

Applying Deep Learning to Hydrological Events

Scott Hamshaw, P.E., Ph.D.

24-May-2018

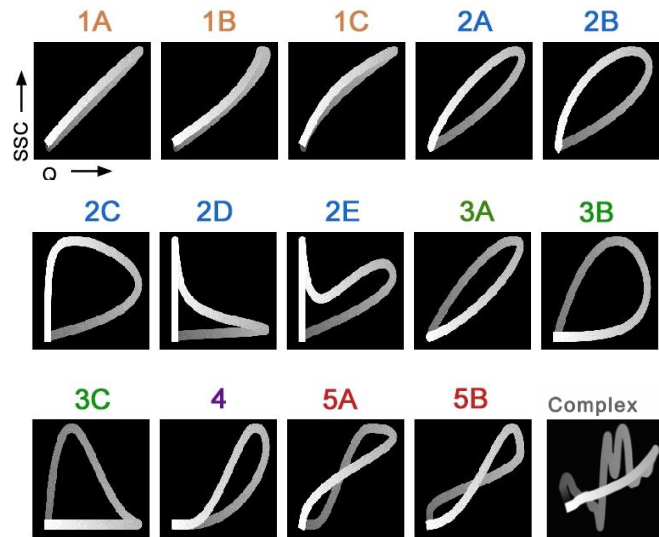
BREE PTAC Meeting



The University of Vermont

Key points from analysis of event hysteresis

- *Untapped potential in data-mining high-frequency water quality sensor data*
- *Can improve constituent load estimates and guide watershed modeling*
- Expanded library of hysteresis patterns



Understand watershed processes

- Sediment sources
- Transport dynamics

Automated Monitoring/Classification

- Shifts in types of events
- Detect key types of events

Research directions and integration into modeling

Event Analysis



- Improved TSS and TP Load Estimates
 - ▣ Regression models
 - ▣ ANN models

Watershed Hysteresis Characterization



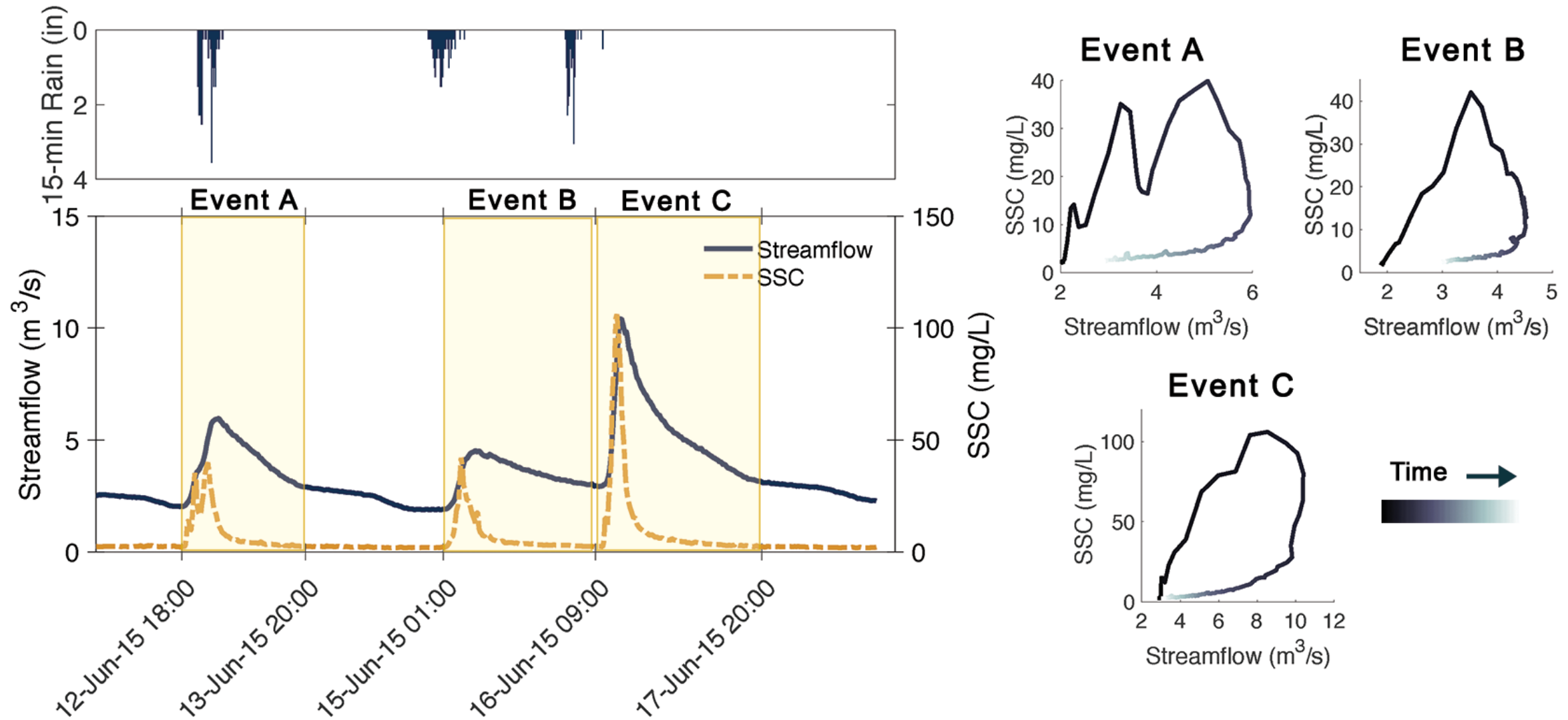
- Inform governance or land use models
 - ▣ Pre-condition map of watersheds to adjust project/BMP selection
 - ▣ Inform spatial cognition of agents

