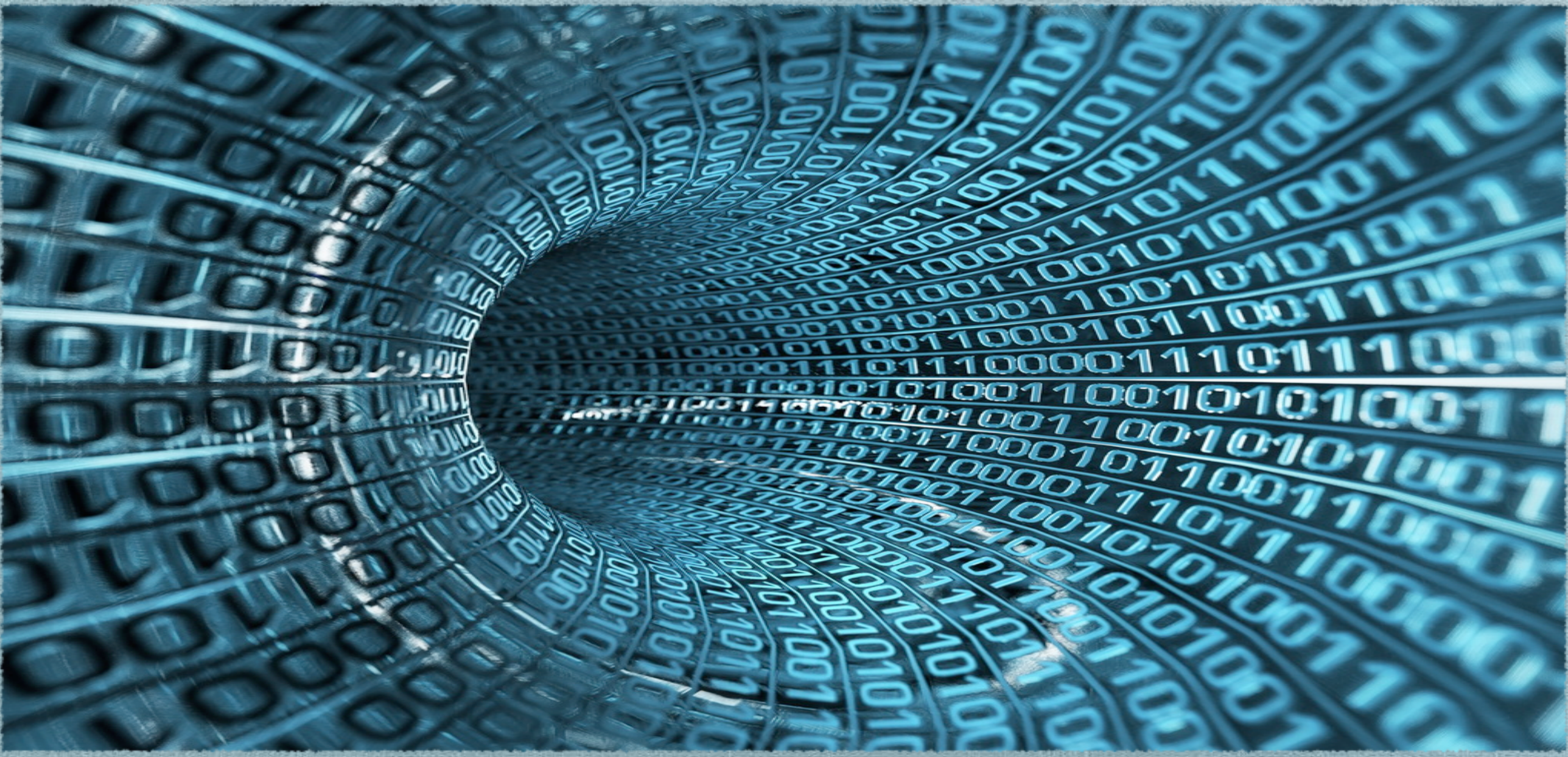
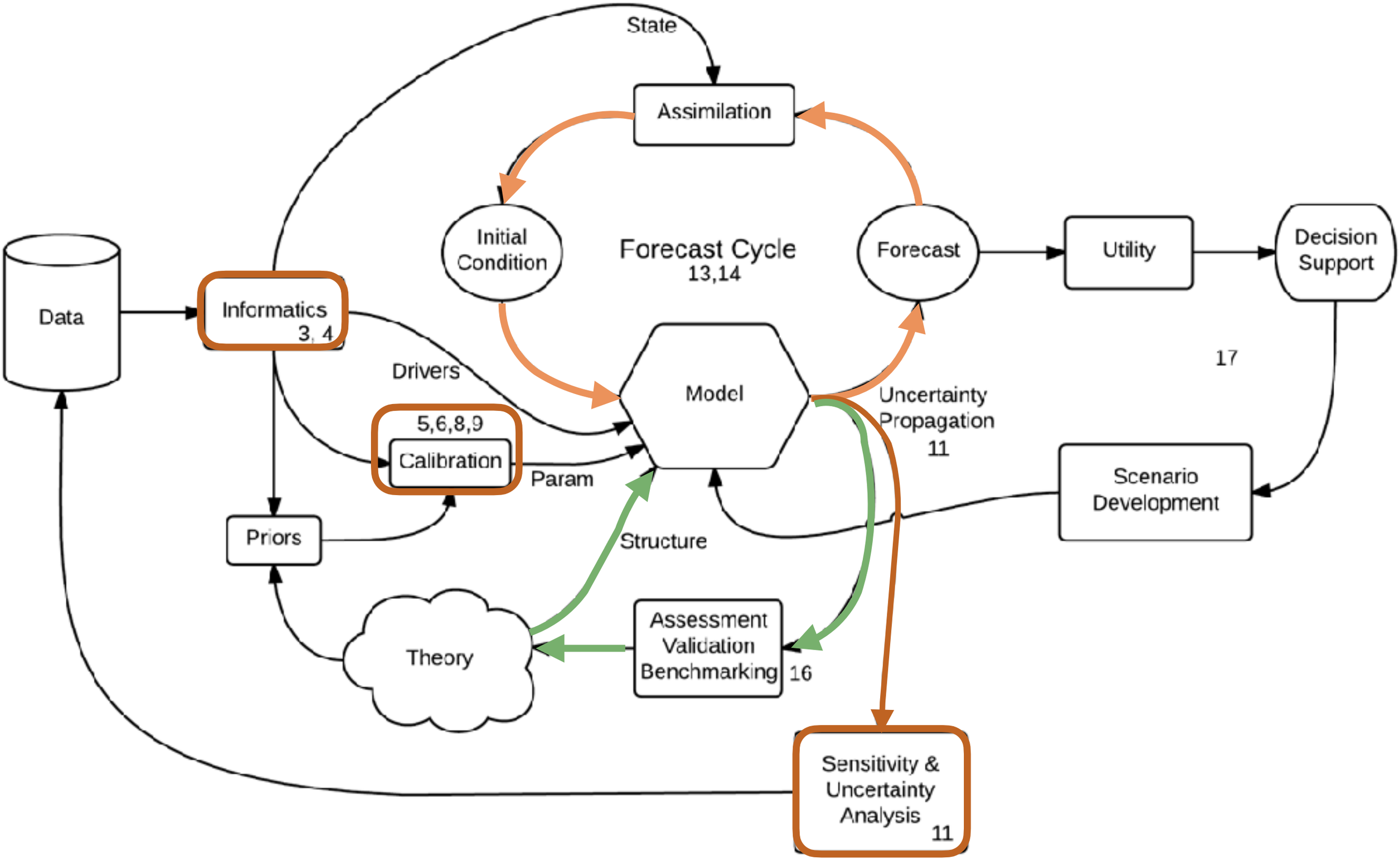


Assessing Model Performance

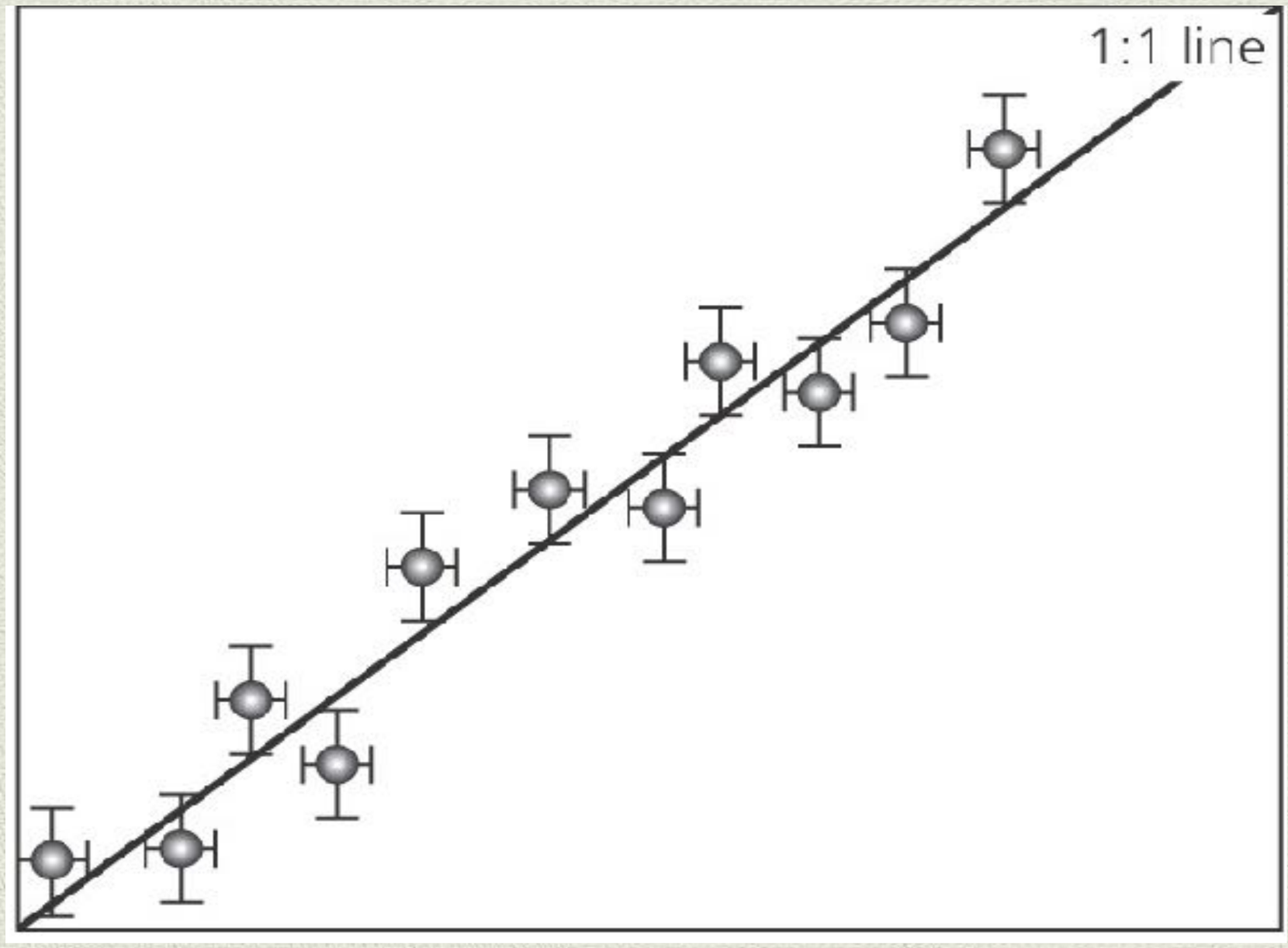

State
Assimilation
Initial Condition
Forecast Cycle
13,14
Forecast
Utility
Decision Support
Data
Informatics
3, 4
Drivers
Model
Uncertainty
Propagation
11
17
5,6,8,9
Calibration
Param
Scenario Development
Priors
Structure
Theory
Assessment
Validation
Benchmarking
16
Sensitivity &
Uncertainty
Analysis
11

