



**Hampshire Avon**  
Demonstration Test Catchment

# Developing a framework for including uncertainty analysis to estimate catchment nutrient fluxes and behaviour.

Charlotte Lloyd

Demonstration  
Test  
Catchments

Lloyd, C. E.M., Freer, J.E., Collins, A.L., Johnes, P. J., Coxon, G.  
and Hampshire Avon DTC team.



[illegible]

```
graph TD; A[Change] <--> B[Change Objectives and Purpose]; B <--> C[Analysis of Change]; A -.-> D([Declare change detection success?]); B --> D; C -.-> D;
```

**The interplay between Evidence, Analysis and Objectives of Change**

**Change**

- Record Length (pre and post)
- Sampling Resolution and Design
- Spatial Variability
- Data Uncertainty
- Laboratory Techniques
- Blanket or targeted interventions

**Change Objectives and Purpose**

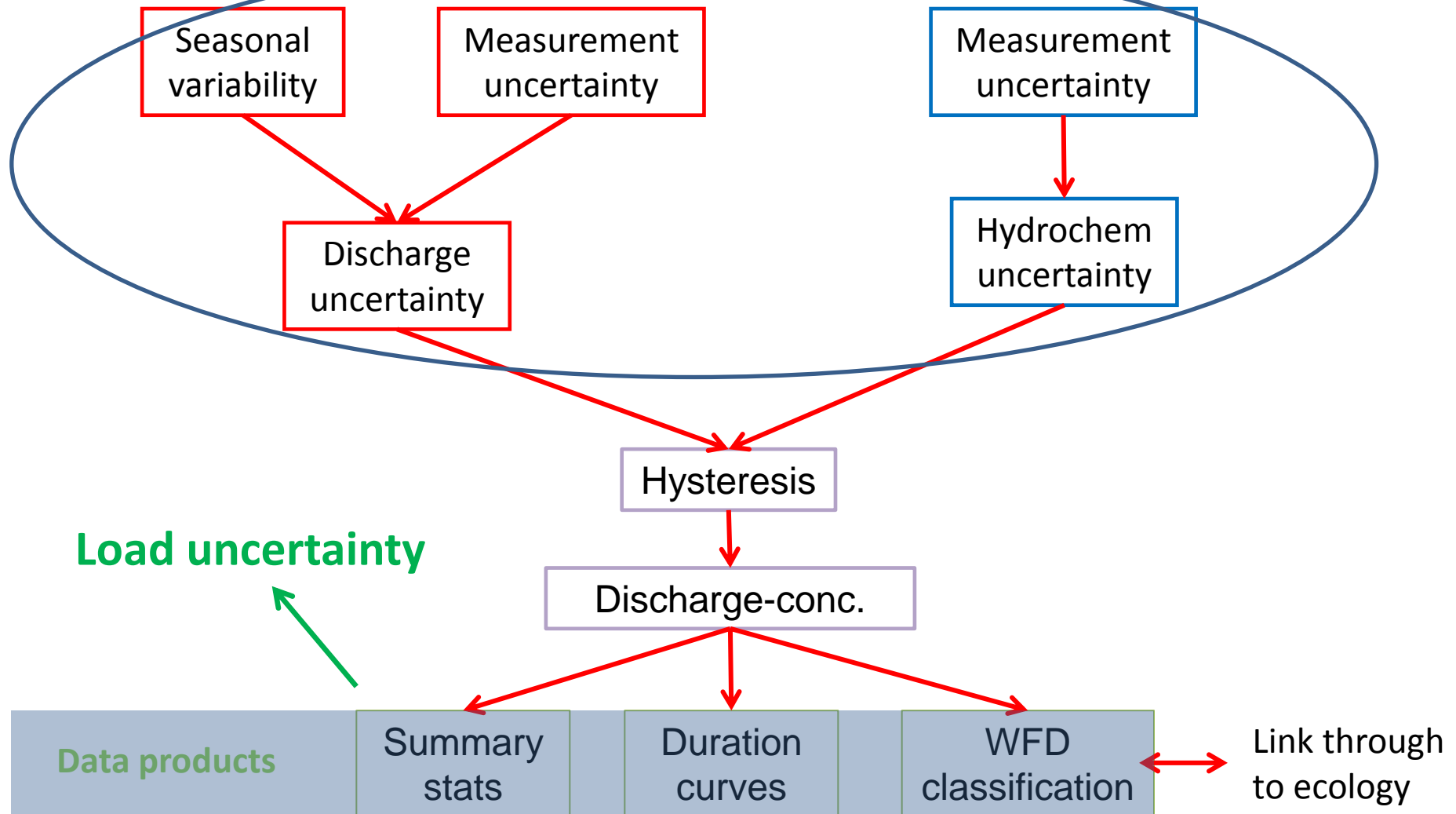
- Pollutant type
- Diffuse or point source contribution
- Storm, Seasonal or Annual
- Abrupt or long term Change
- Multi-factor or single factor