Automated Classification of Event C-Q hysteresis



- Apply to other response variables
 - ▣ DOC
 - ▣ Nitrate
 - ▣ Soil Moisture

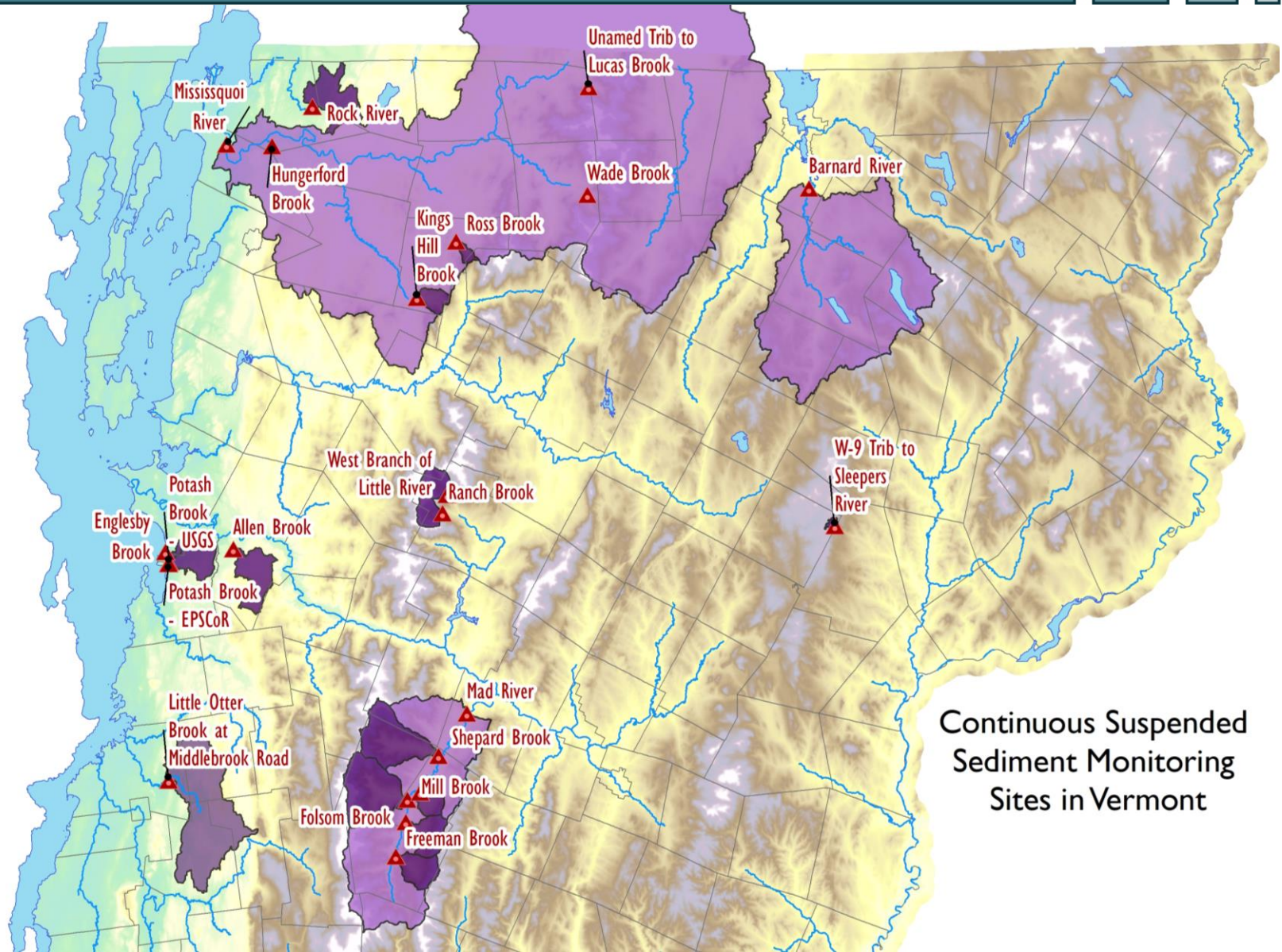
Using Hysteresis Analysis to Characterize Hydrological Events



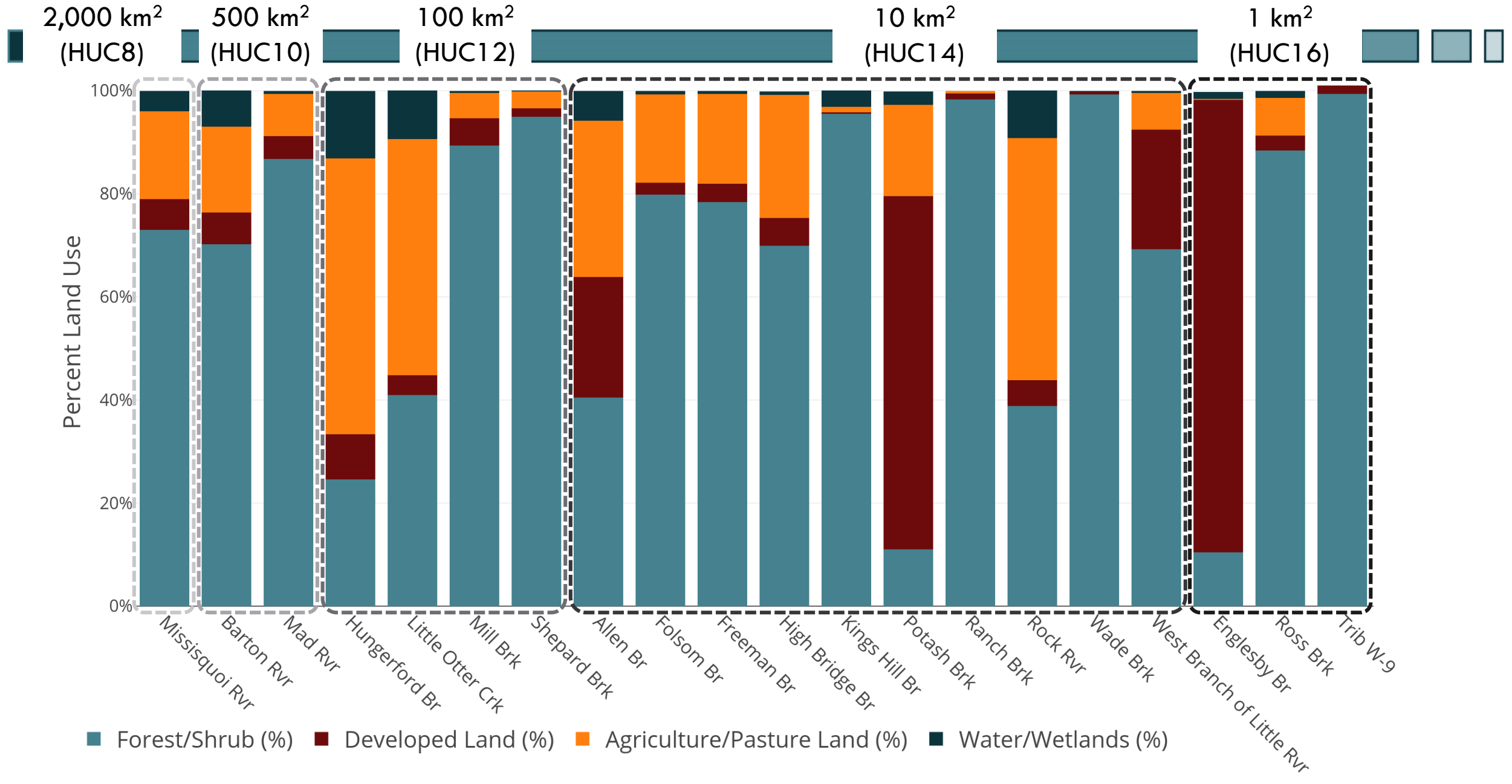
Expanding research out into new watersheds

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- Range of:
 - ▣ Land Use/Cover
 - ▣ Geology
 - ▣ Soils
 - ▣ Drainage Area
 - ▣ Topography