- Range of values
- Units
- General pattern in time & space

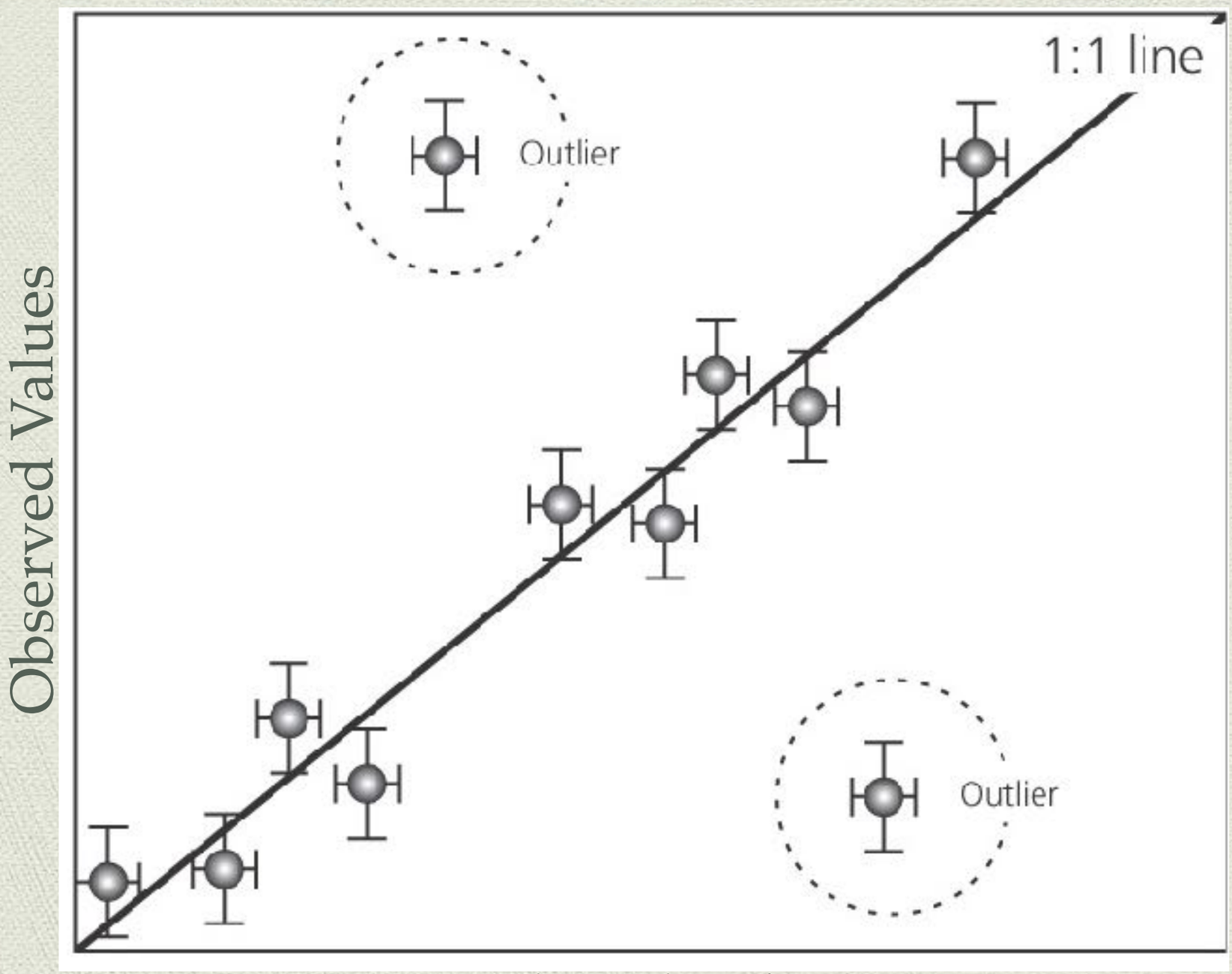
Step 1: Is the model output reasonable?