**Analysis of Change**

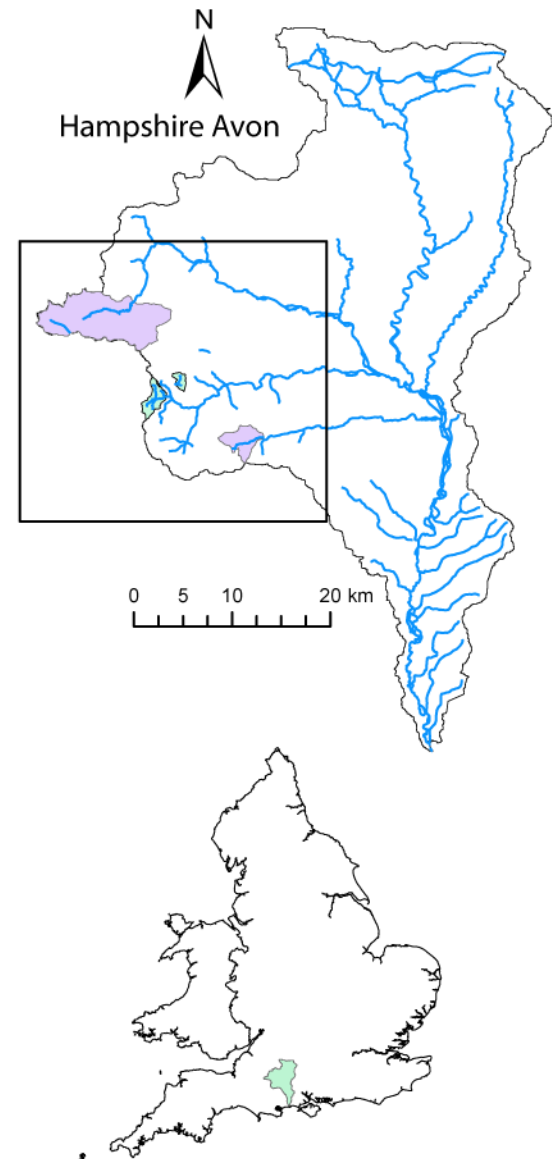
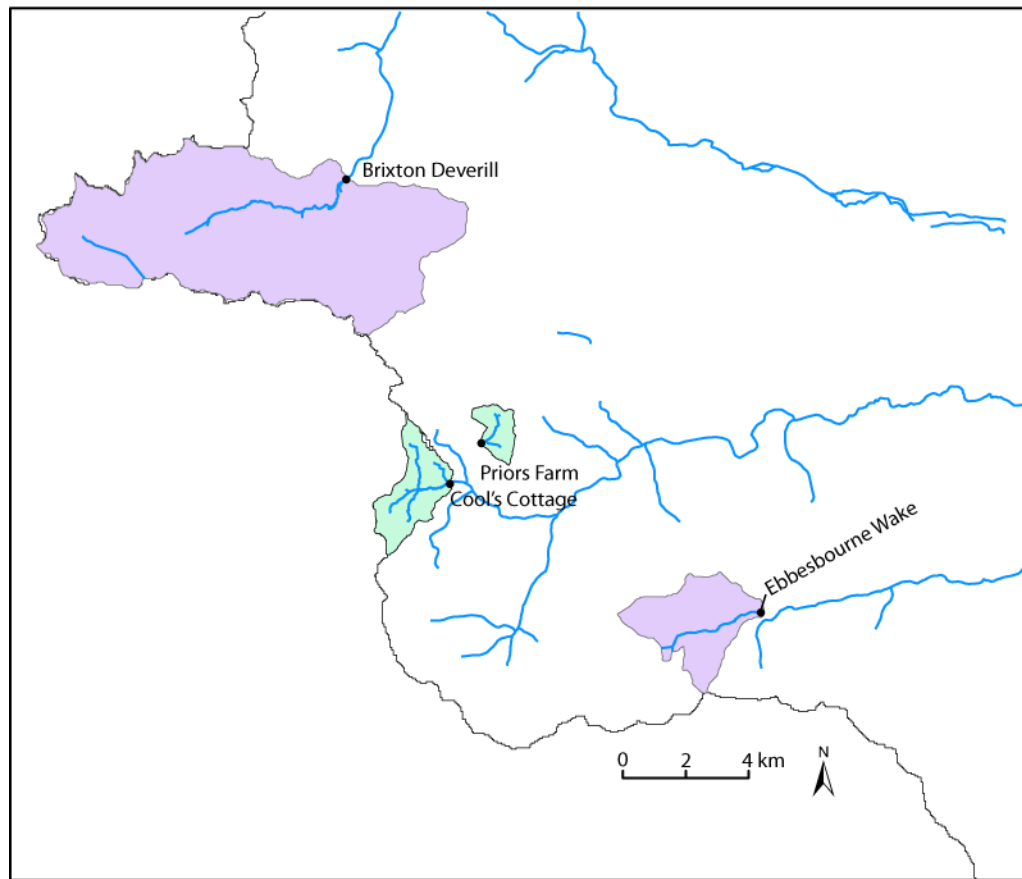
- Non-normality
- Non-stationarity
- Autocorrelation
- Outliers
- Sparseness and missing data
- Hysteresis

**Declare change detection success?**

# Conceptual framework



# Field sites



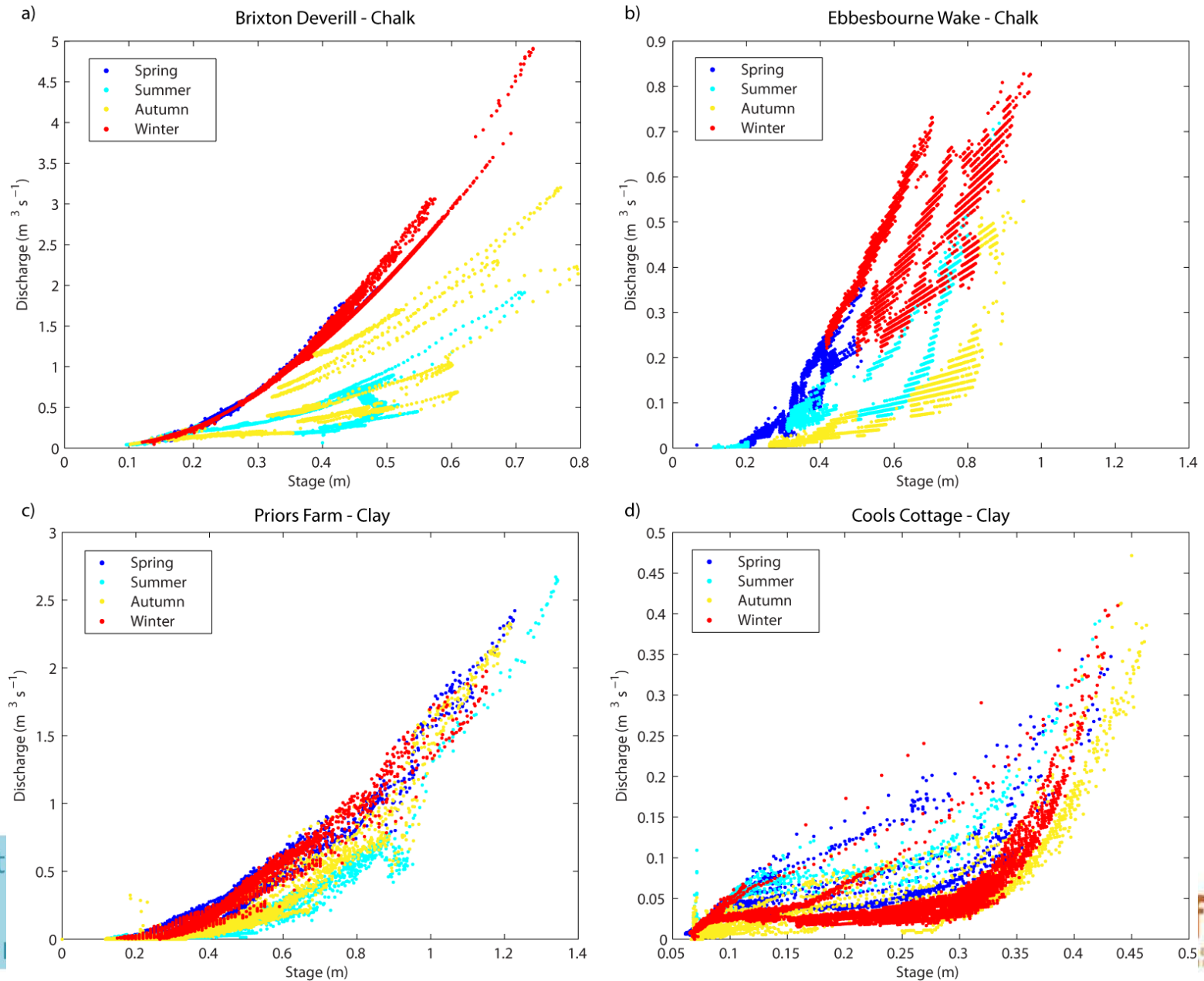
# Data

- 15 min paired stage-height and velocity
- 30 min nitrate (Nitratax) and TP (Phosphax)
- Daily nitrate and TP (ISCO and laboratory)