A more varied set of watersheds

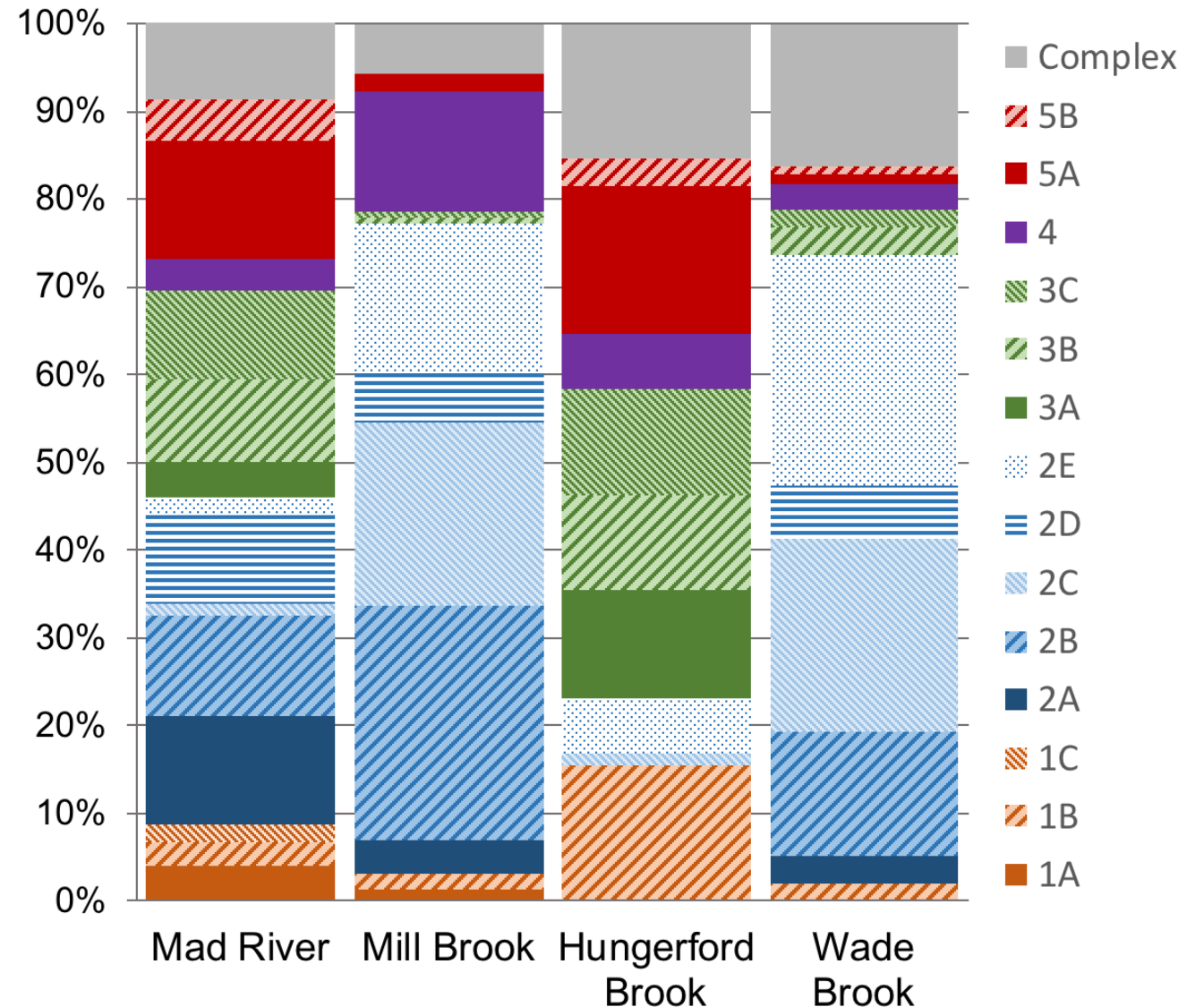
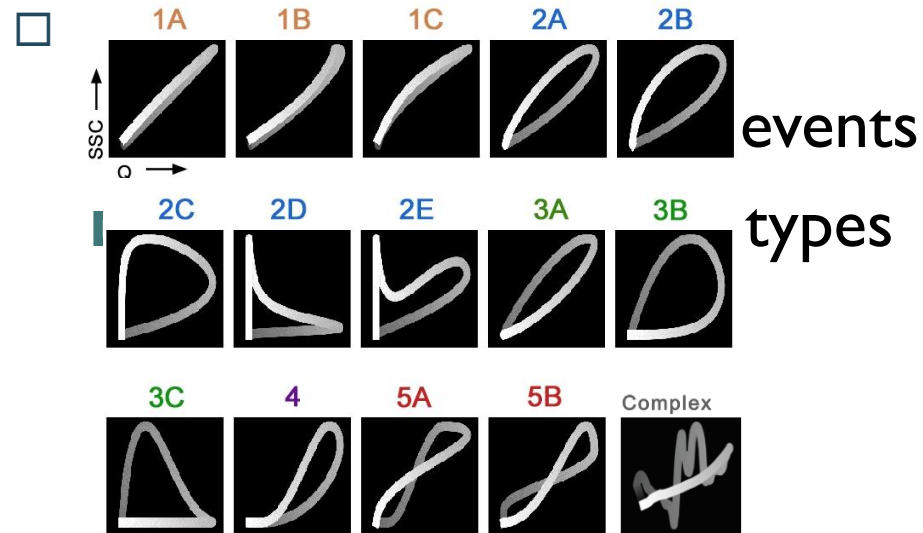


Clear differences in dynamics between watersheds

- Need to account for effects of:

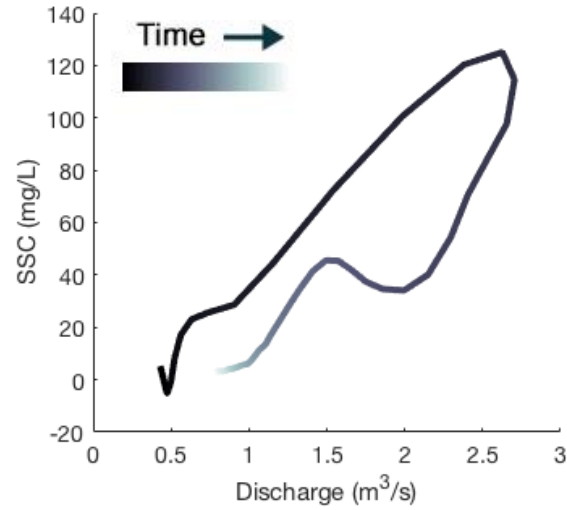
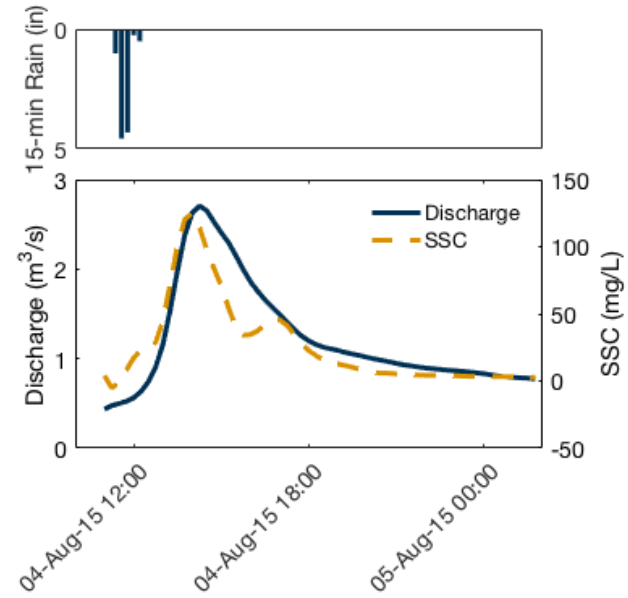
- Spatial Scale