Step 2: Graphical comparisons to data

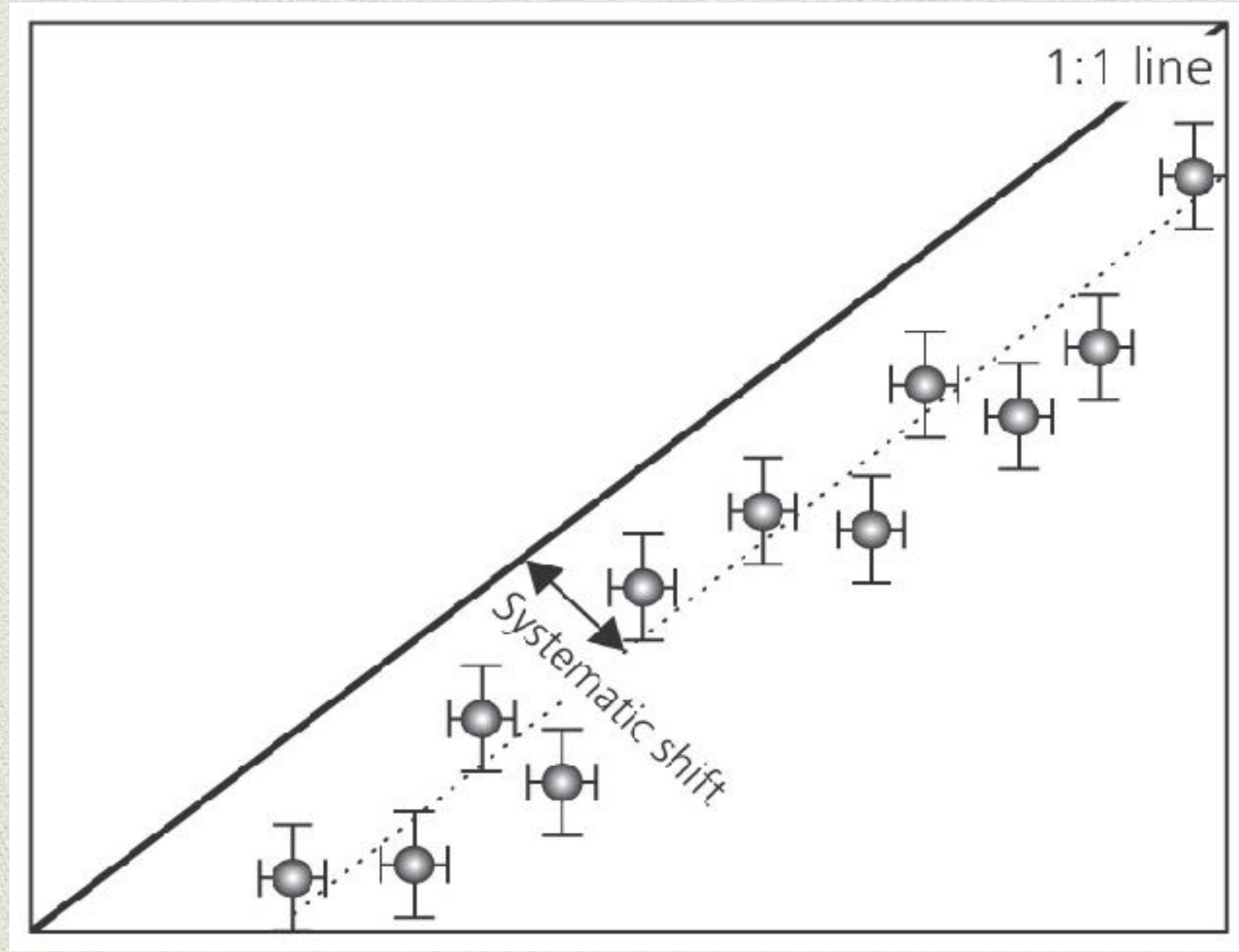
Accuracy of Prediction



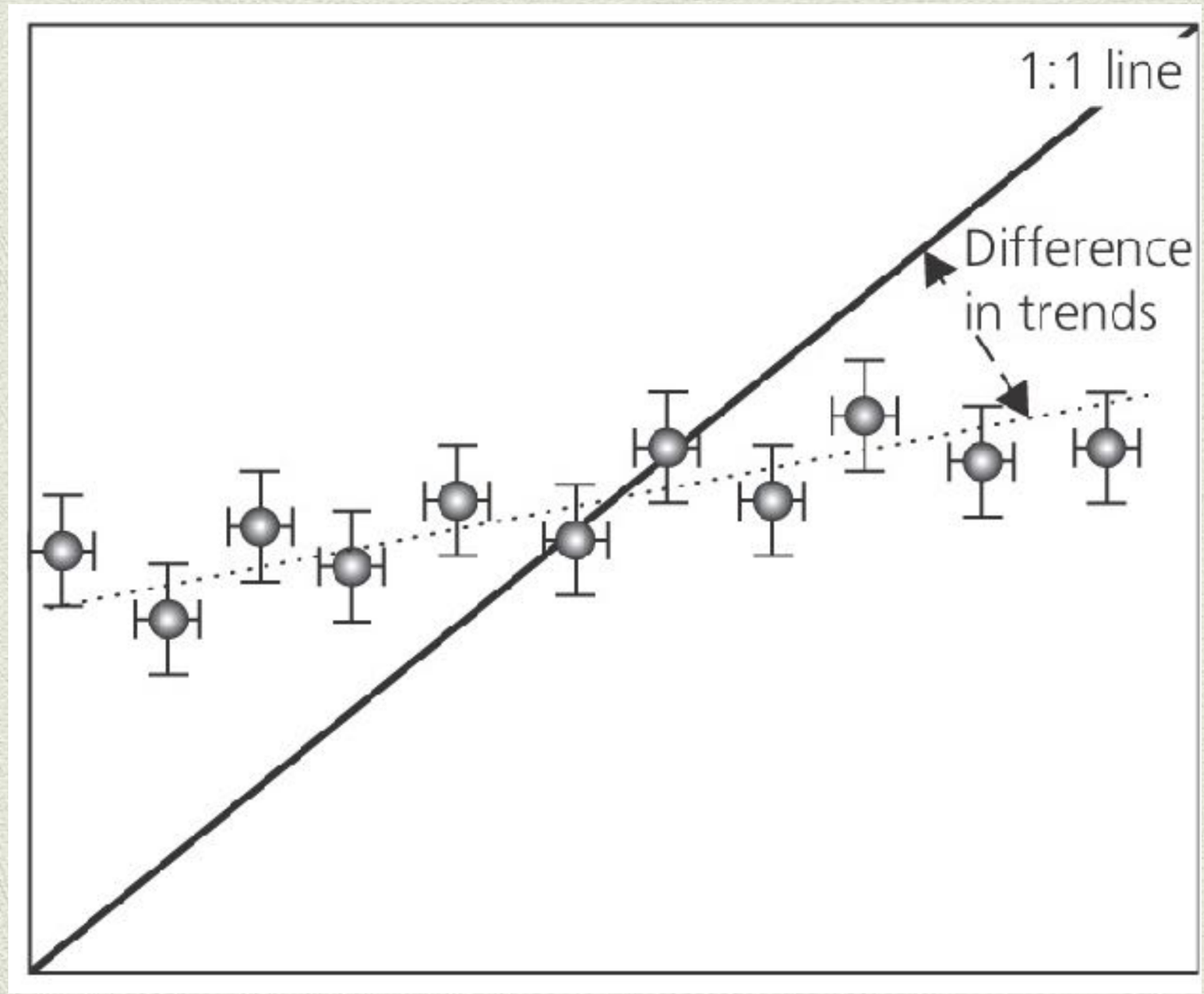
Identify Outliers



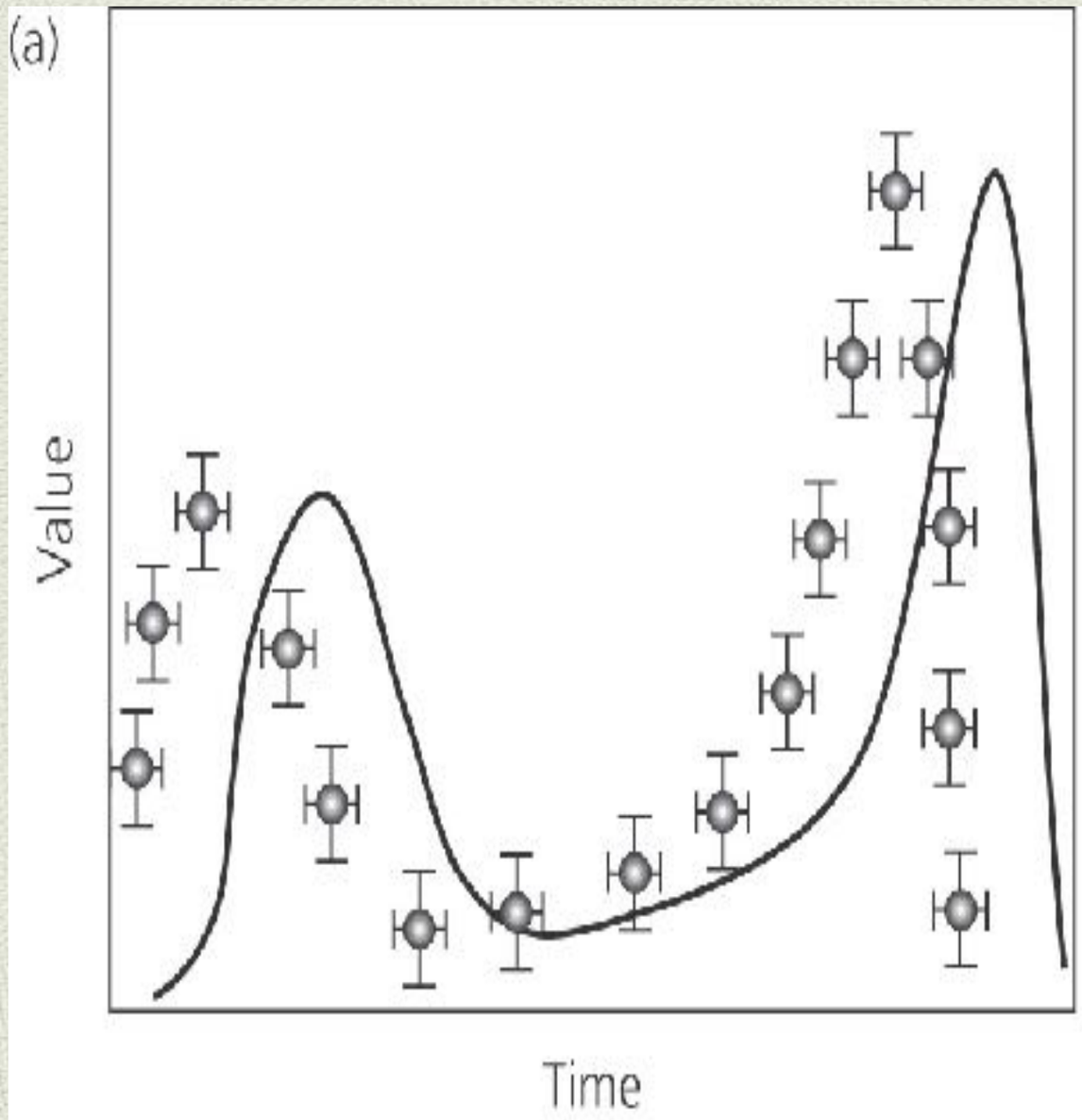
Assess Biases

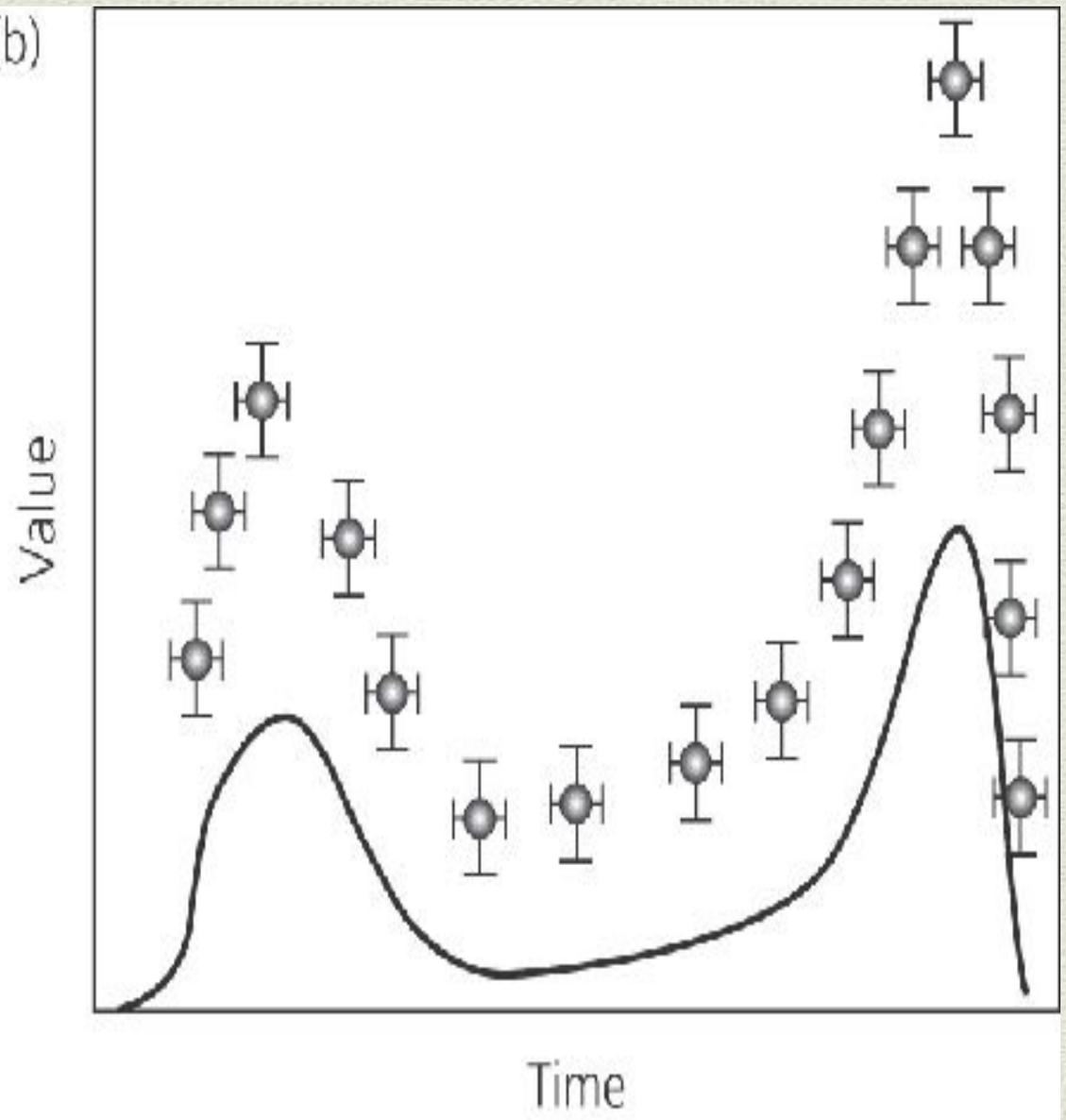


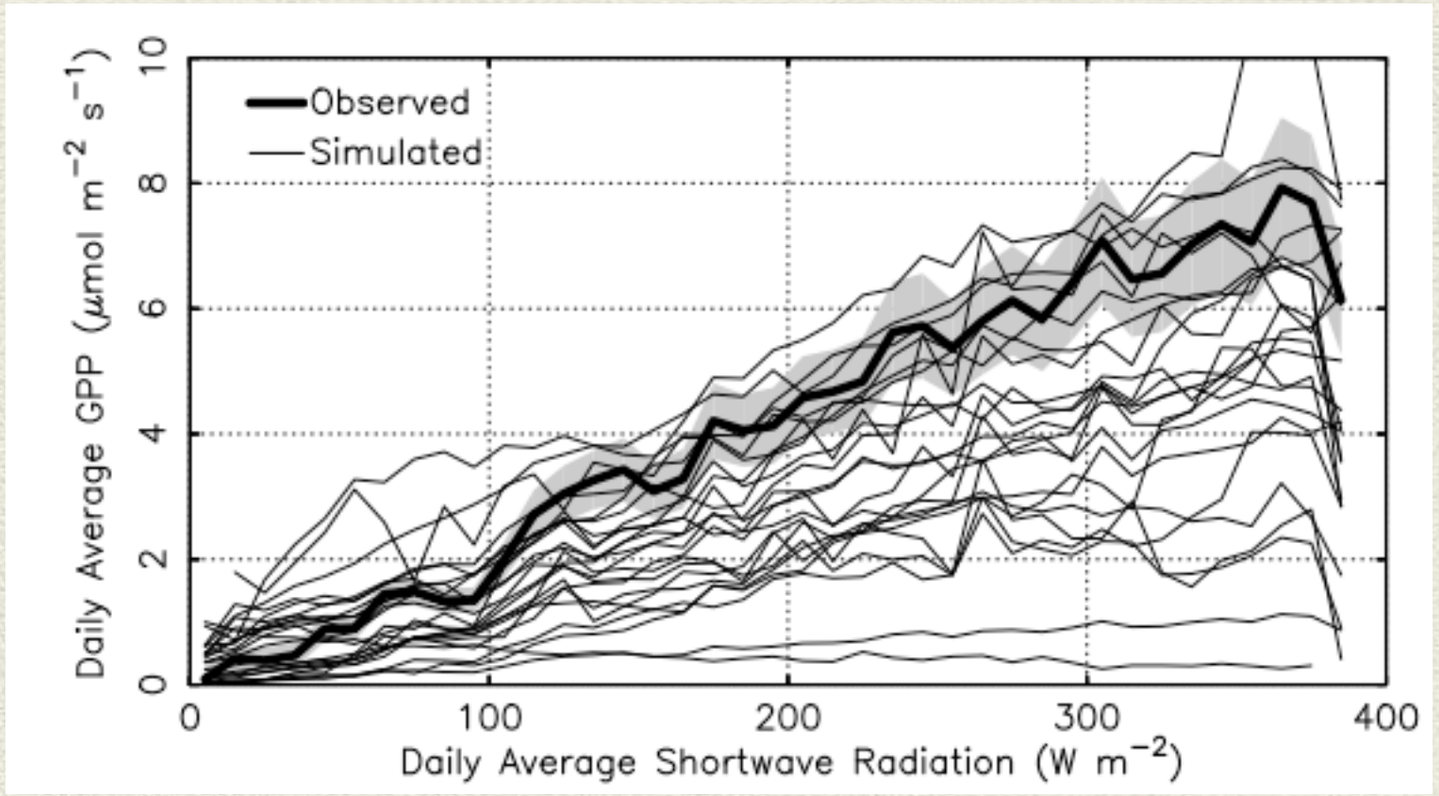
Miscalibration

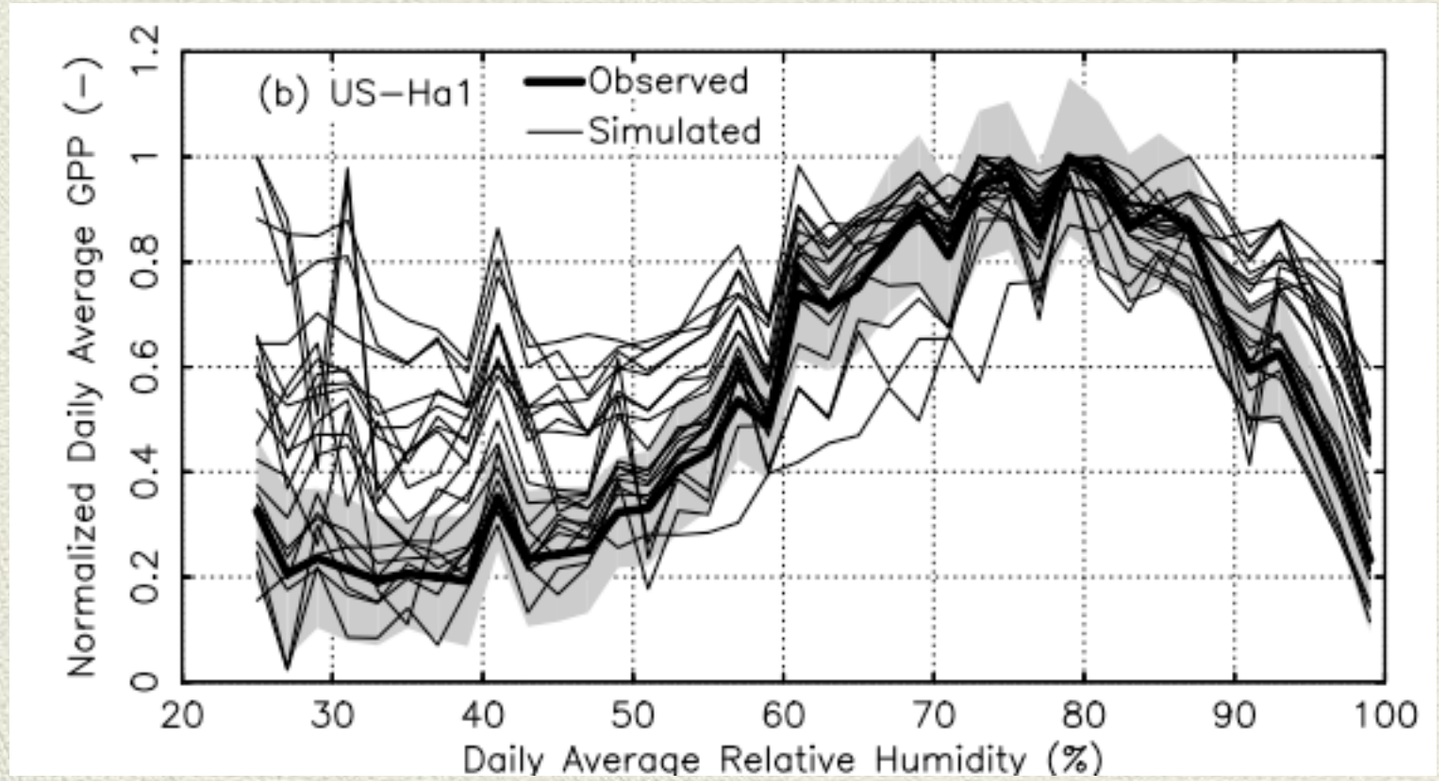