# Stage-discharge relationships

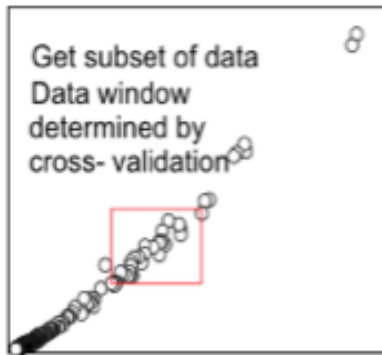


# Discharge uncertainty – non-parametric method

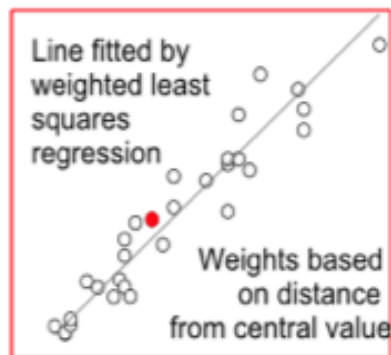
Choose a stable period (usually winter) so that uncertainty represents observational uncertainty not structural changes.

## Lowess fitting – calculate curve and variance

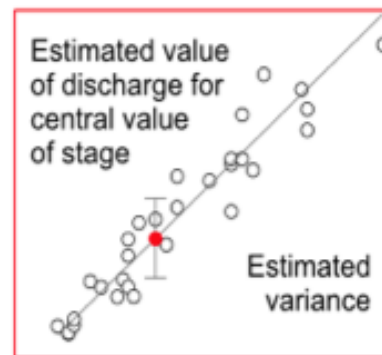
a. Original Data



b. Local fitting within window



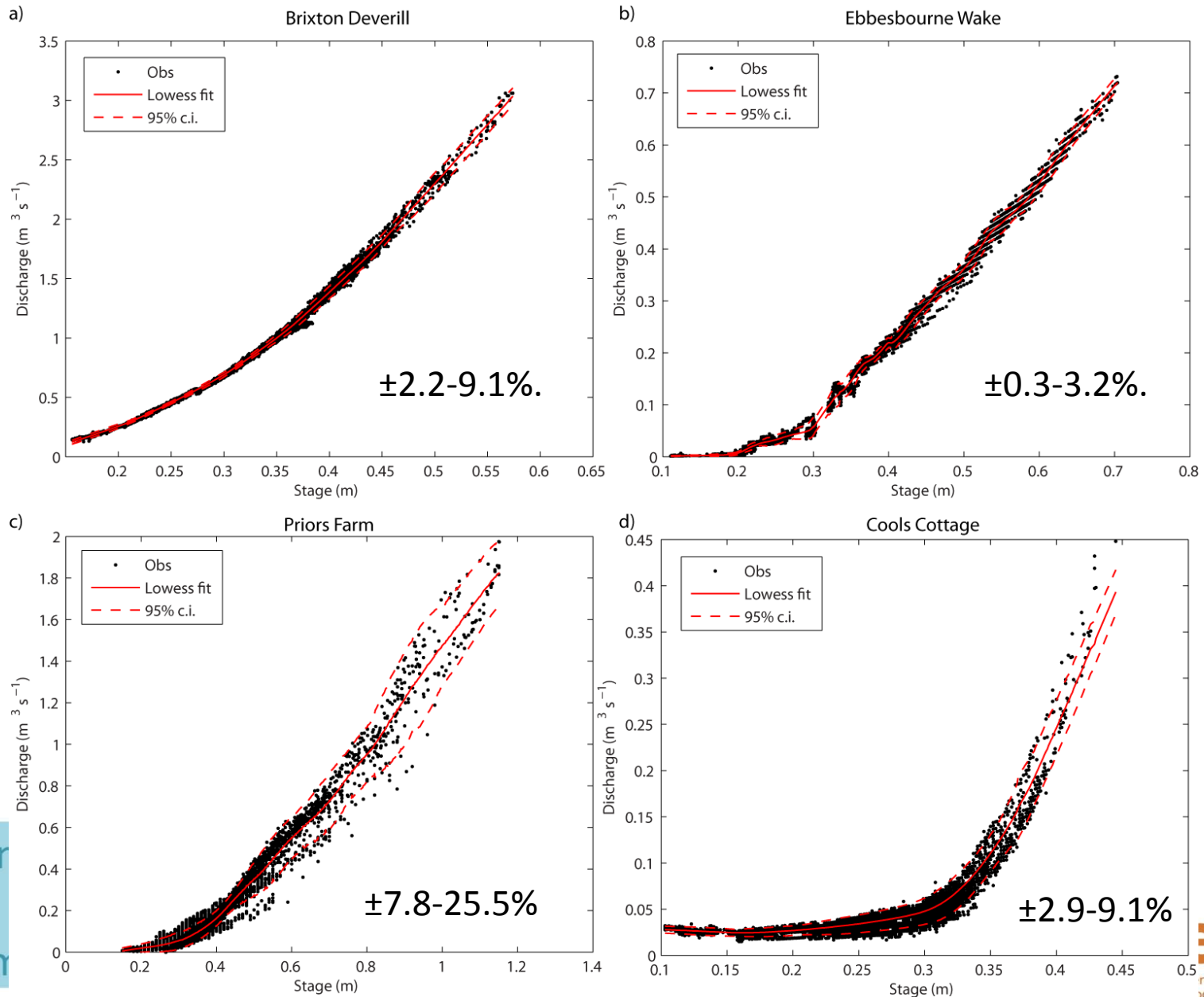