- Season

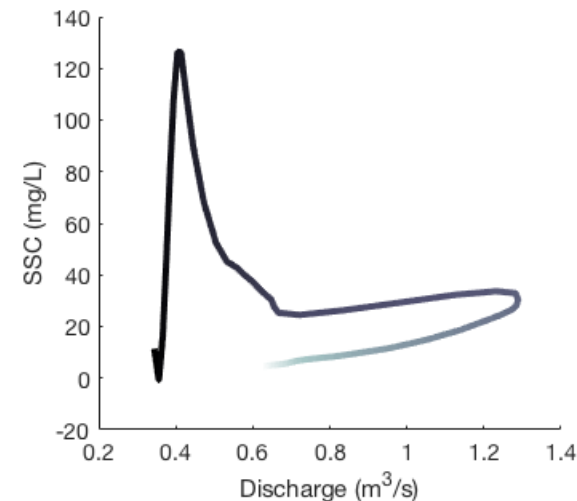
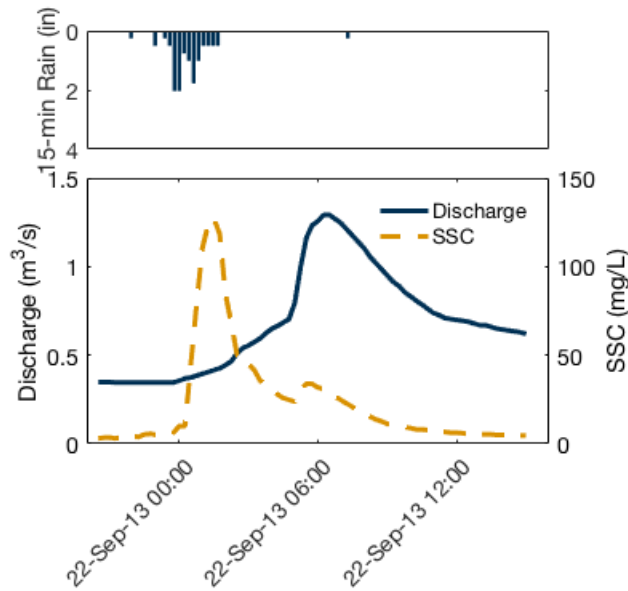


An Example: Two storm events to illustrate event sediment dynamics

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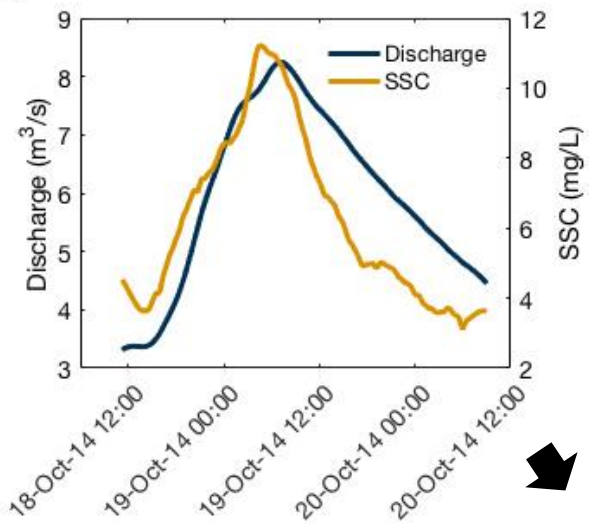


- Streamflow activated (channel network) sediment sources important

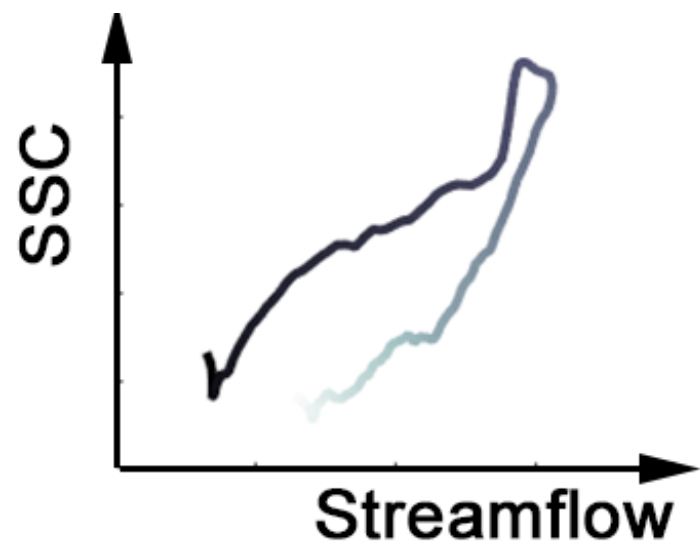


- Connected, rainfall activated, nearby sediment sources important

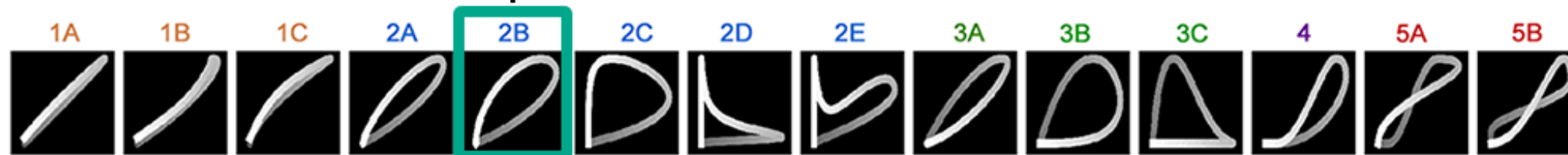
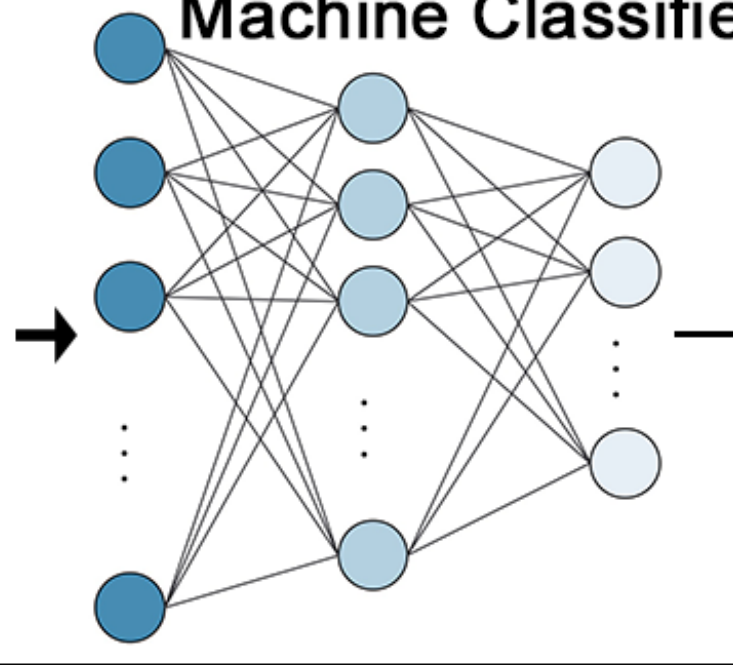
a)



