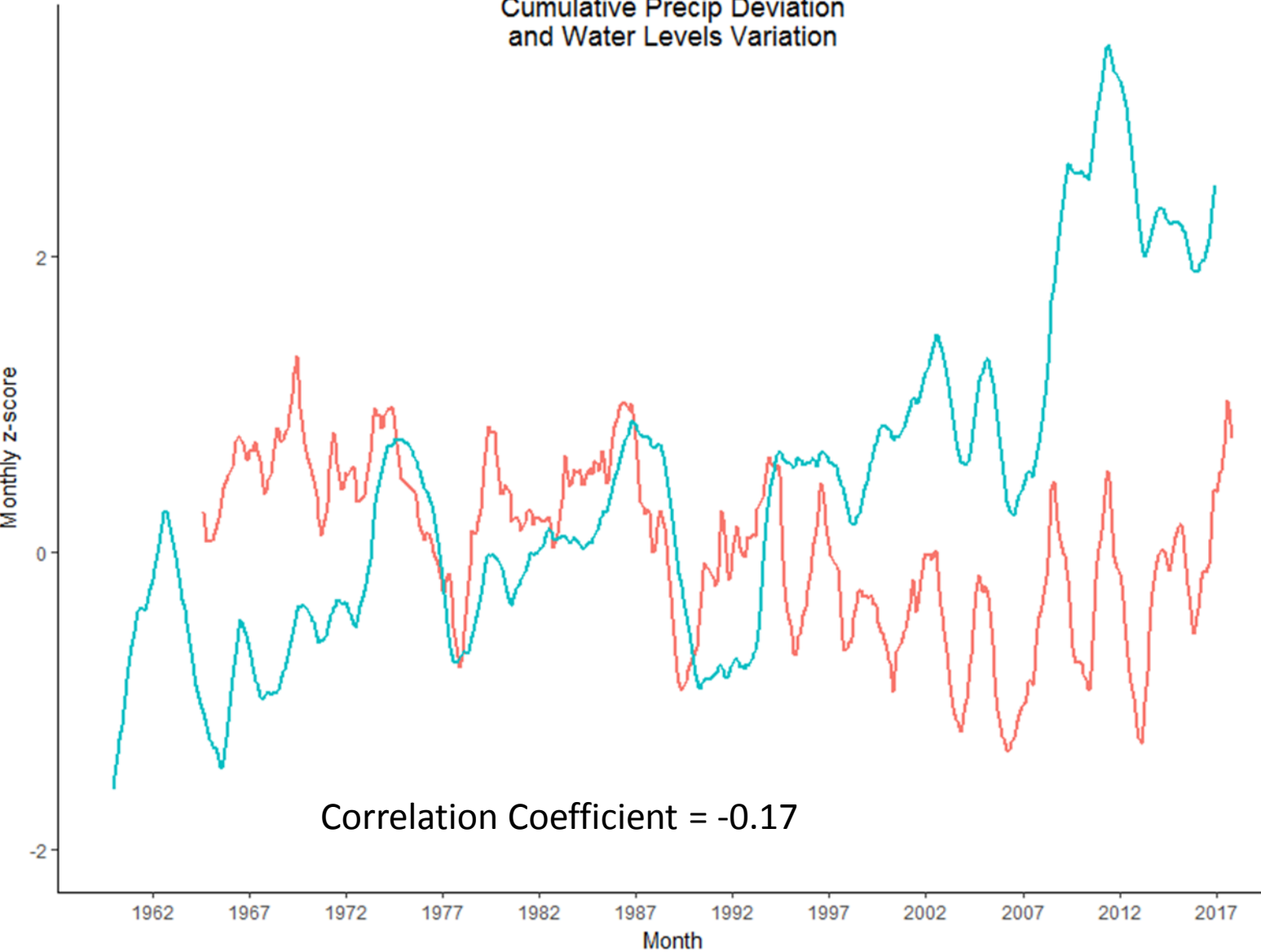
Correlation Coefficient = 0.21

Observation Type
440713089320801
Precip



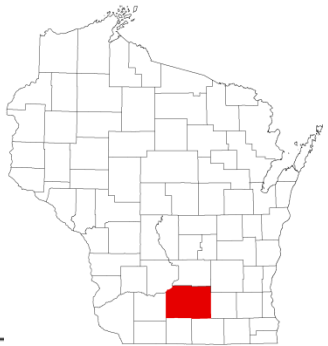
Waushara County

1 Year Average Monthly Z Score
Cumulative Precip Deviation
and Water Levels Variation



Observation Type

- 431312089475301
- Precip