c. Estimated value and variance



d. Global lowess fit



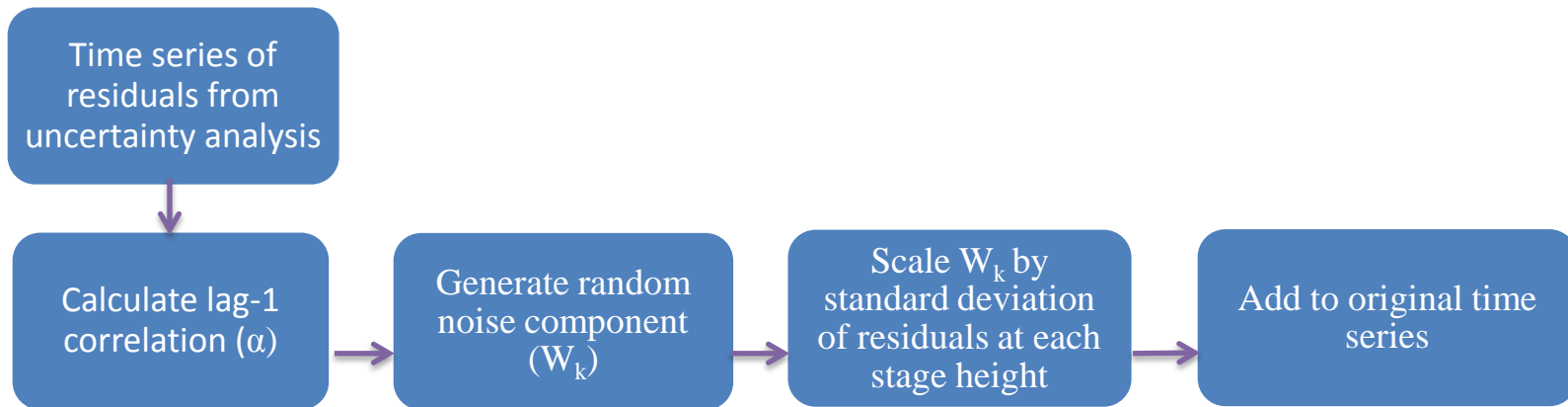
# Discharge uncertainty



Demor  
Test  
Catchm



# Error modelling



## 1st order autoregressive model

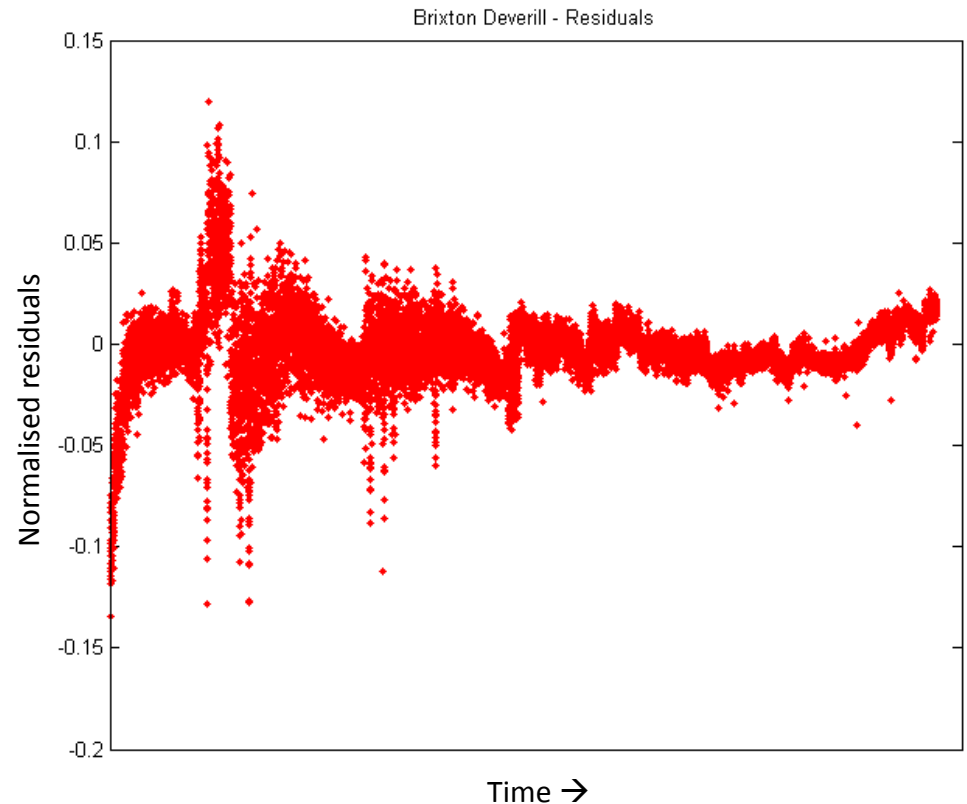
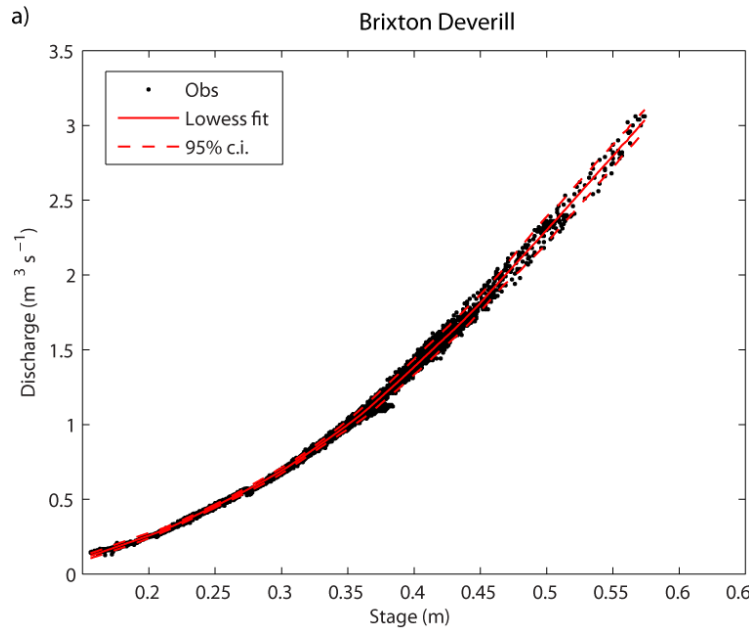
$$q_k = \alpha q_{k-1} + \sqrt{1 - \alpha^2} W_k$$

where  $q_k$  is the error at time  $k$ ,  $\alpha$  is temporal autocorrelation and  $W_k$  is random white noise at time  $k$ .