Automated event classification system



Restricted Boltzmann Machine Classifier



Implementing Deep Learning into hydrological event analysis

□ Model algorithms & architecture

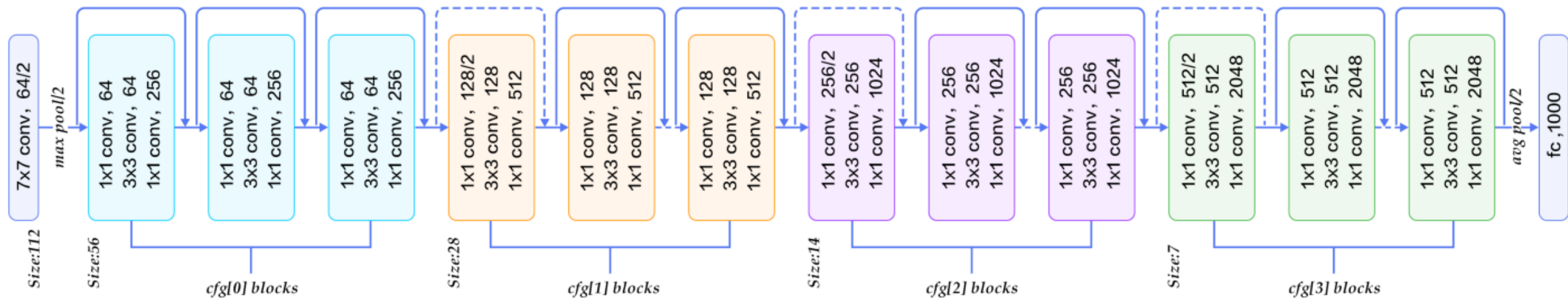
■ Convolutional Neural Networks (CNNs)

■ 3-D CNNs

■ Autoencoders

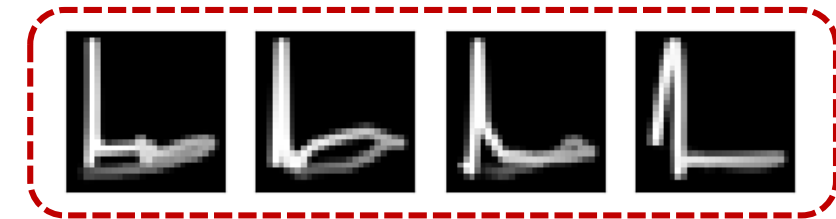
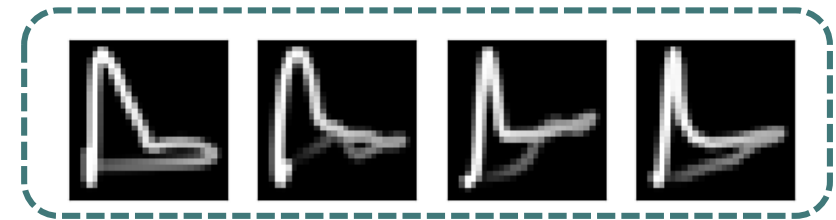
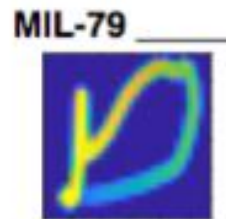
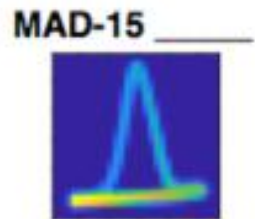
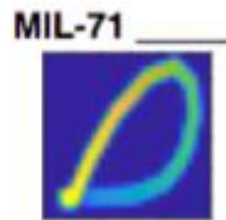
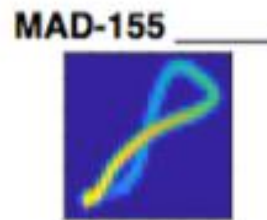
- Increase in accuracy over previous results
- Near 70% classified correctly