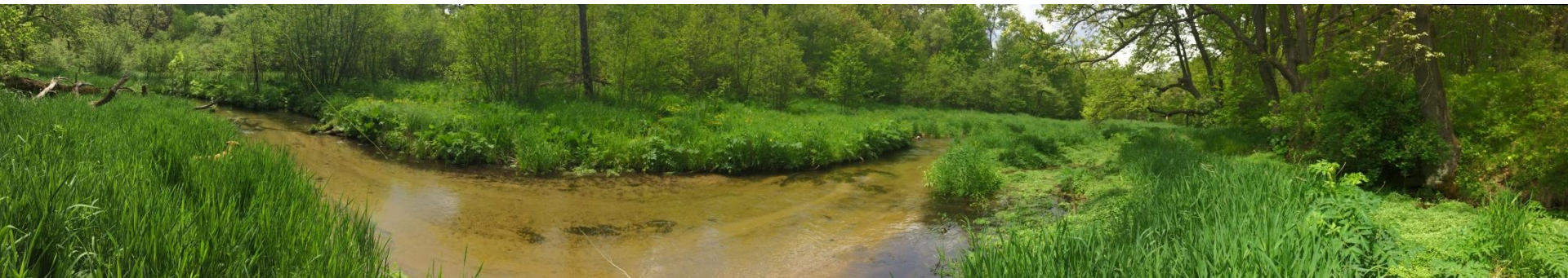


Dane County



Understanding and Forecasting Variation on Water Levels

- Interpreting Precipitation
- Drivers of Water Level variation
- Forecasting Future Water Levels

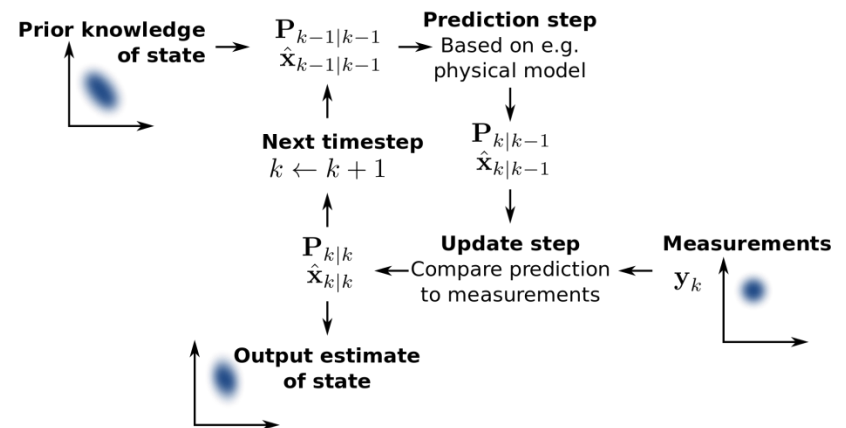
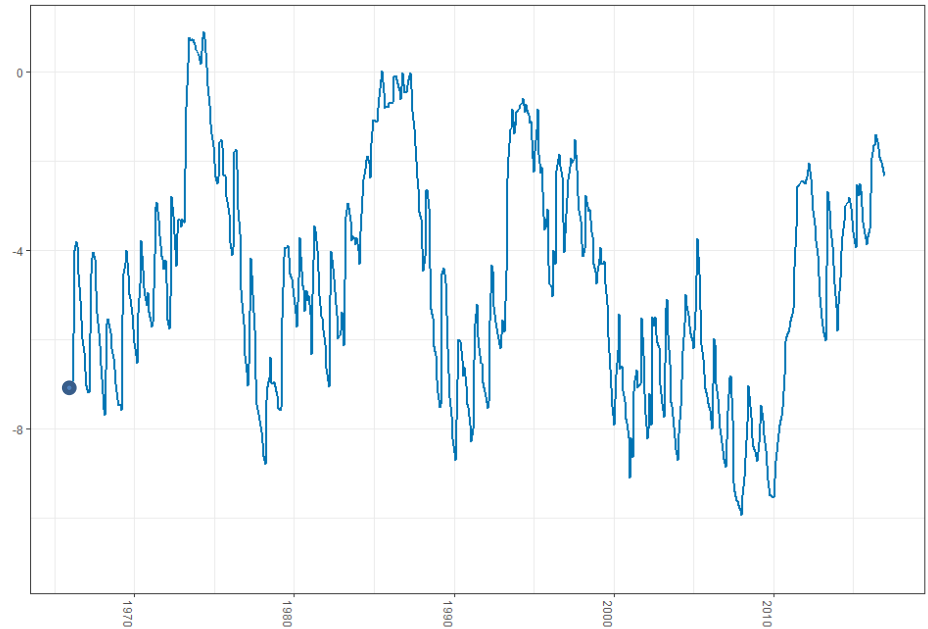