Dynamics & Drivers





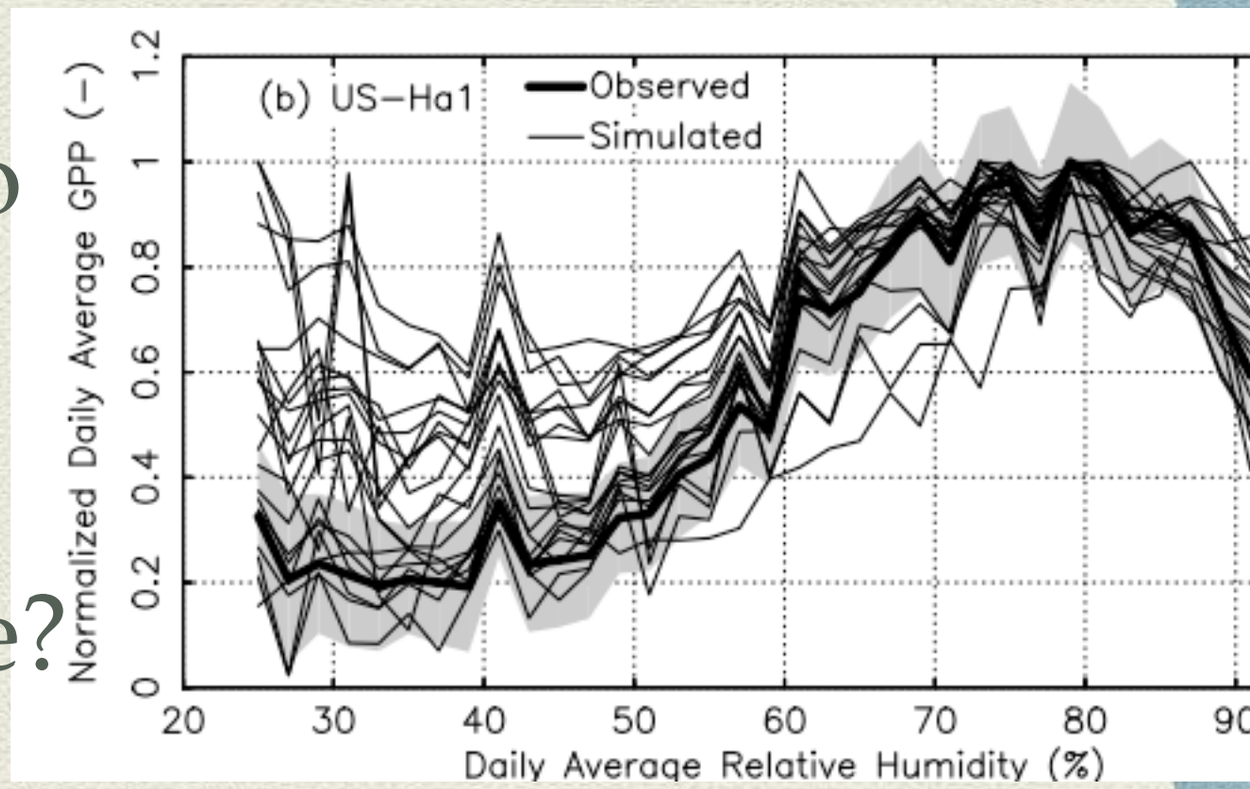




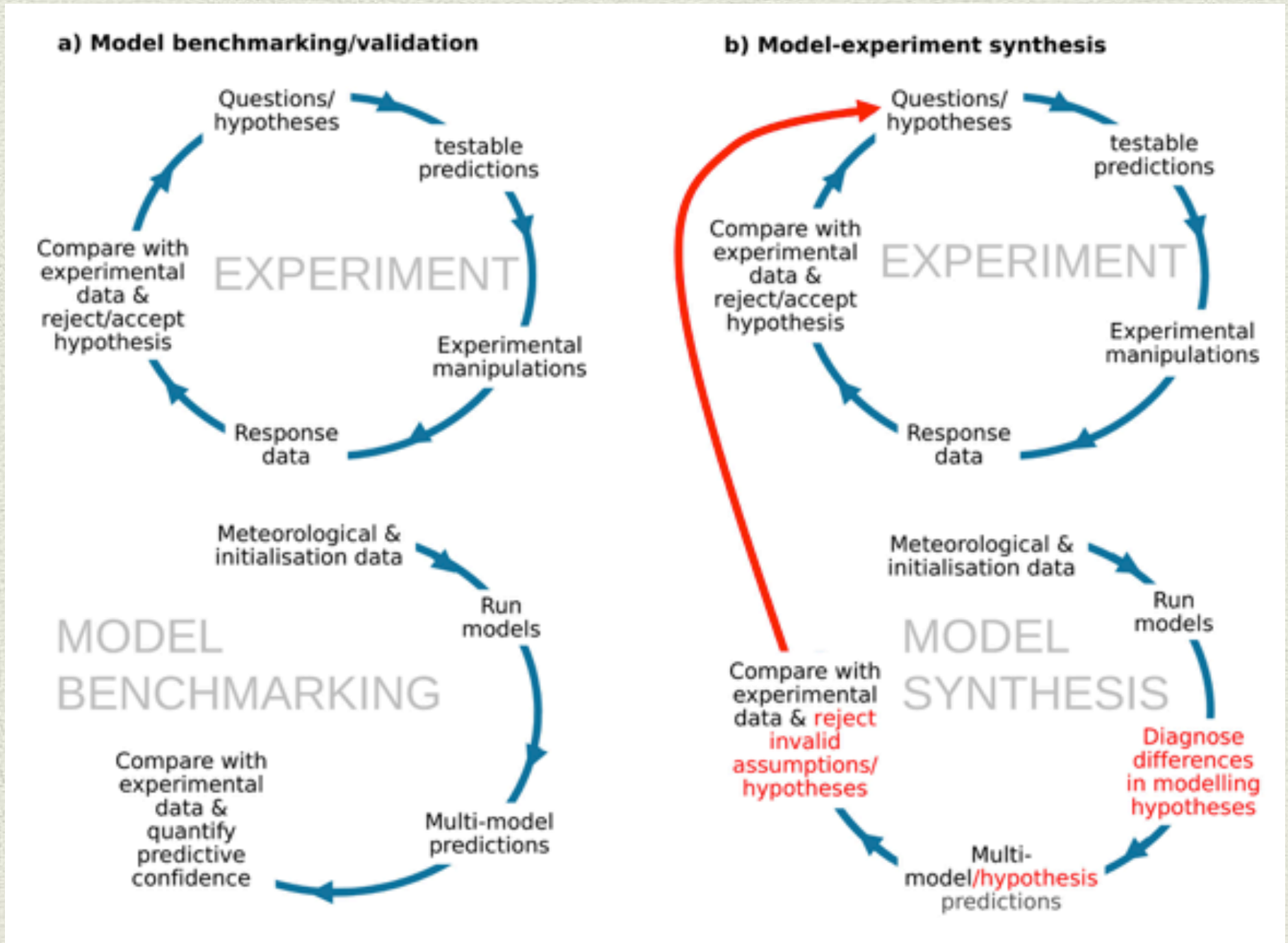
Schaefer et al. 2012 JGR-B

Diagnosing a model is Hypothesis Testing

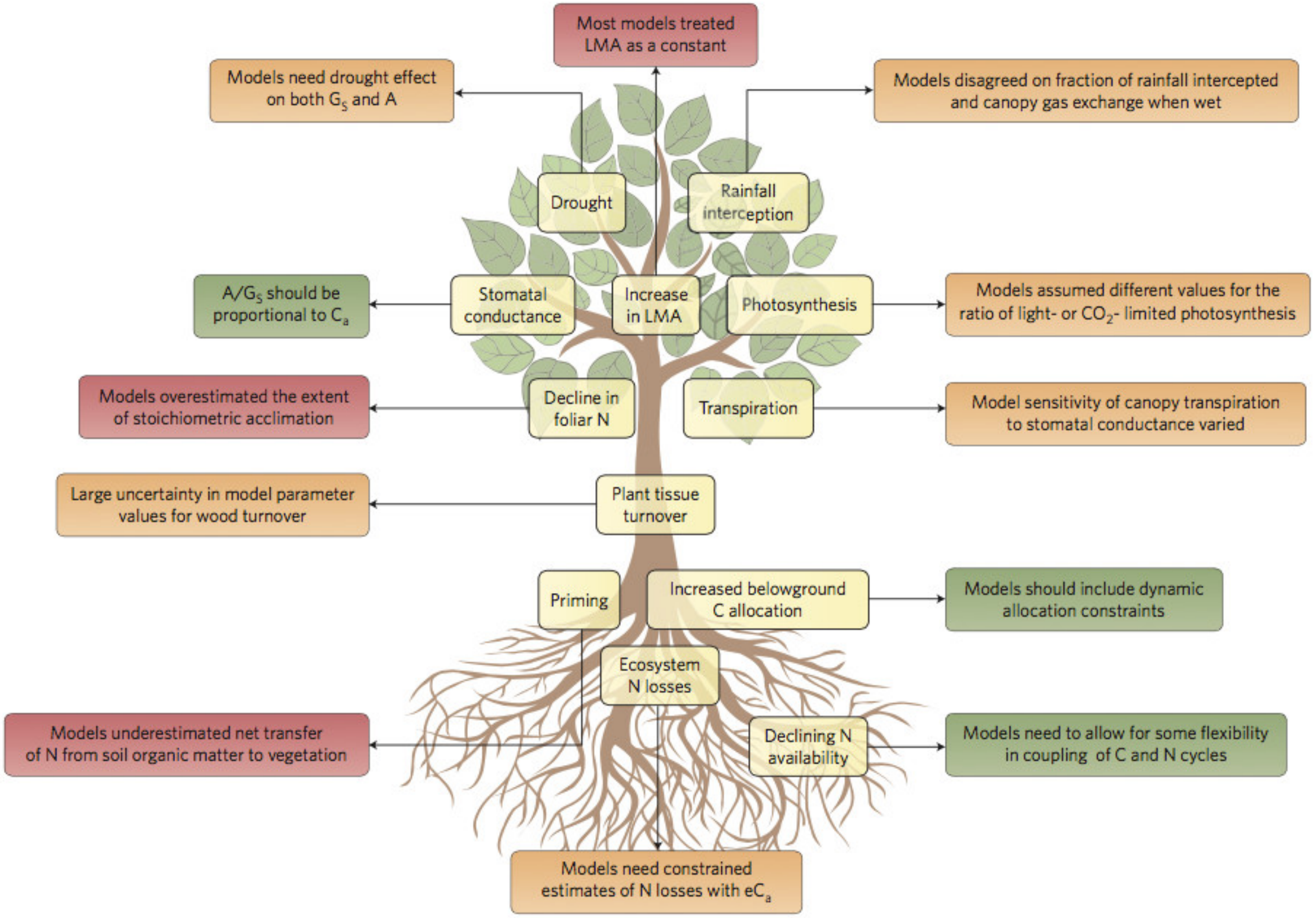
- Why would a model fail at low humidity?
- Stomatal sensitivity too low?
- Too much soil moisture?
- What experiments would I run in the model to test this?