ResNet50 Architecture



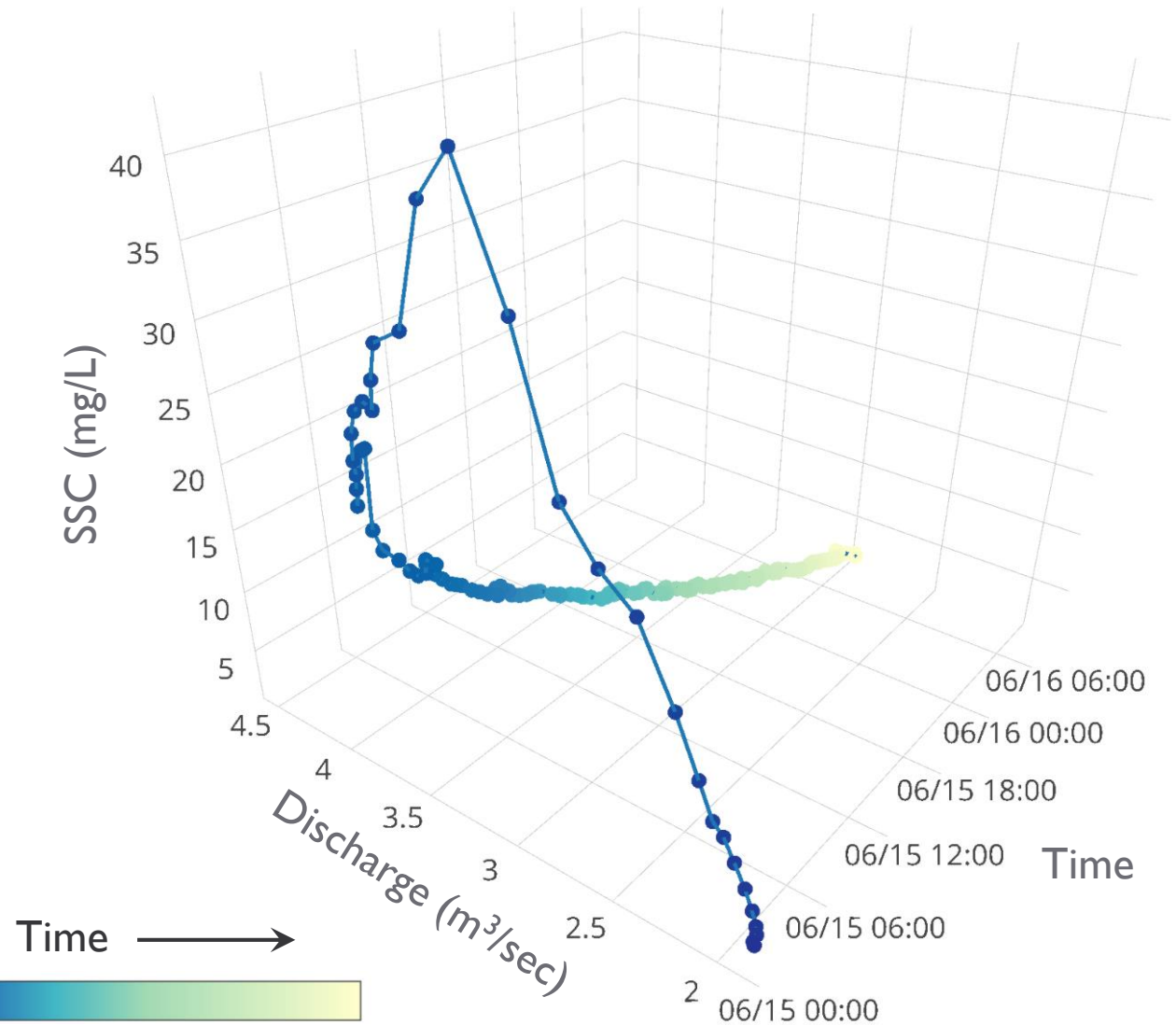
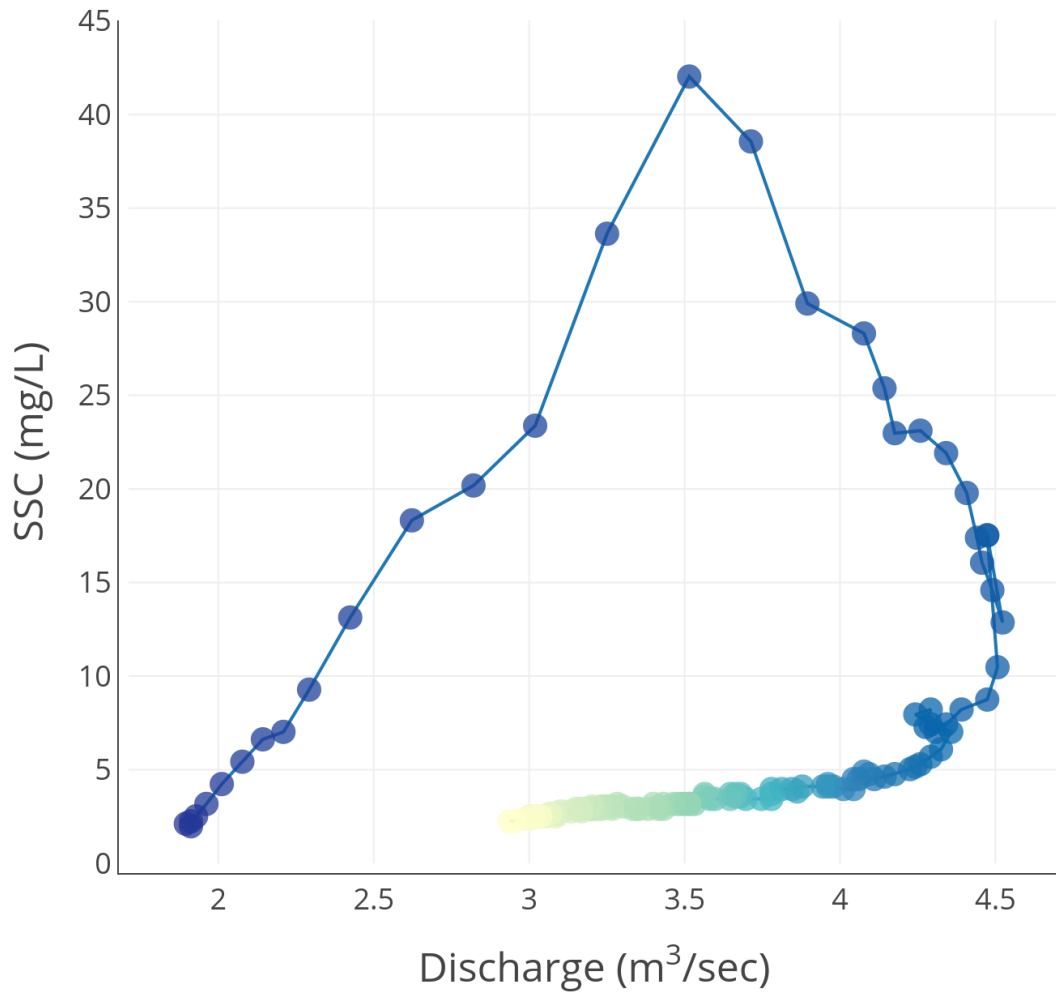
Implementing Deep Learning into hydrological event analysis

- Model algorithms & architecture
 - ▣ Convolutional Neural Networks
 - ▣ 3-D CNNs
 - ▣ Autoencoders
- New Classes (pattern library)
 - ▣ Clustering of encoded features
 - ▣ Crowdsourcing tests



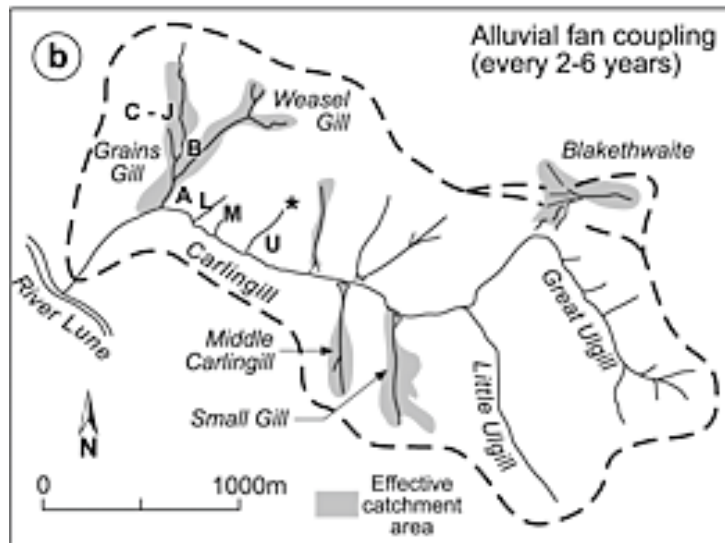
Challenge: very data hungry methods!