Understanding and Forecasting Variation on Water Levels

Bayesian Structural Time Series

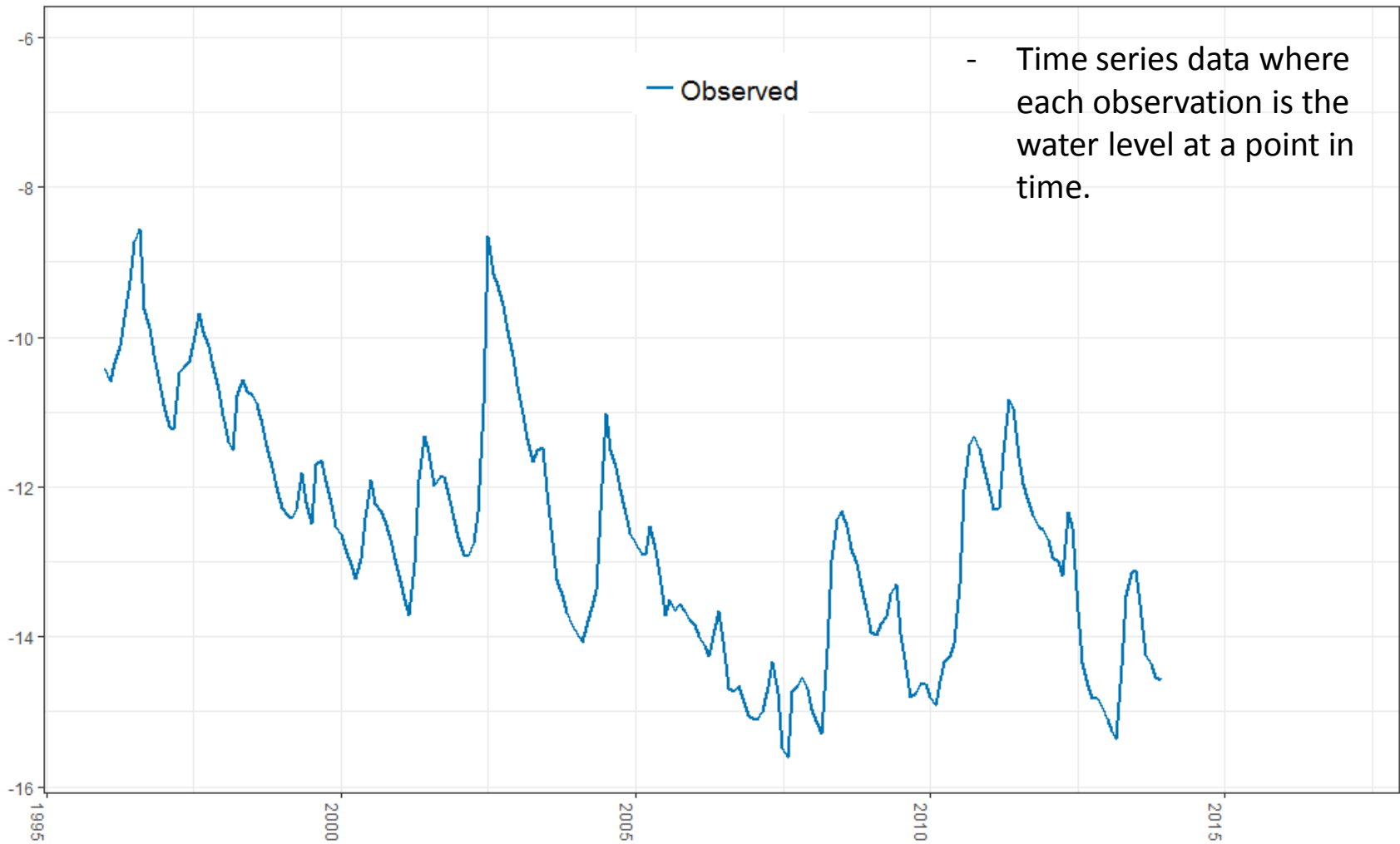
- Machine learning technique for time series data forecasting
- Models the probability of the next step based on past likelihoods
- Incorporates seasonal cycle
- Provides likelihood that that prediction will fit the observation
- Predictor data can be added to substantially improve the model.
 - Precip deviations
 - Monthly Precip
 - Evapotranspiration
 - Withdrawals
 - Global weather indices