Focus on key assumptions



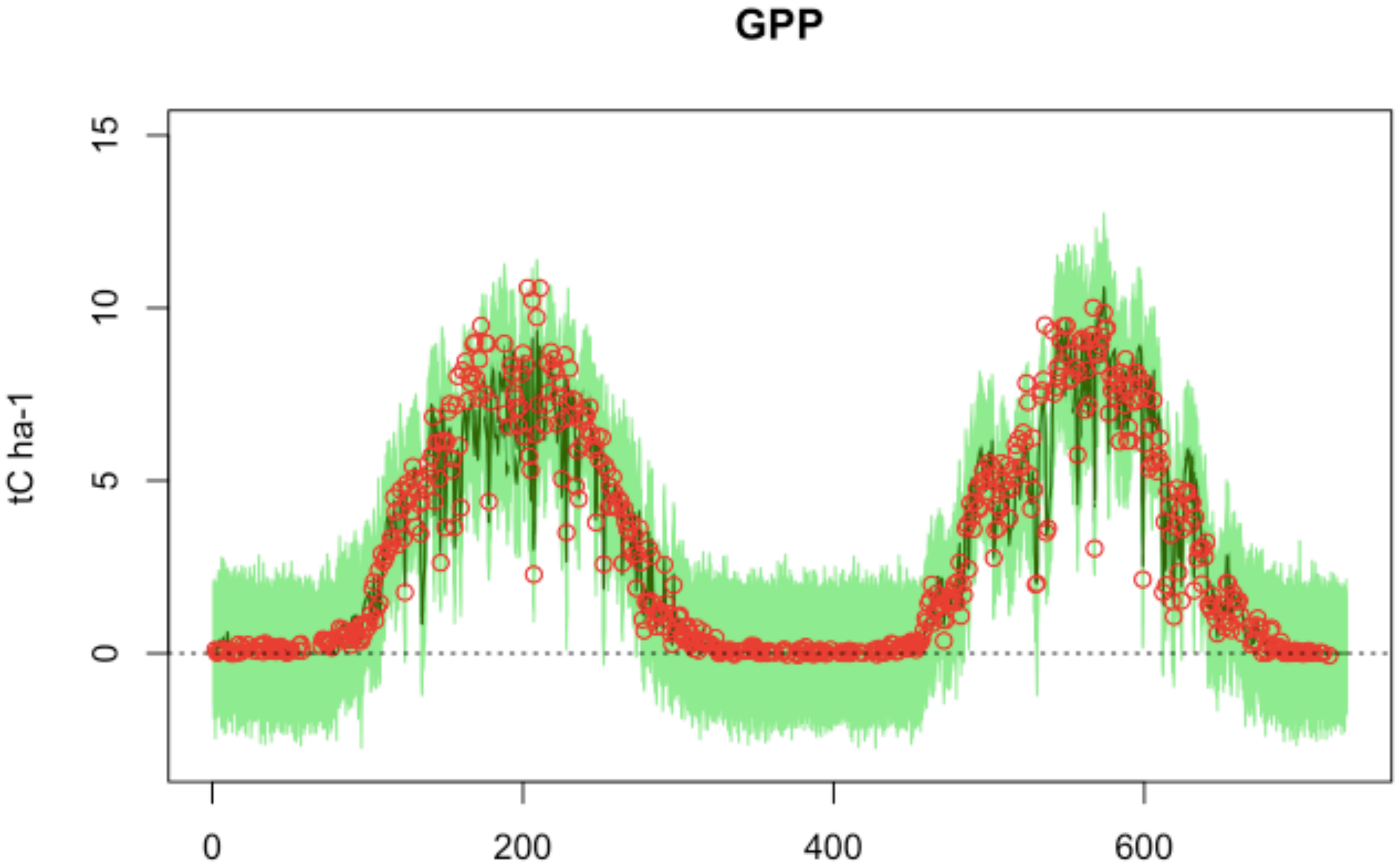
Walker et al 2014



Medlyn et al NCC 2015

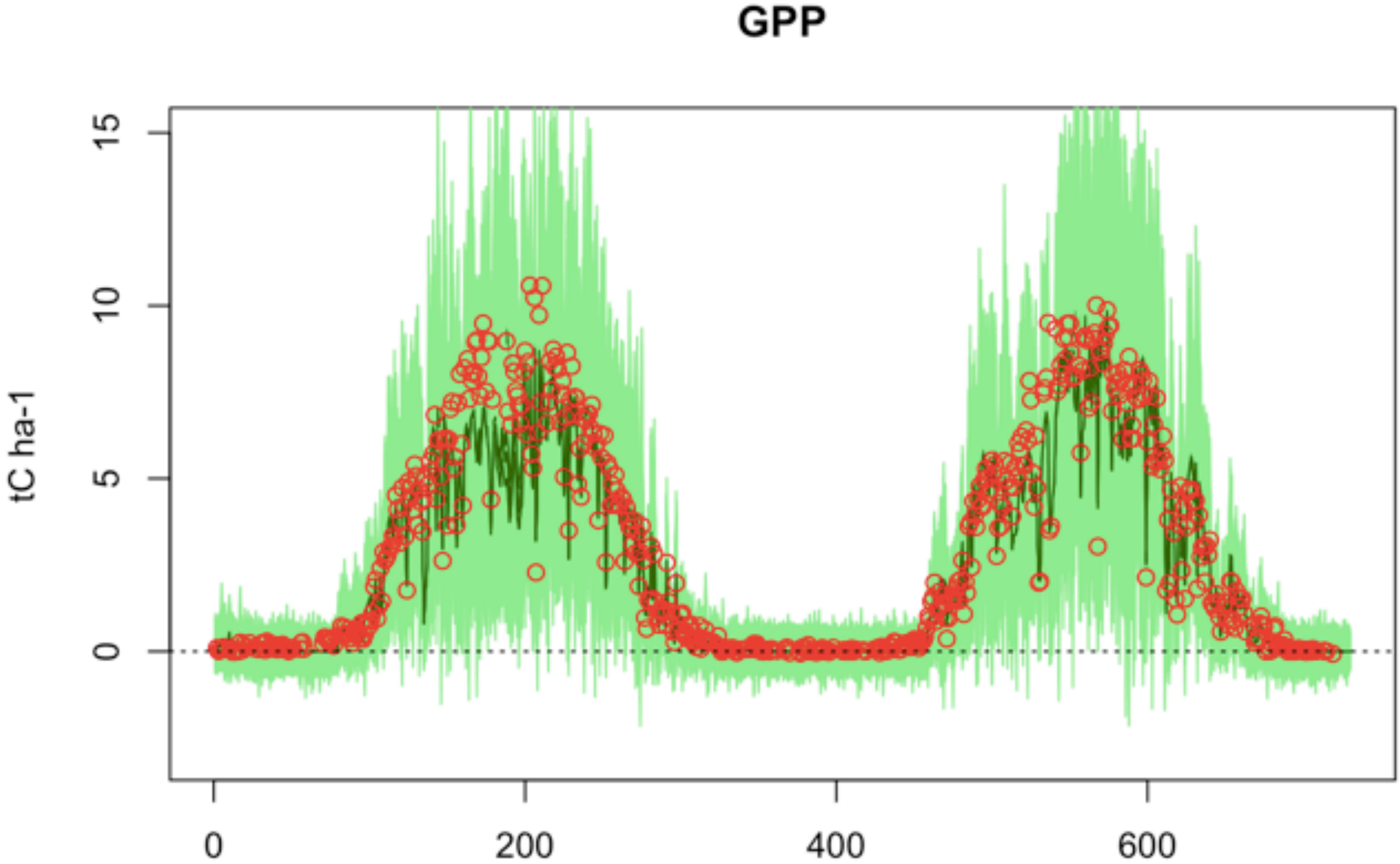
"data simulated under a model should look similar to data gathered in the real world."
Conn et al 2018

IN THE FITTING, WE ASSUMED IID NORMAL ERRORS



Does that seem like an adequate description of the data?

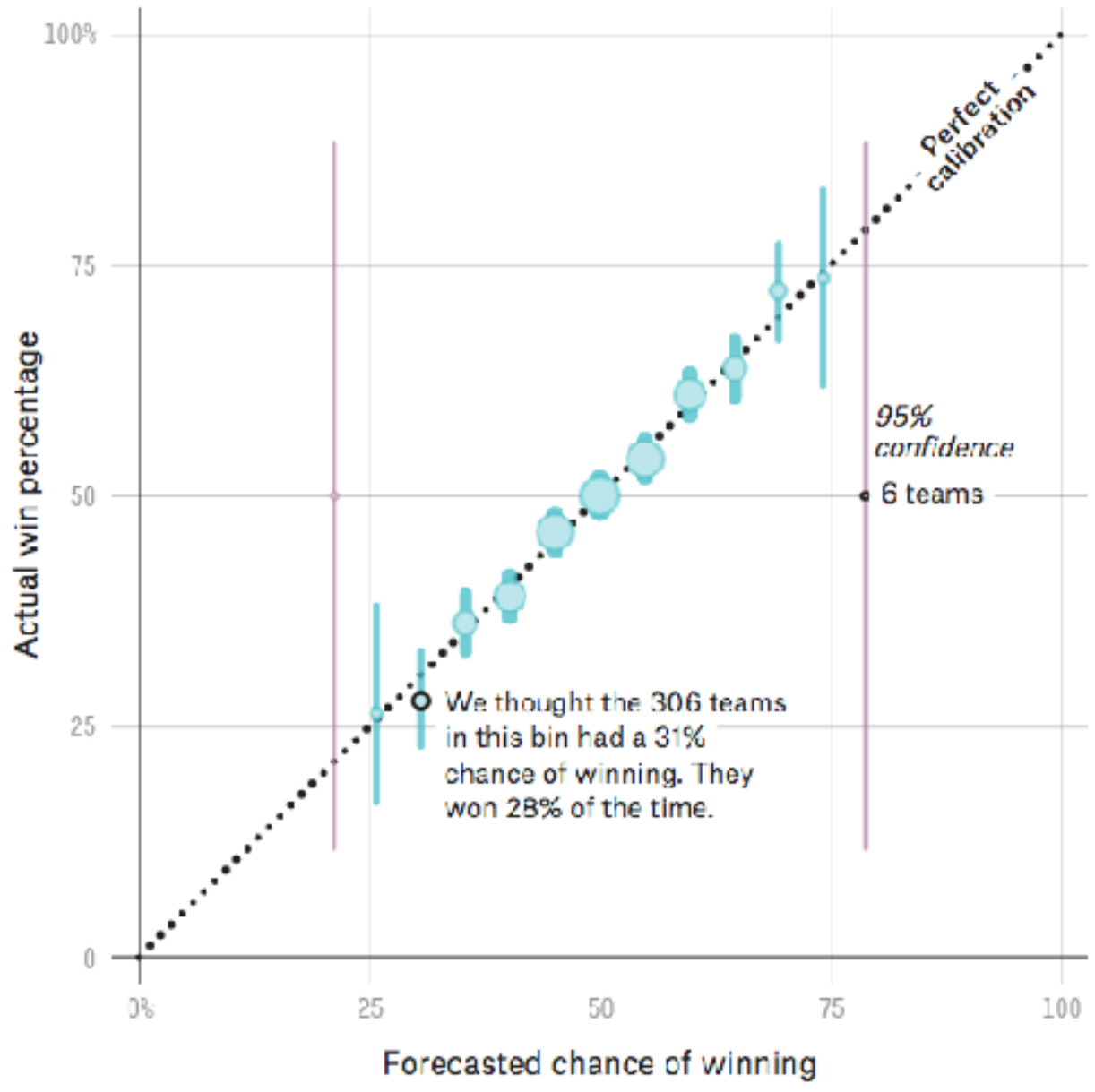
IN THIS FITTING, WE ASSUMED EXPONENTIAL ERRORS WITH NON-CONSTANT VARIANCE



Does that seem like an adequate description of the data?

PUBLISHED APR. 4, 2019, AT 7:00 AM

How Good Are FiveThirtyEight Forecasts?