2-D vs 3-D “Trajectories” of Events



Continue work for testing hypothesis

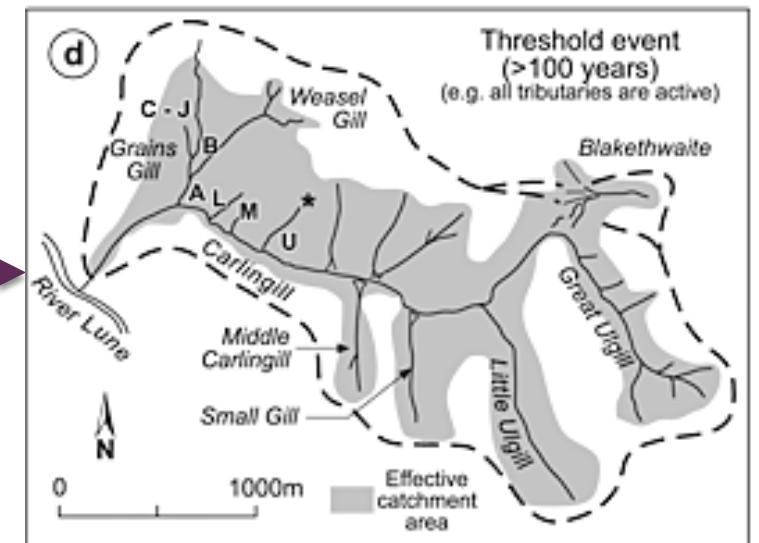
- C-Q plot (and their sequence) encodes information about where erosion is taking place in watershed and it's transport downstream



Fryirs, 2013 ESPL

VARIABLE

- Sediment Source Areas
 - Location
 - Supply
 - Connectivity
- Susp. Sediment Yield
- SS – Q Relationships

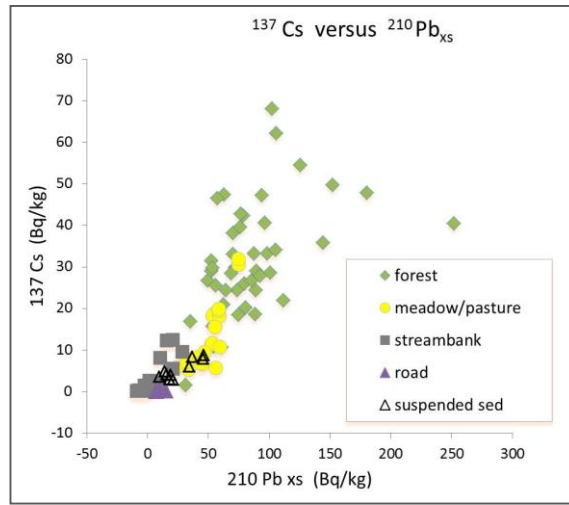


Fryirs, 2013 ESPL

How do we determine from where riverine sediments originate?

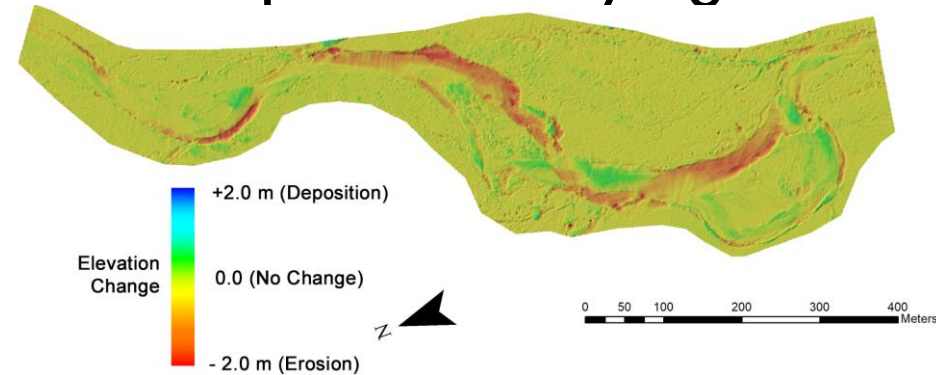
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□ Sediment Tracers