Understanding and Forecasting Variation on Water Levels

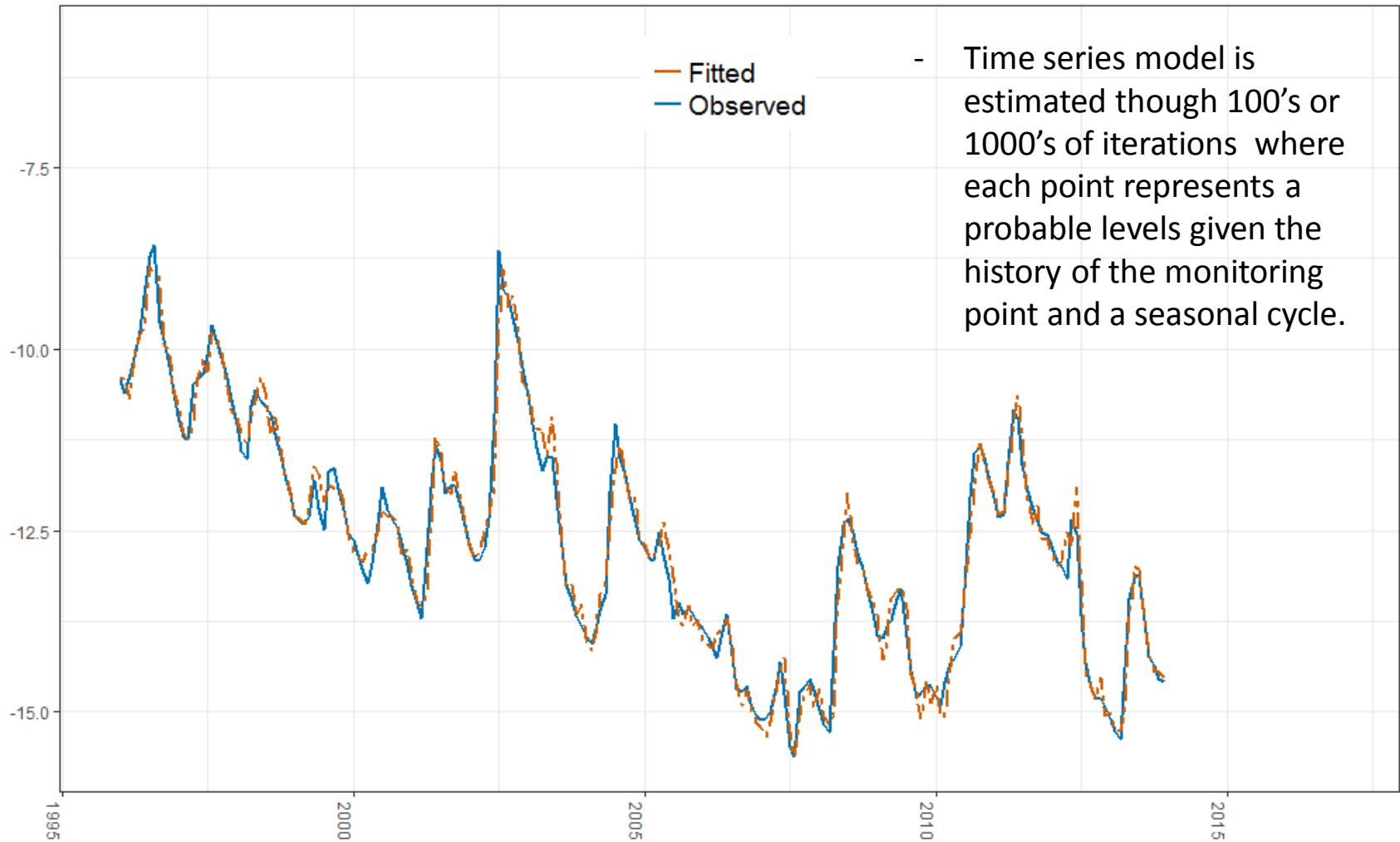
Forecasted and Observed Water Levels at WS-19/08E/15-0008