Lag-1 autocorrelation

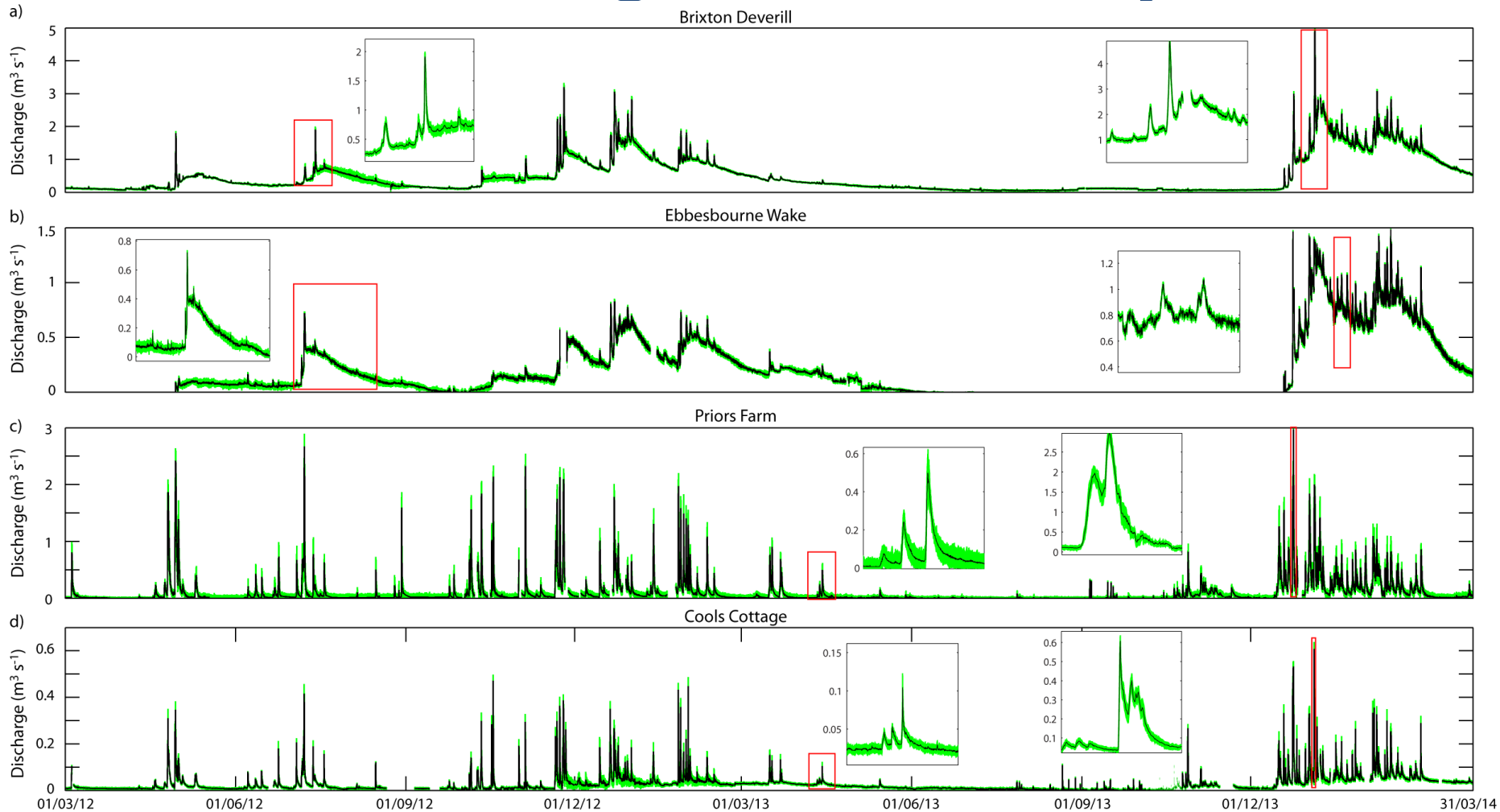
White noise (scaled by st. dev)

# Discharge uncertainty



Residuals have structure over 2 seasons however are independent over short time-scales → represented as random noise.