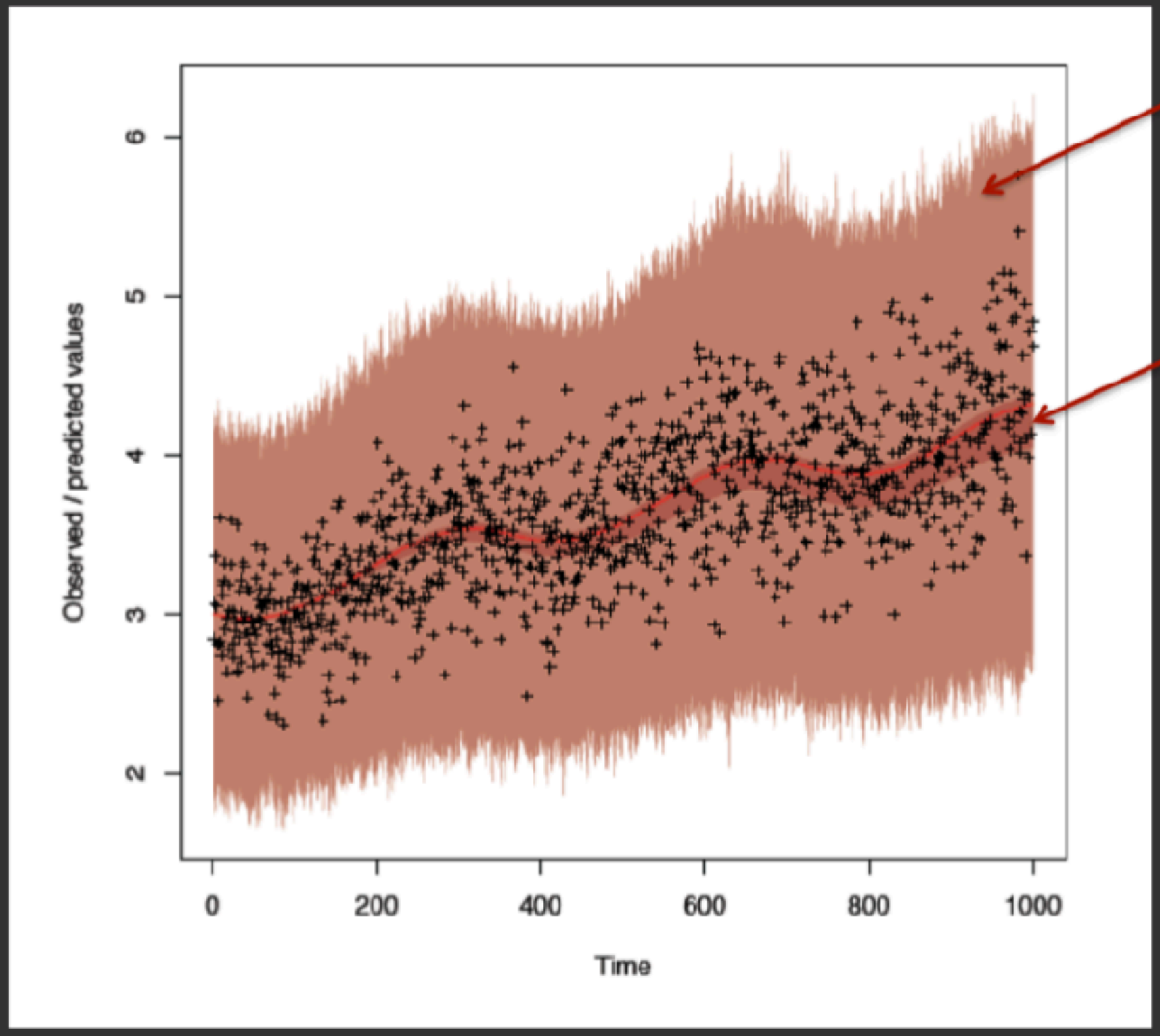
https://projects.fivethirtyeight.com/checking-our-work/

Bayesian p-value / prediction interval

- Posterior predictive distribution is the uncertainty of the „true“ value

- **Prediction interval** is the expected variance of the observed values = PPD + error
  - Shows us what distribution we would expect for the data

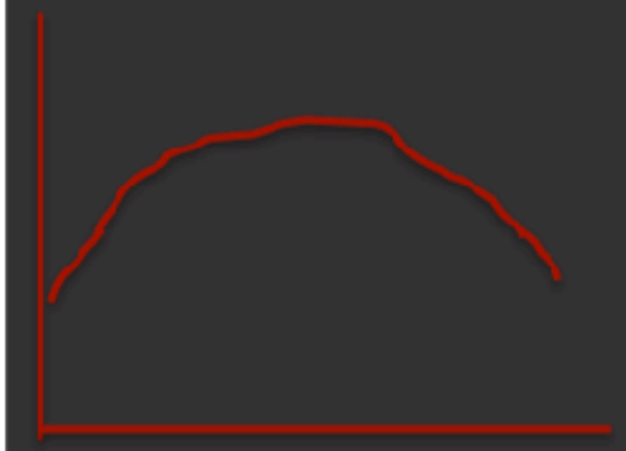
- Bayesian p-value is when we use PPD + error to calculate the value of the cdf of the observed data
  - Distribution should be flat (uniform)
  - „Bayesian residuals“



PPD + Error

Posterior Predictive Distribution

Distribution of ecdf values for residuals



0

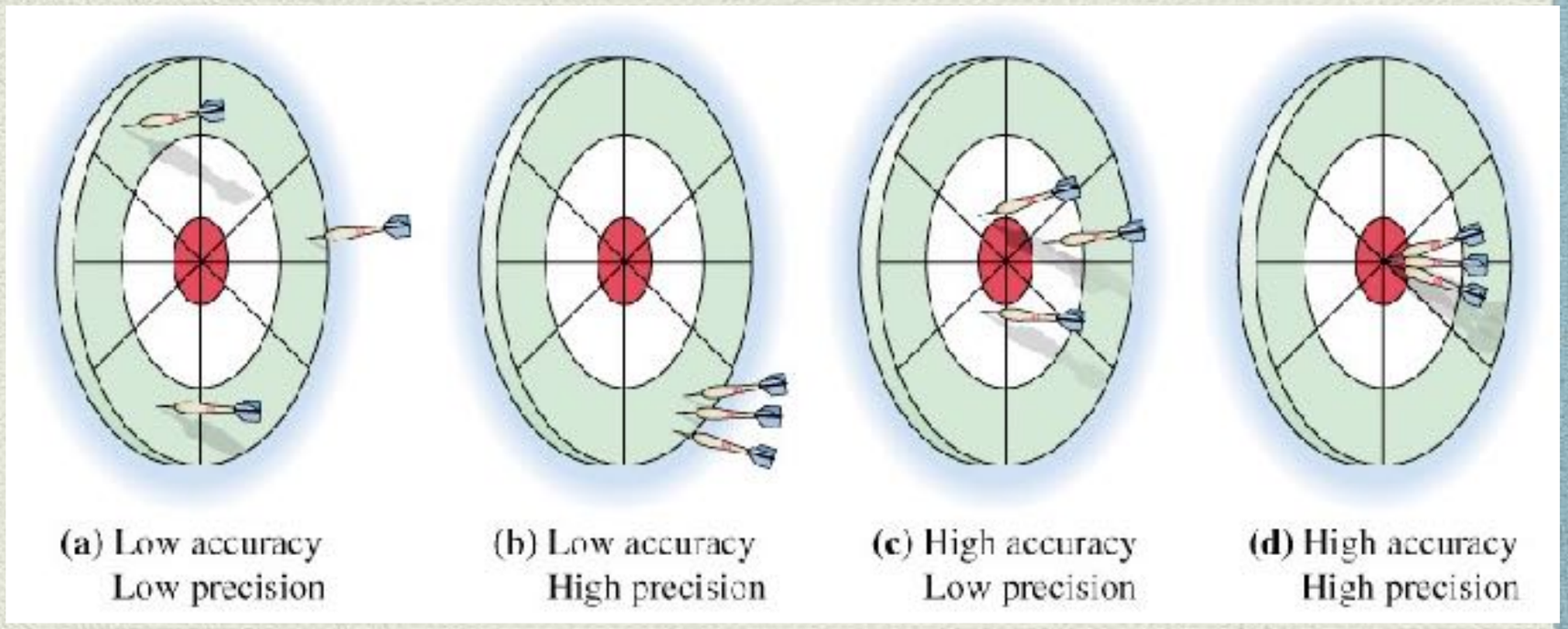
1

Step 3: Quantitative Skill Assessment

Error Statistics

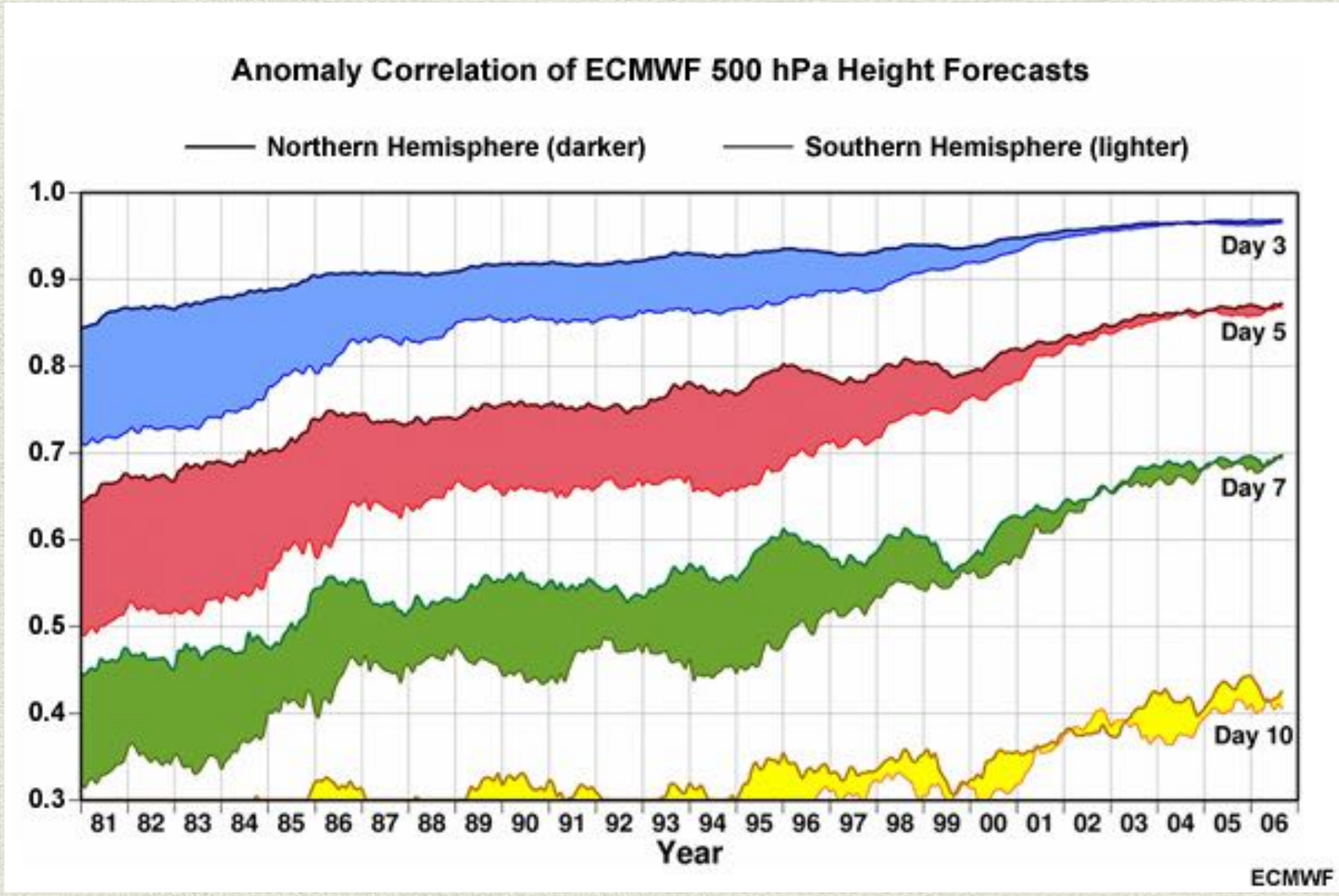
- Root Mean Square Error (RMSE)
- Bias
- Correlation (r)
- R^2
- Regression slope