Understanding and Forecasting Variation on Water Levels

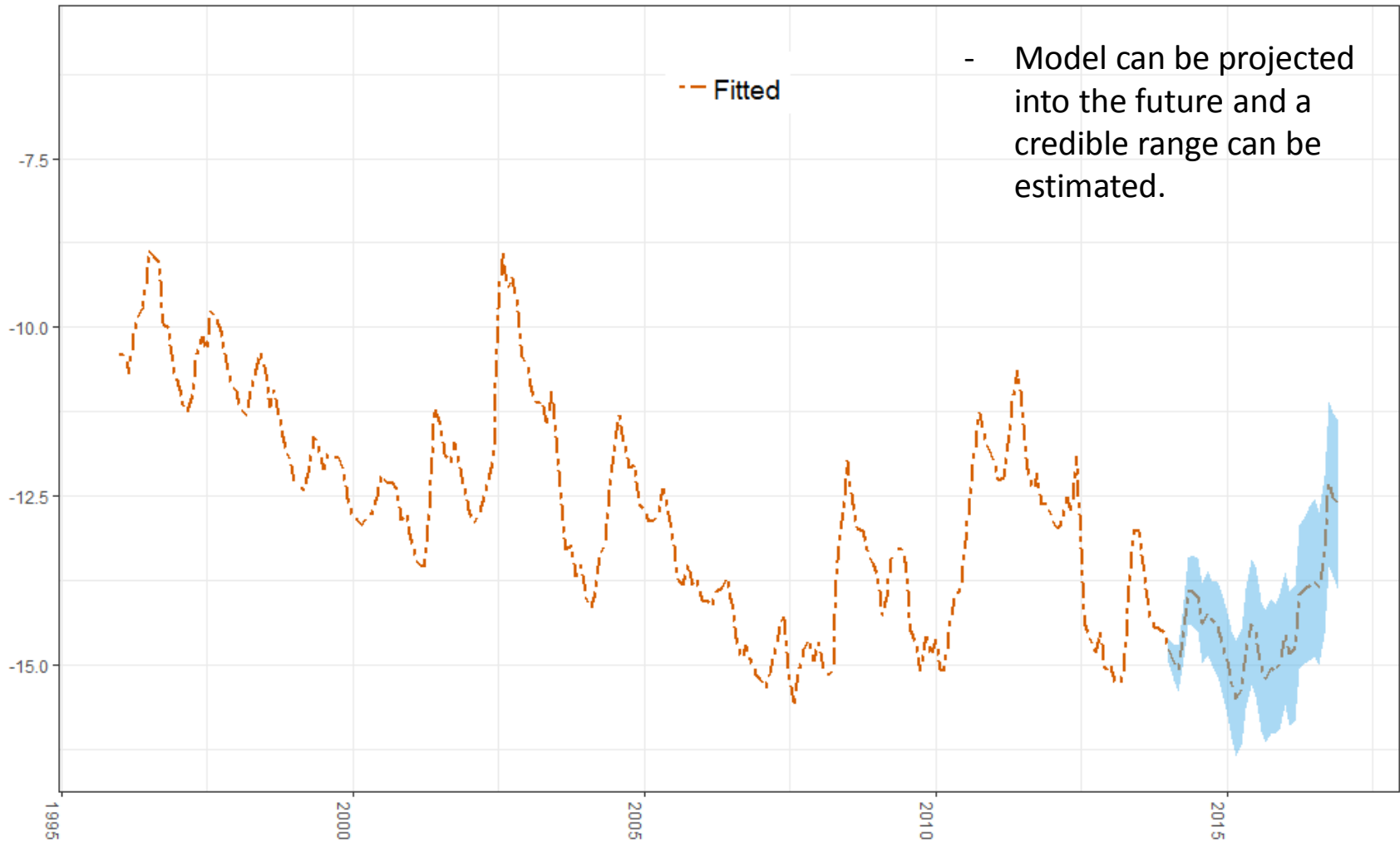
Forecasted and Observed Water Levels at WS-19/08E/15-0008