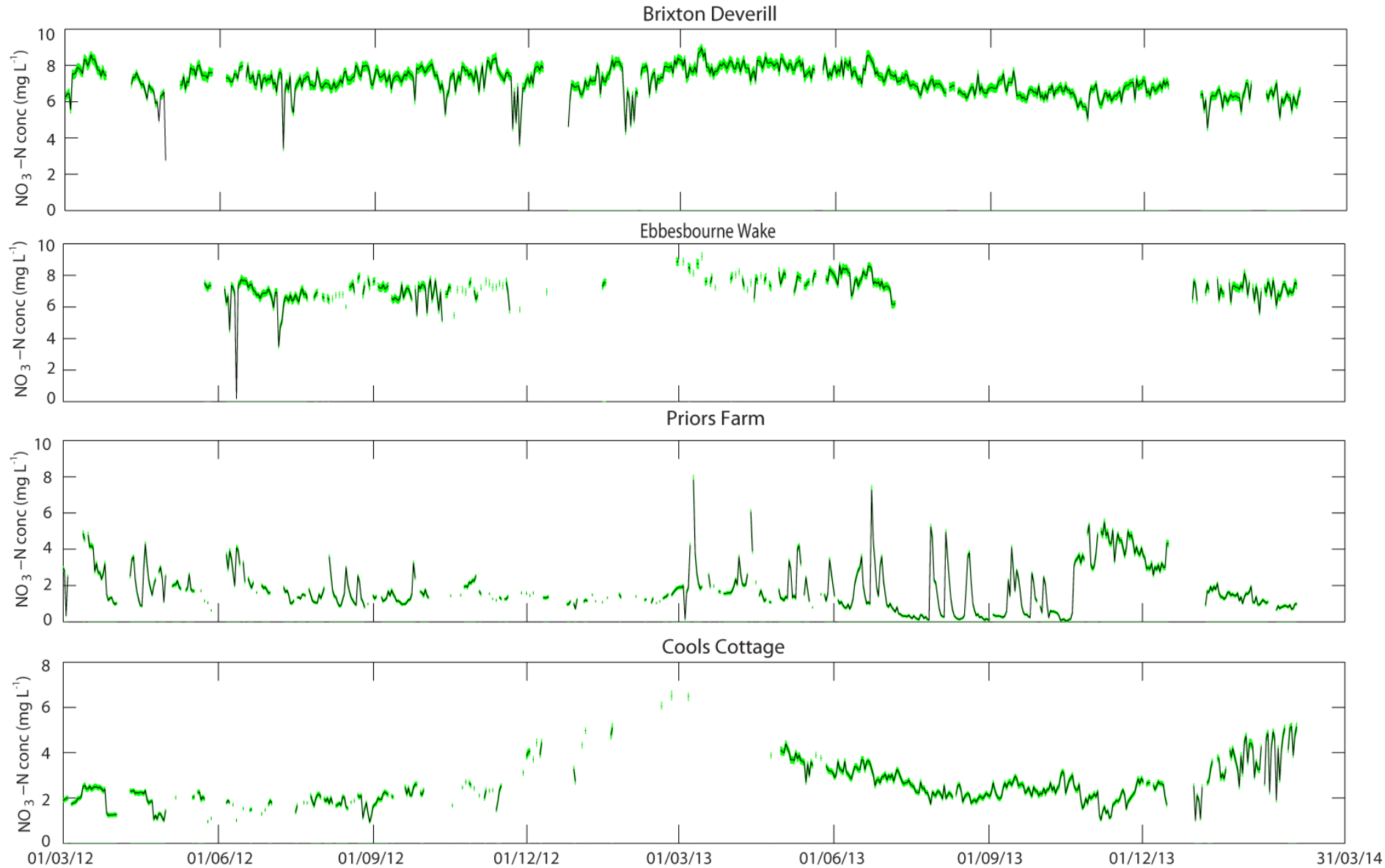
# Discharge uncertainty



# Nutrient uncertainty - Lab

- Repeated lab standards at a variety of concentrations
- Calculate the mean error and standard deviation of error
- No temporal autocorrelation in the errors as samples run in a random order
- Errors were heteroscedastic → larger st. dev errors (%) at lower concentrations