$$RMSE = \sqrt{\frac{1}{N}\sum_{i=1}^{N}(y_i - x_i)^2}$$

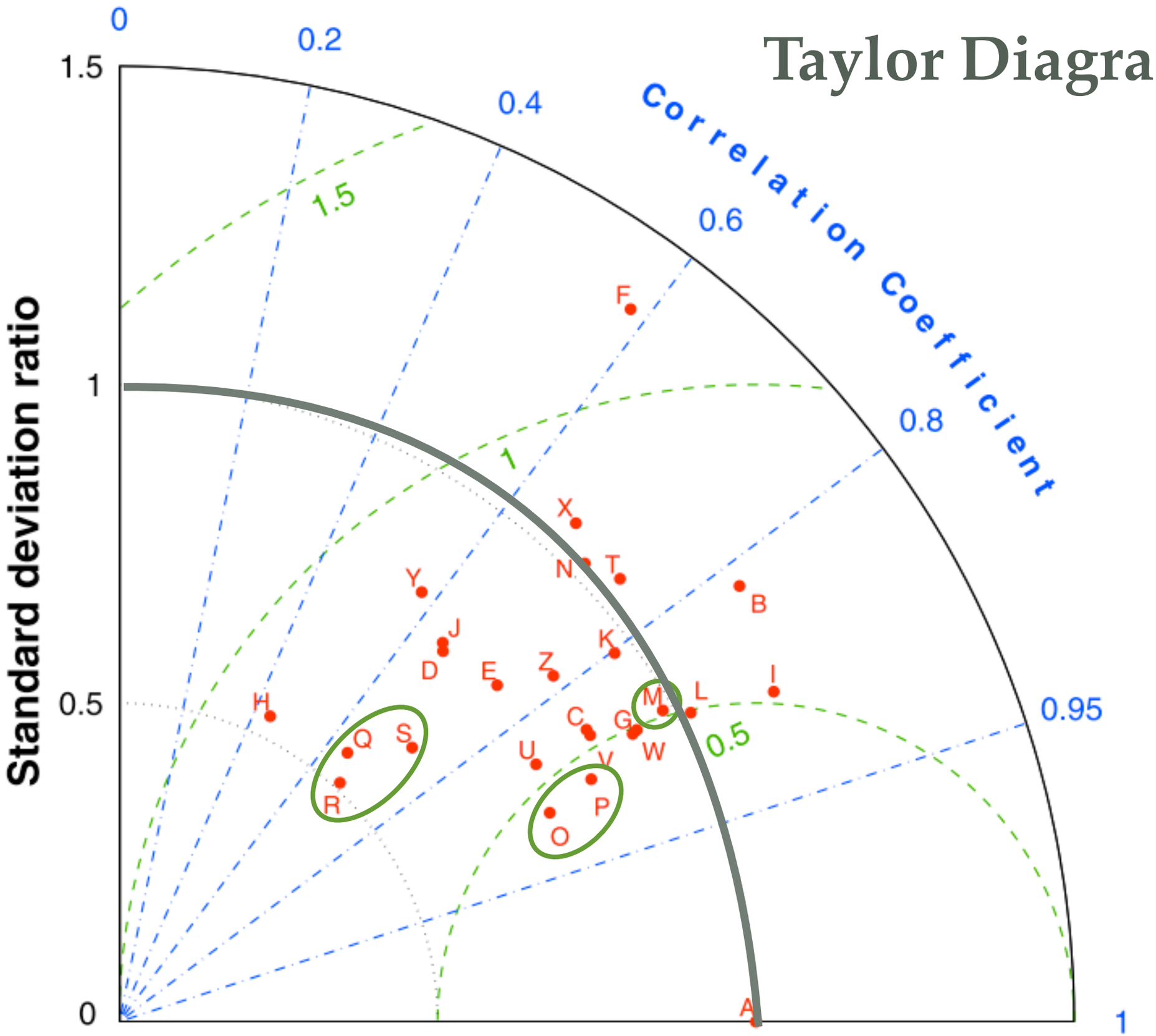


Proper: based on the metric used for calibration
Local: depends on data that could actually be collected

Correlation



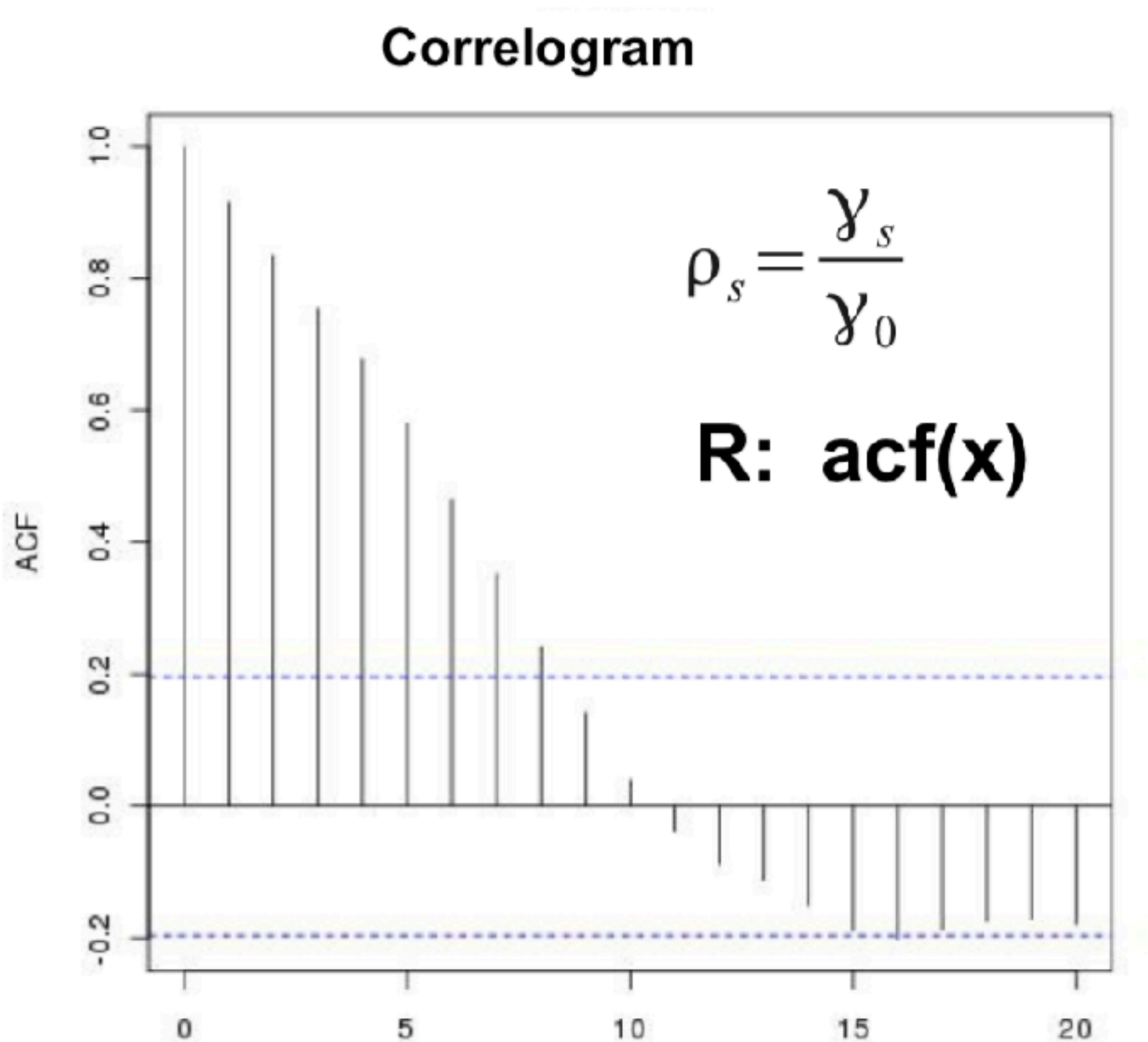
Taylor Diagram



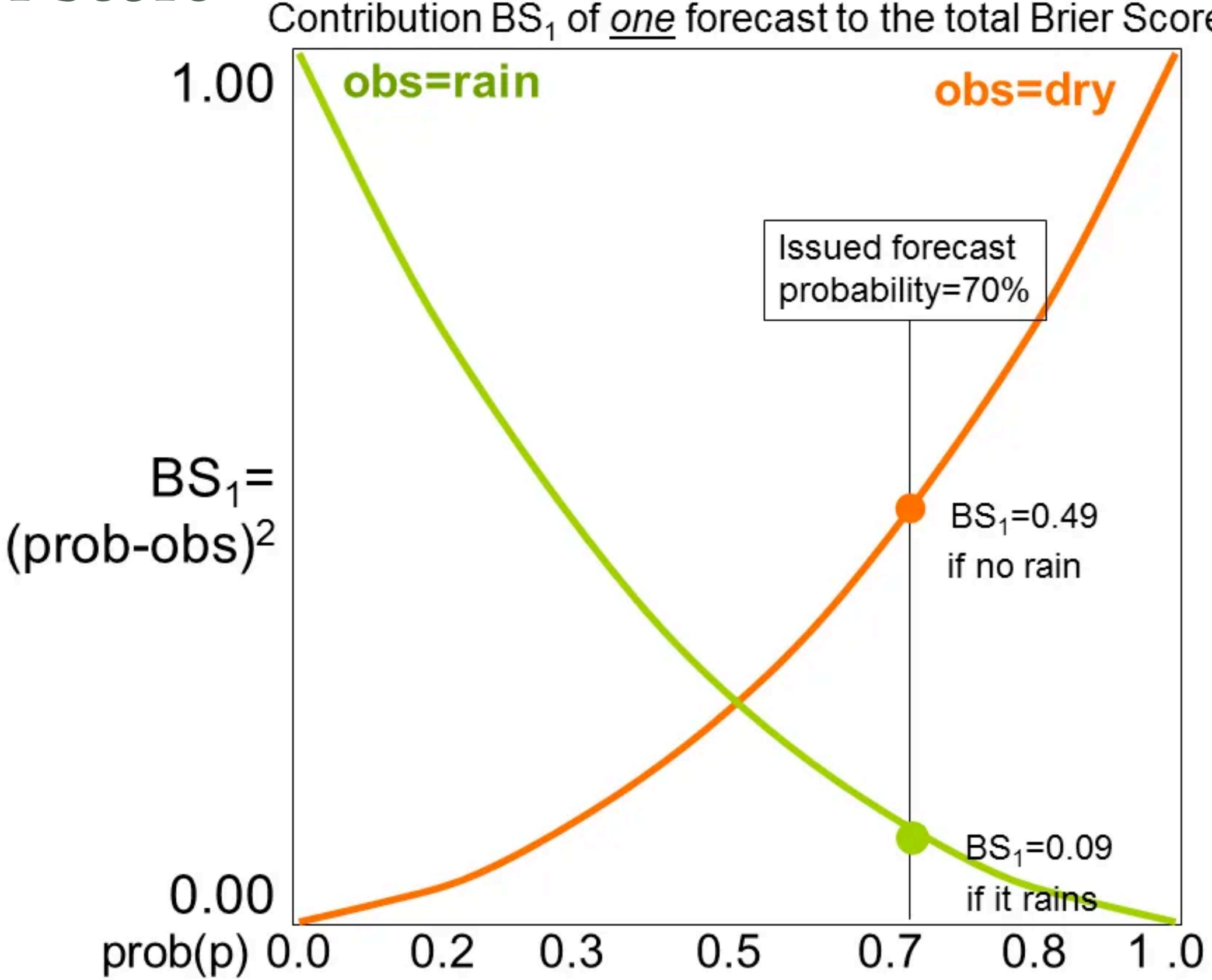
A OBSERVED
B AGROIBIS
C BEPS
D BIOMEBGC
E CAN–IBIS
F CNCLASS
G DLEM
H DNDC
I ECOSYS
J ED2
K ISAM
L ISOLSM
M LOTEC
N LPJ
O MEAN_ALL
P MEAN_DIURN
Q MODIS_ALG
R MODIS_5
S MODIS_5.1
T ORCHIDEE
U SIB
V SIBCASA
W SIBCROP
X SSIB2
Y TECO
Z TRIPLEX

Schaefer et al. 2012 JGR-B

Autocorrelation



Brier score


Contribution BS_1 of *one* forecast to the total Brier Score
1.00
obs=rain
obs=dry
Issued forecast
probability=70%
BS_1=
$(prob-obs)^2$
BS_1=0.49
if no rain
BS_1=0.09
if it rains
0.00
prob(p) 0.0 0.2 0.3 0.5 0.7 0.8 1.0