Understanding and Forecasting Variation on Water Levels

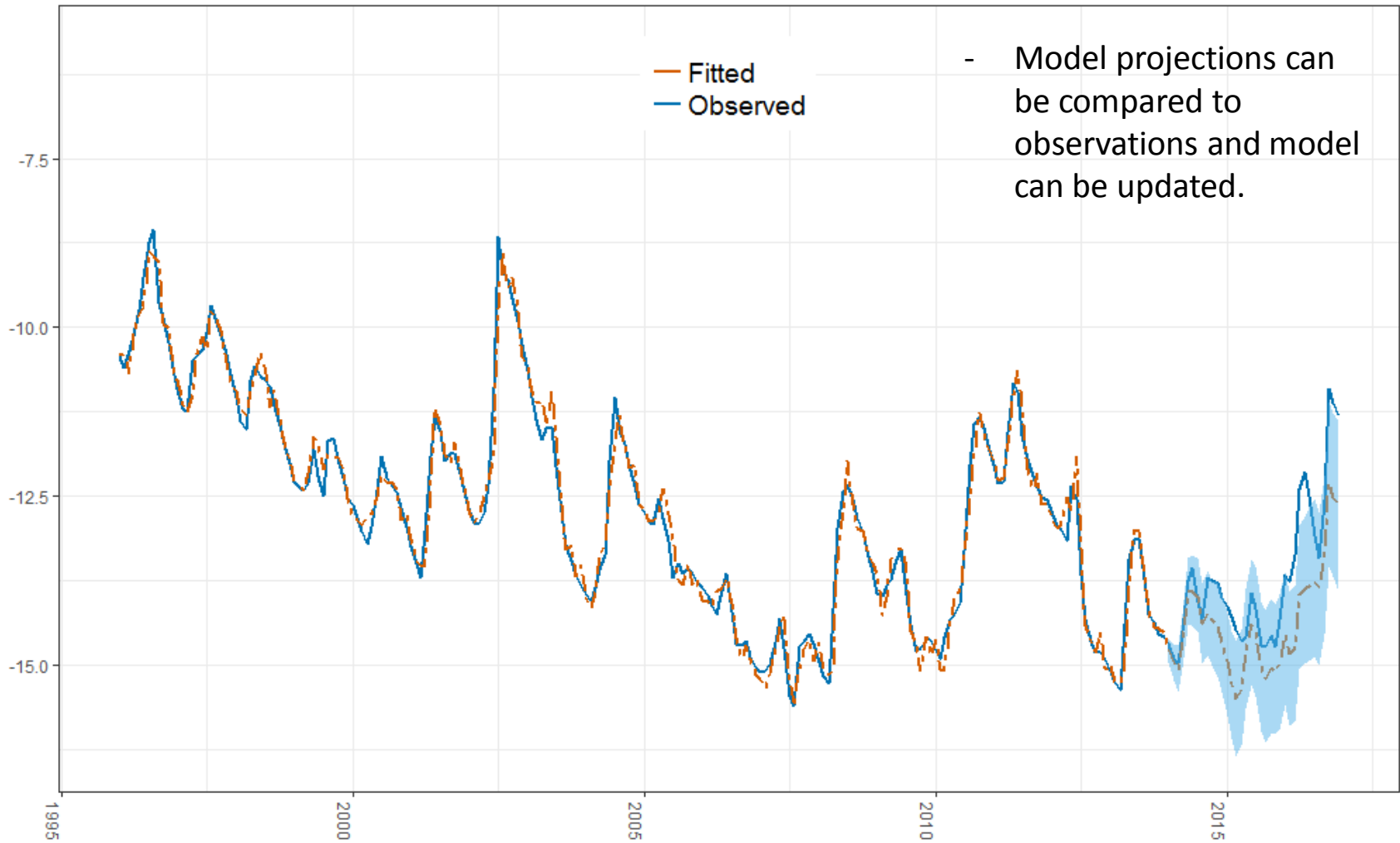
Forecasted and Observed Water Levels at WS-19/08E/15-0008