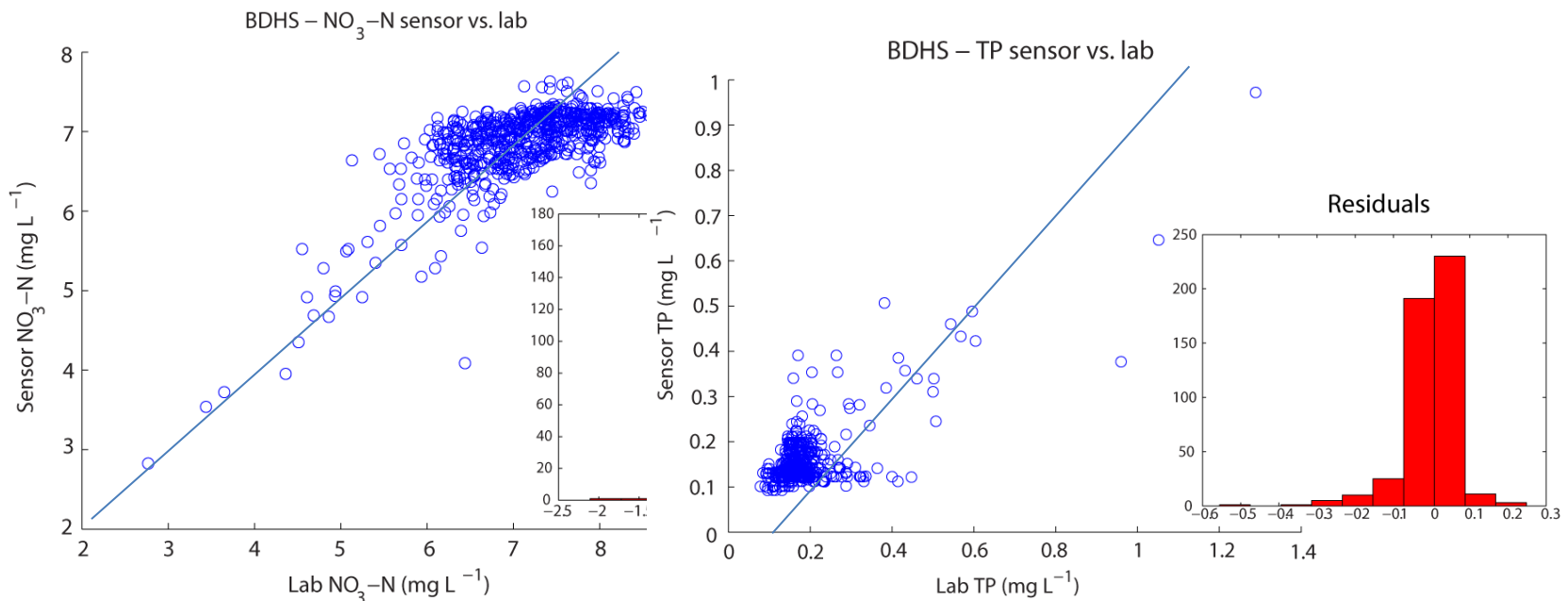
# Nutrient uncertainty



# Nutrient uncertainty

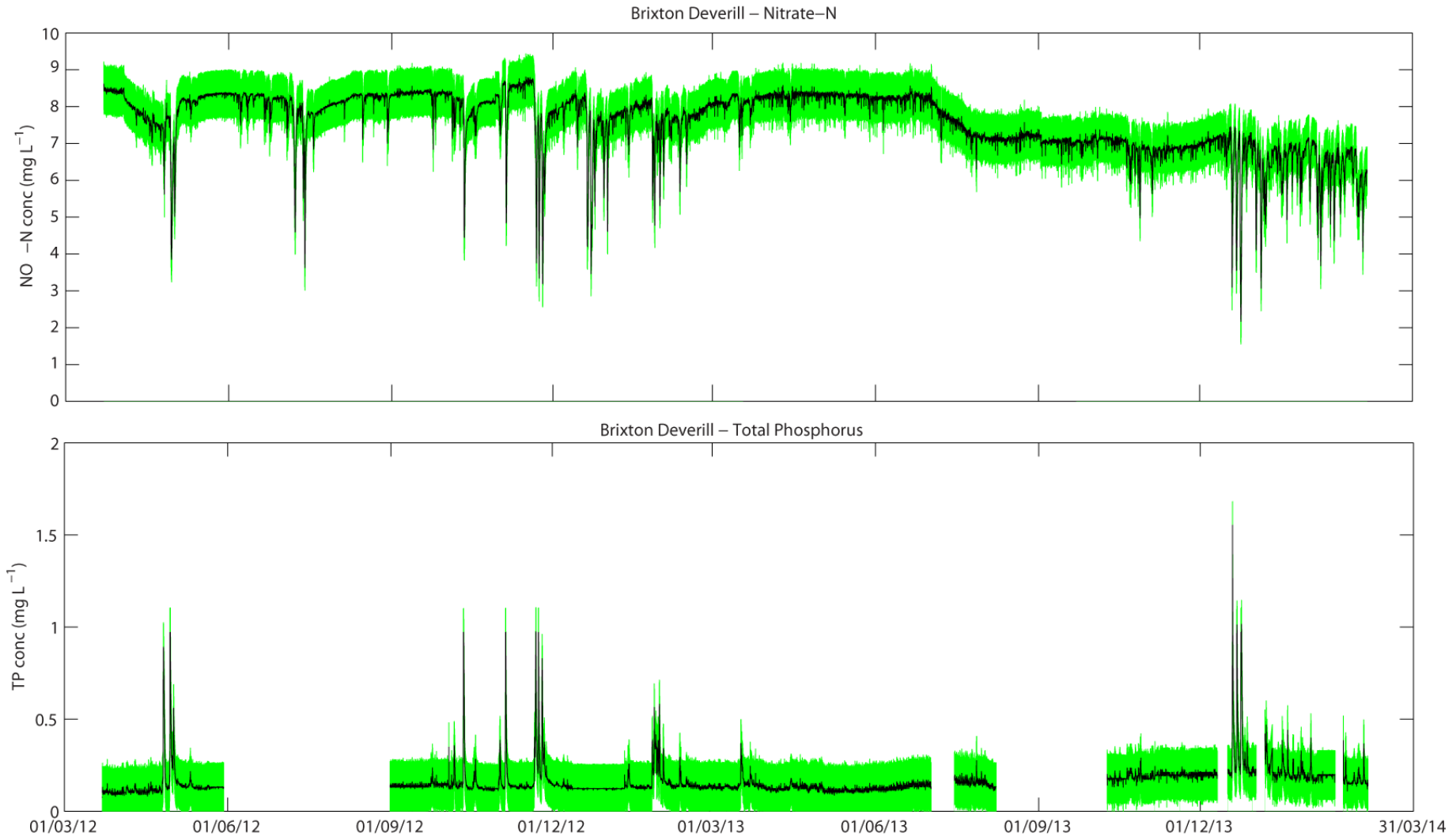
## - Sensor

- Sensor data series paired with lab samples (within 5 mins)
- All 100 iterations of lab data used to calculate error statistics for sensor data → uncertainty cascaded through analysis