Continuous Ranked Probability Score

$$\mathrm{CRPS}(F, x) = -\int_{-\infty}^{\infty} \left(F(y) - \mathbb{1}\{y \geq x\}\right)^2 \mathrm{d}y$$

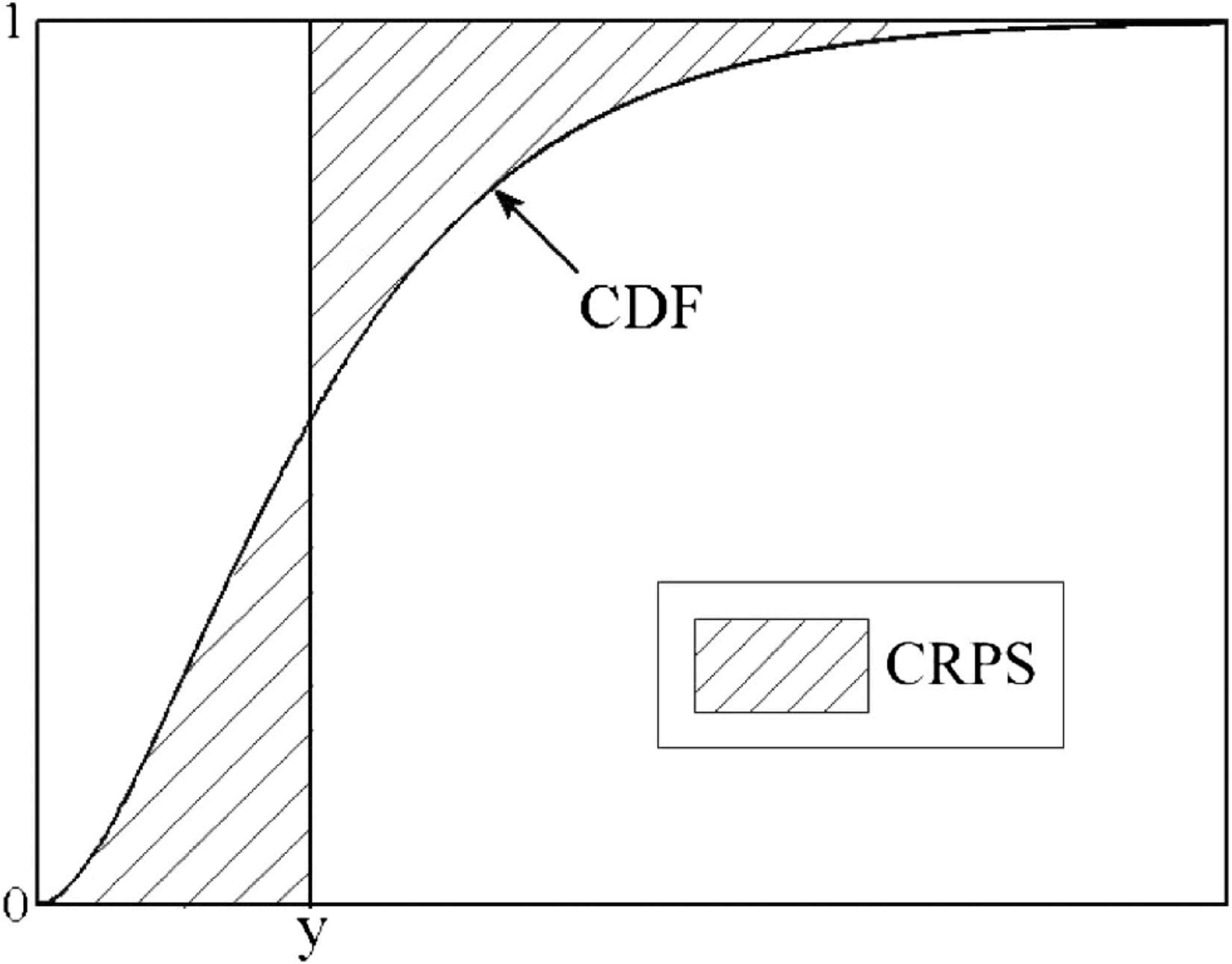
Ensemble member

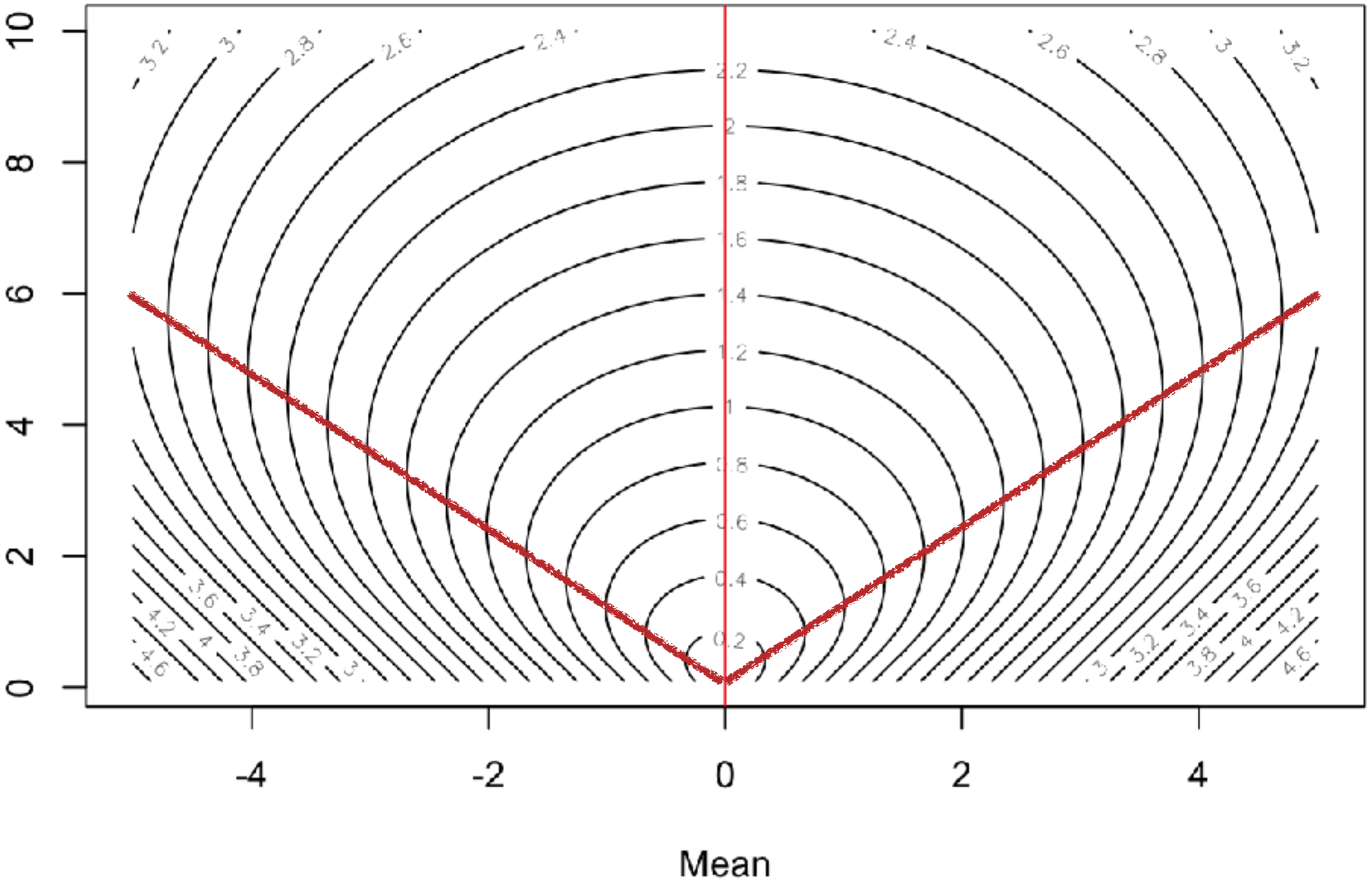
Data

$$\mathrm{CRPS}(\hat{F}_m, y) = \frac{1}{m}\sum_{i=1}^{m} |X_i - y| - \frac{1}{2m^2}\sum_{i=1}^{m}\sum_{j=1}^{m} |X_i - X_j|$$

Mean Absolute Error

Penalty for ensemble spread


1
CDF
CRPS
0
y

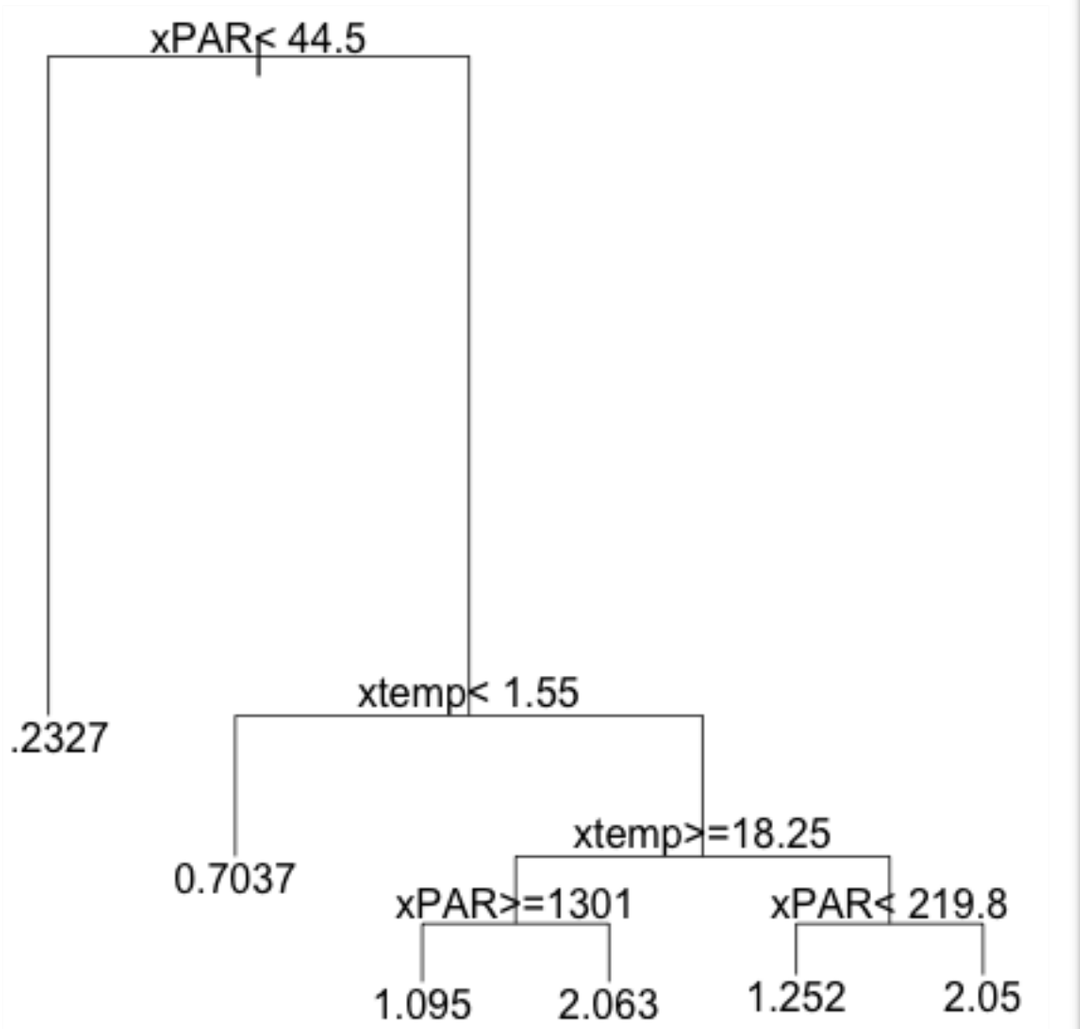


https://github.com/eco4cast/neon4cast-scoring/blob/main/CRPS_example_JRT.Rmd

Data mining the residuals

- Wide variety of Data Mining algorithms in use
- Potentially useful for generating hypothesis about when/where model fails
- "Correct" the forecast
- Hybrid models increasing
- CART
- GAM
- Random Forests
- Boosted regression trees
- Artificial Neural Network
- Support Vector Machines

CART



Random Forest