Understanding and Forecasting Variation on Water Levels

Forecasted and Observed Water Levels at WS-19/08E/15-0008





Understanding and Forecasting Variation on Water Levels

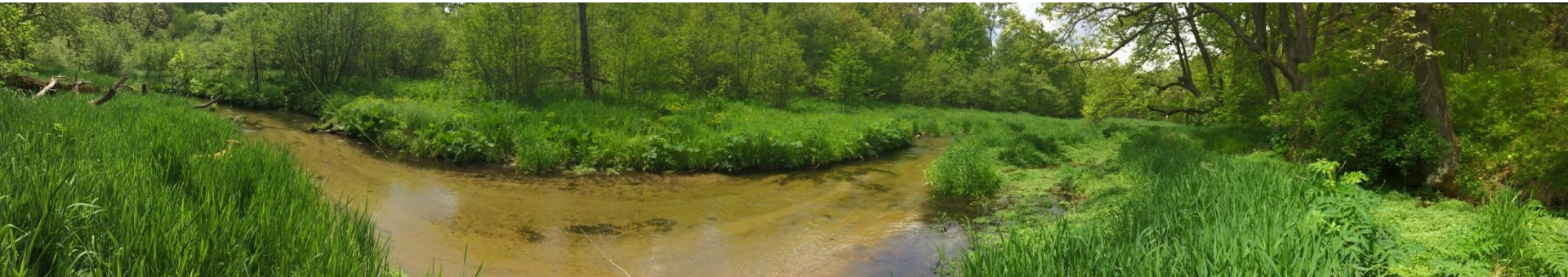
Thank You

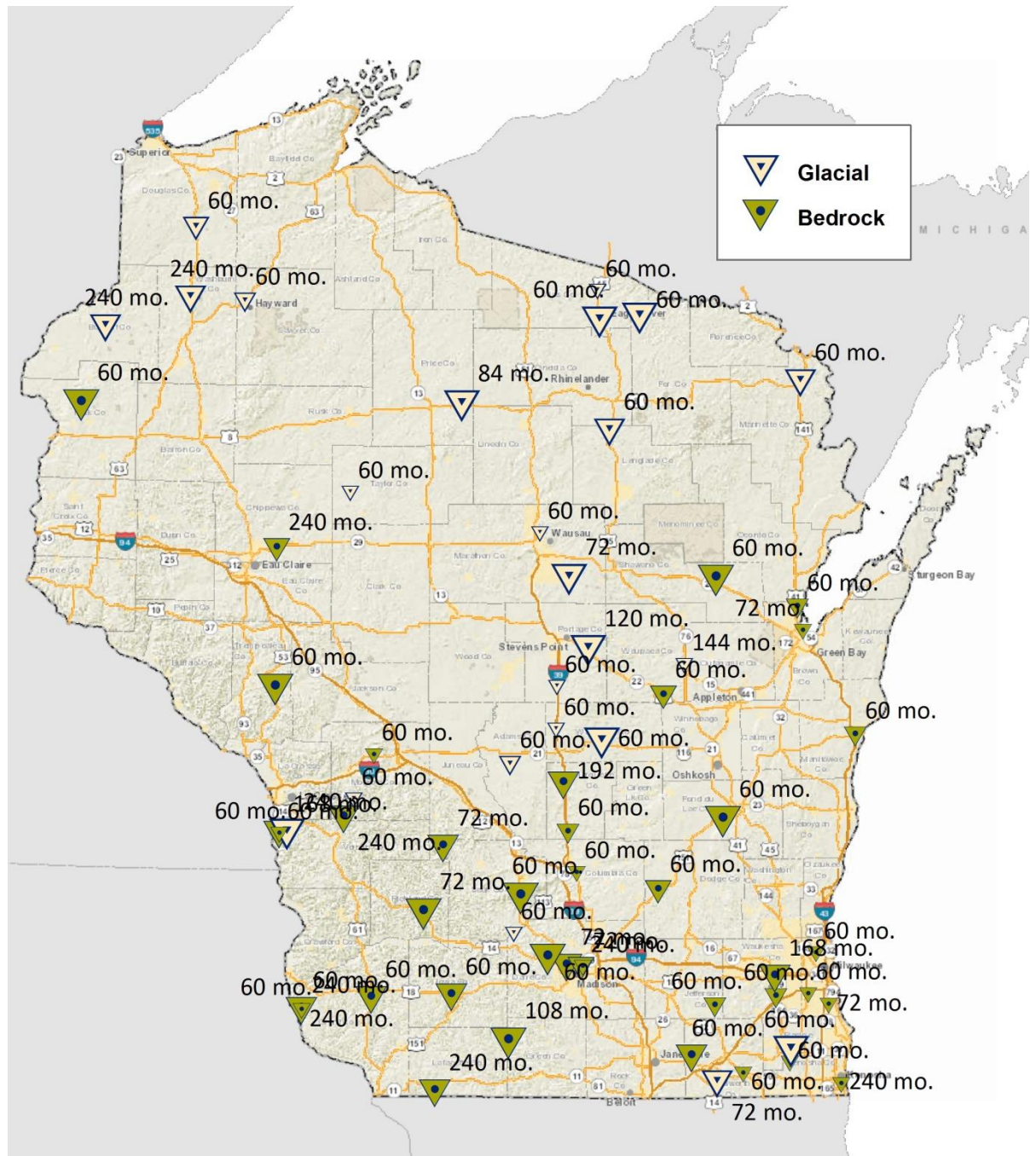