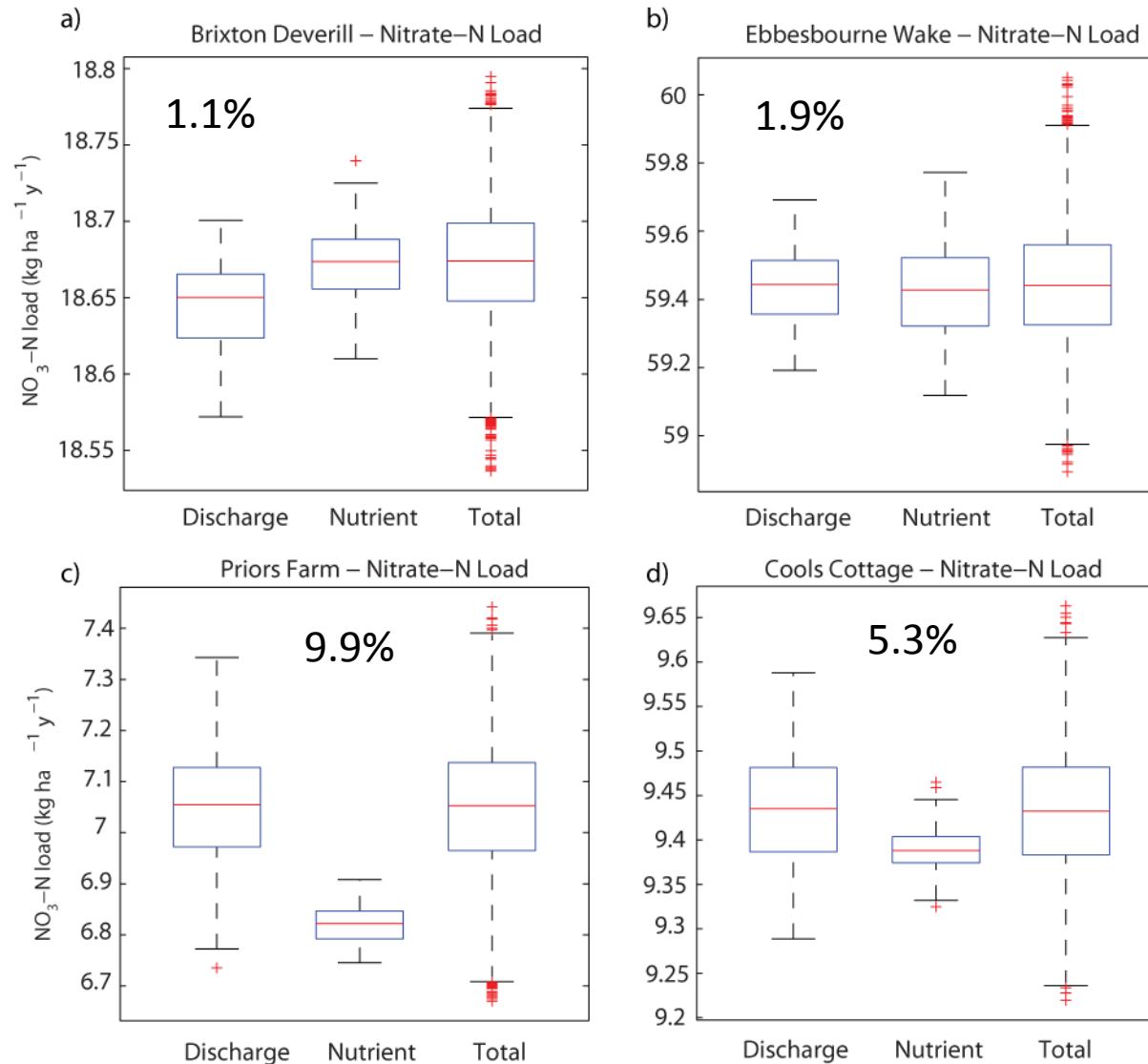




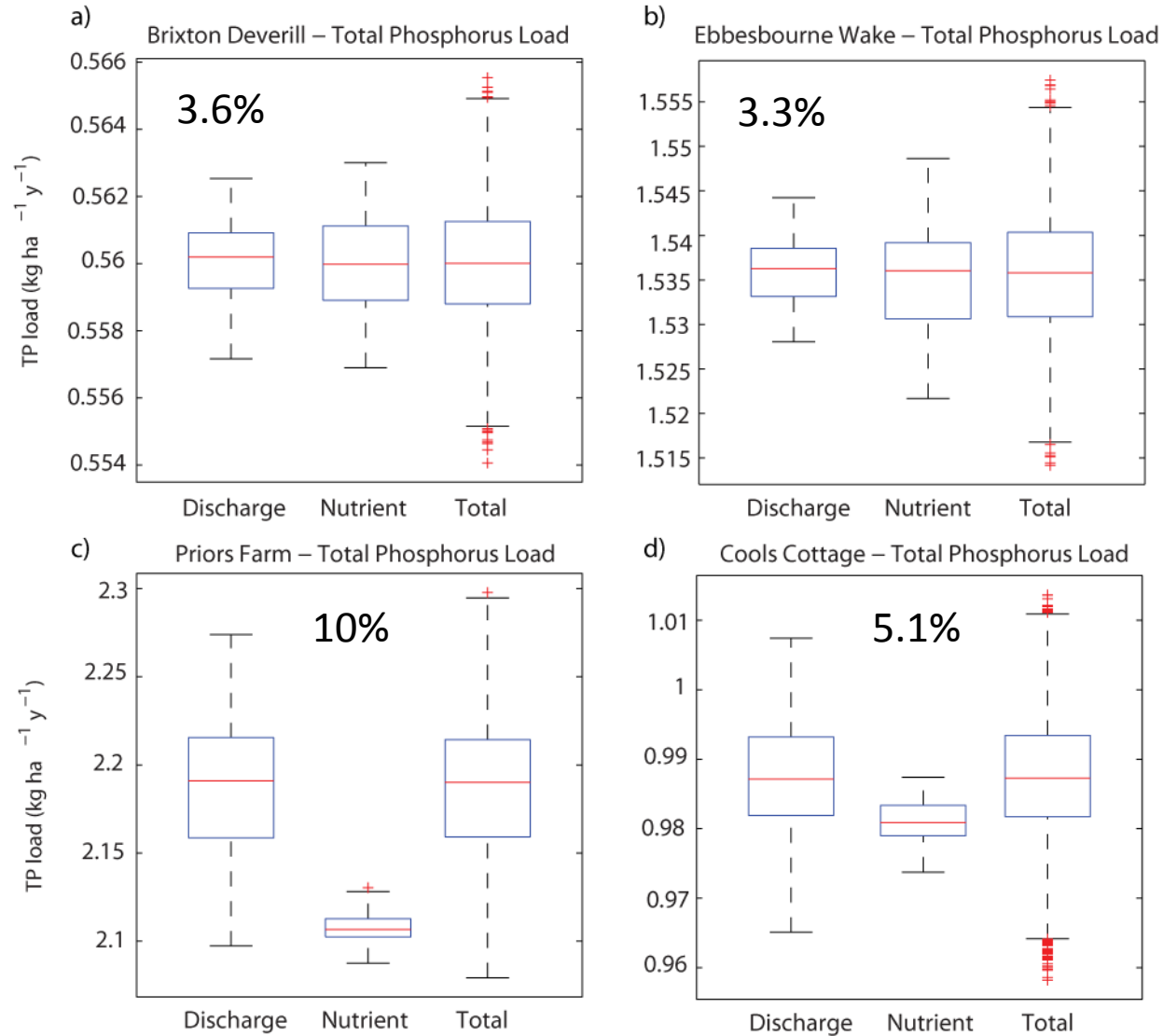
# Nutrient uncertainty- Sensor



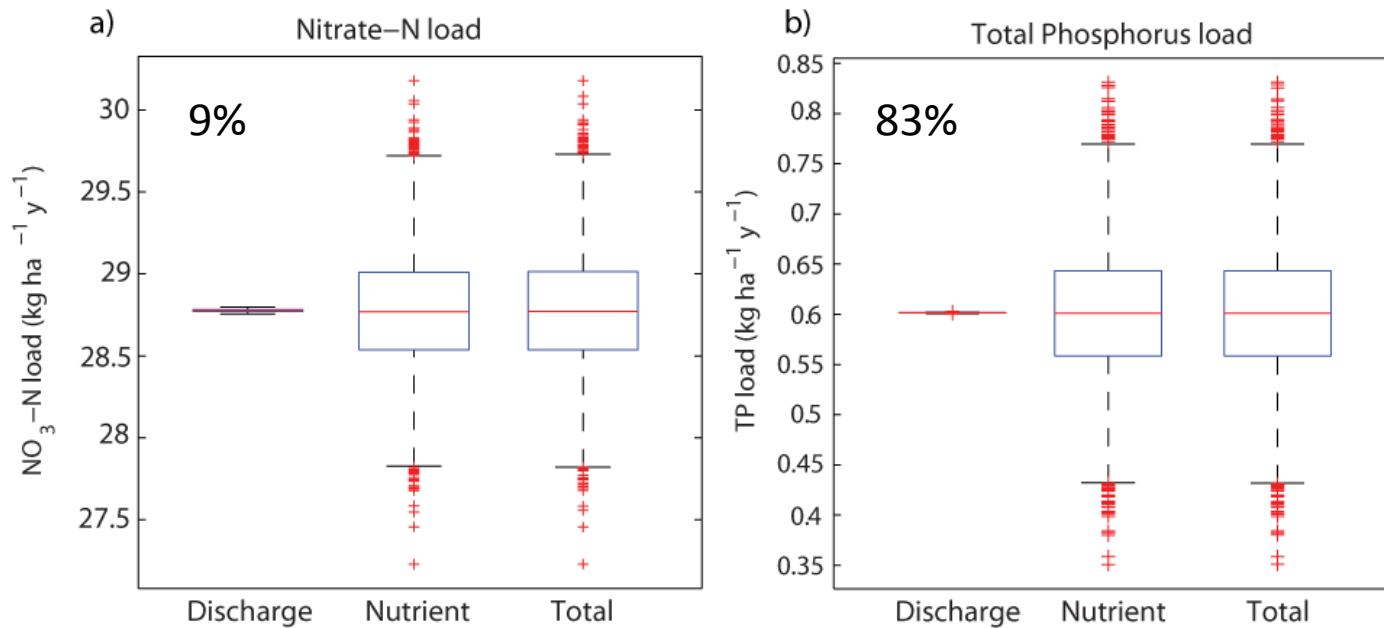
# Impact on load estimation



# Impact on load estimation



# Impact on load estimation - sensor



# Summary

- Using sensors to gain new insights into catchment processes
- Many challenges associated with sensor data collection and management
- Challenge the notion that high-resolution data is always better
- Highlights issues associated with the use of single rating curves
- Proposed a new methodology using non-parametric techniques which are more flexible
- Allows heteroscedasticity and temporal correlation to be taken into account
- Which is better: frequent but fuzzy data or infrequent and precise?