For more information, contact

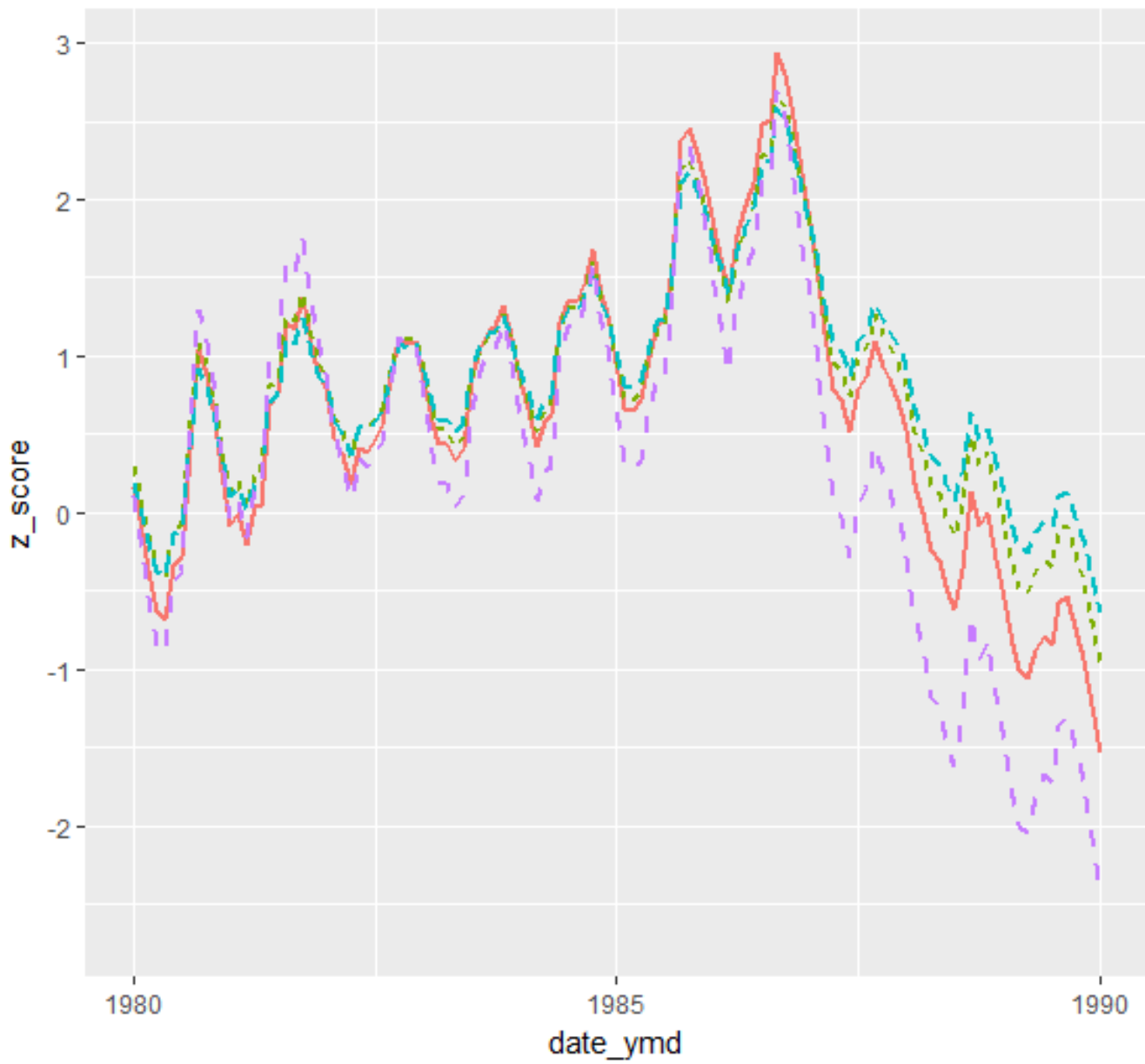
Bob Smail

608-267-3581

Robert.smail@wisconsin.gov







- obs_type
- rollmean_120_dev_cmdv
 - rollmean_180_dev_cmdv
 - rollmean_240_dev_cmdv
 - rollmean_60_dev_cmdv