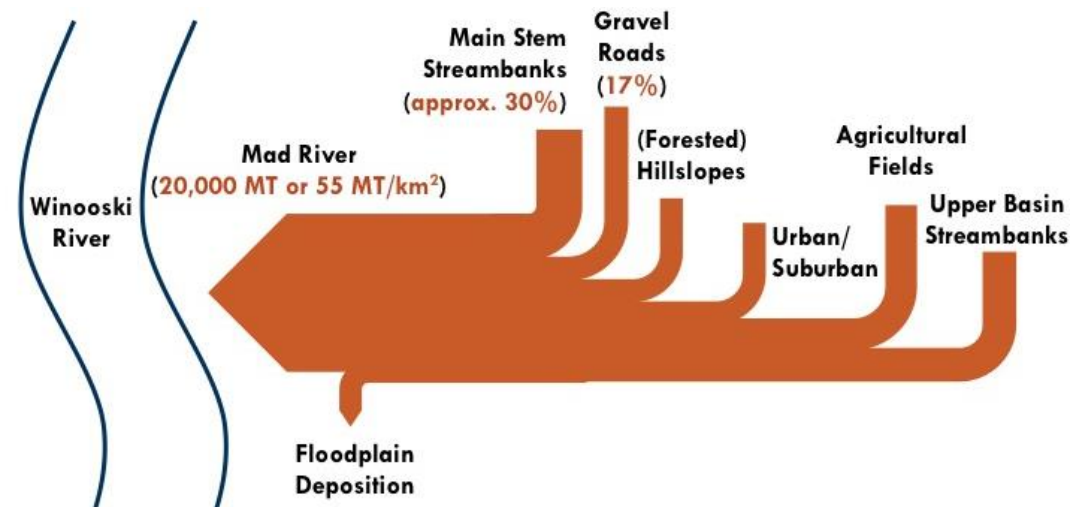


Kristen Underwood

□ Repeat Surveying

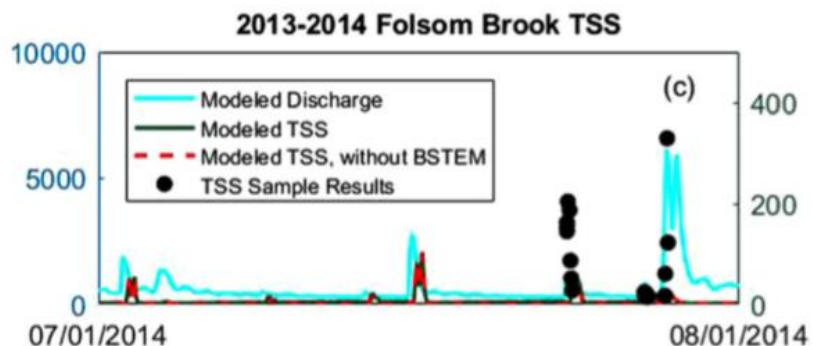


□ Sediment Budget

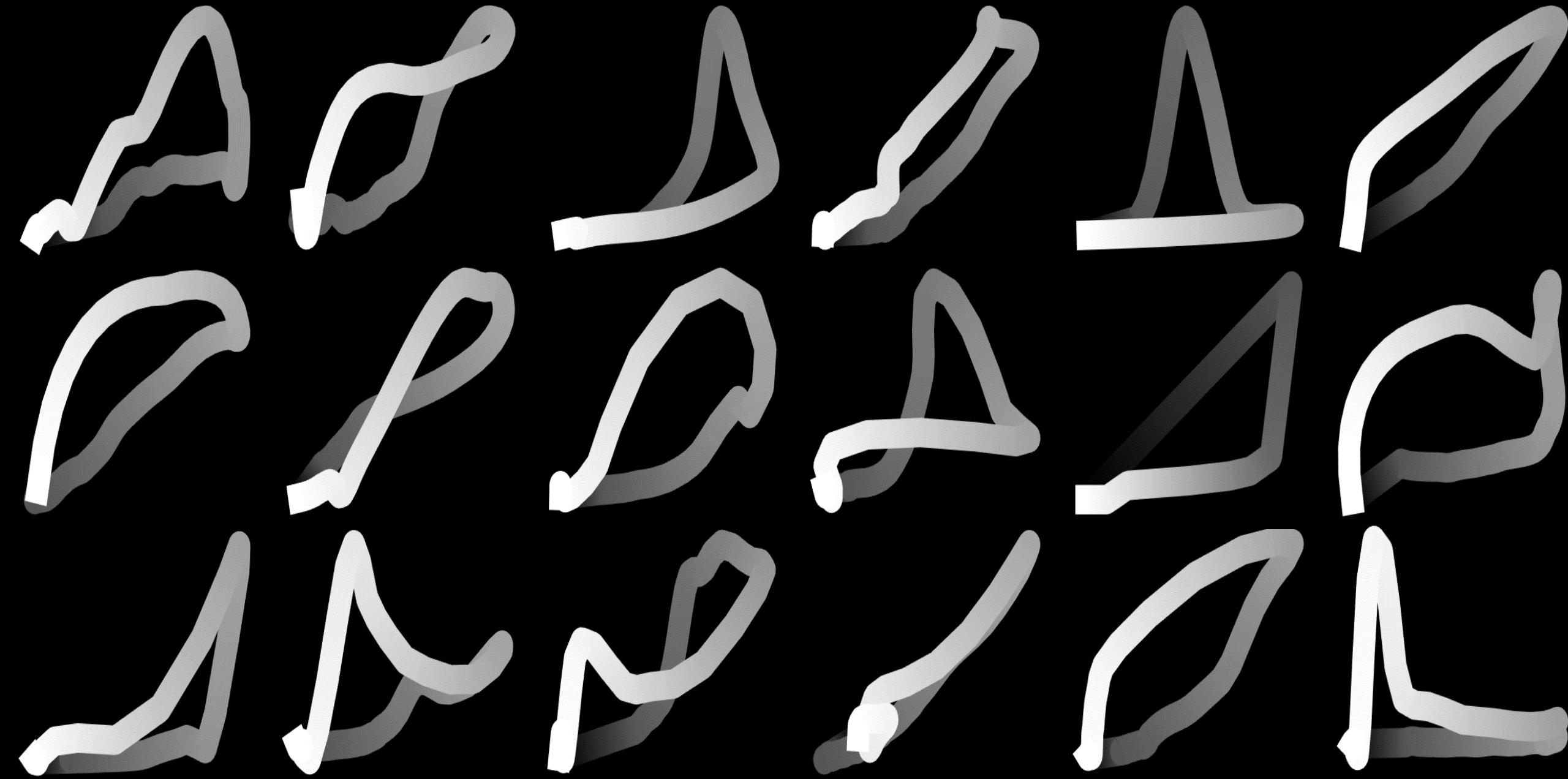


Stryker et al. 2017

□ Watershed Modeling



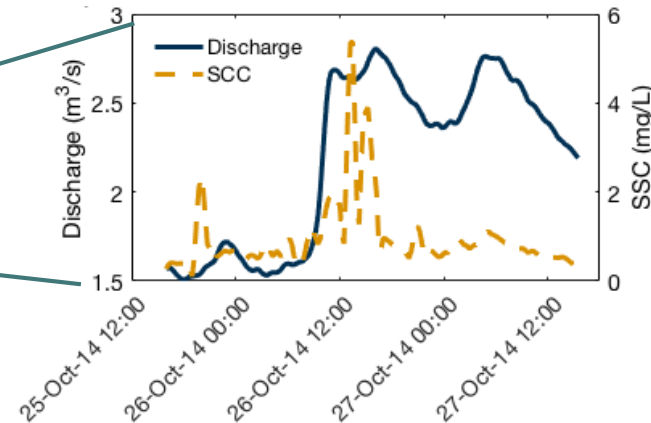
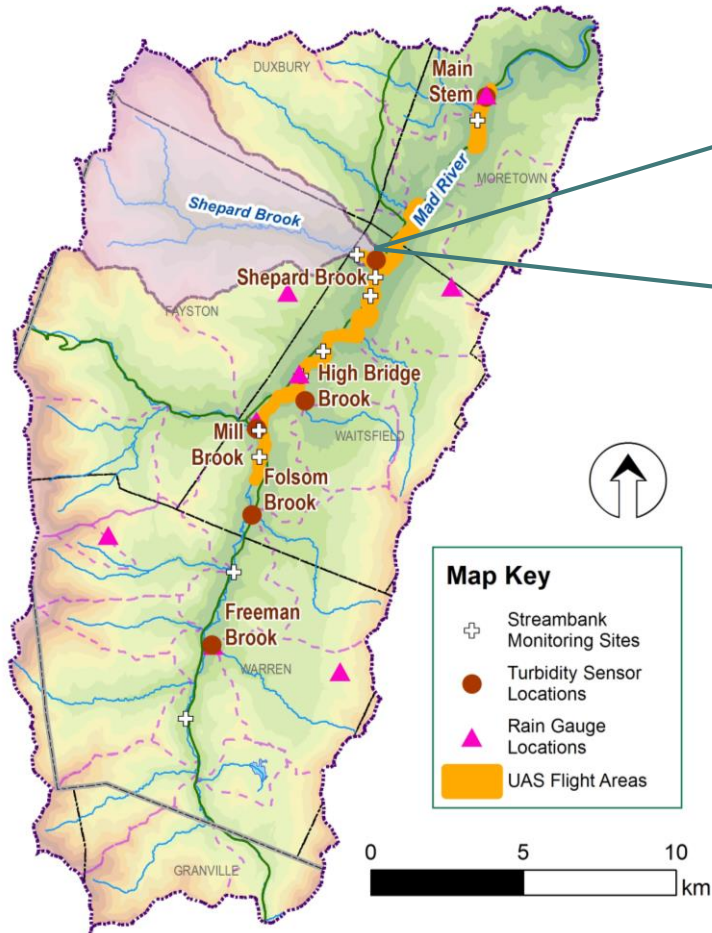
What if we let the watershed tell us what is going on?



What if we let the watershed tell us what is going on?

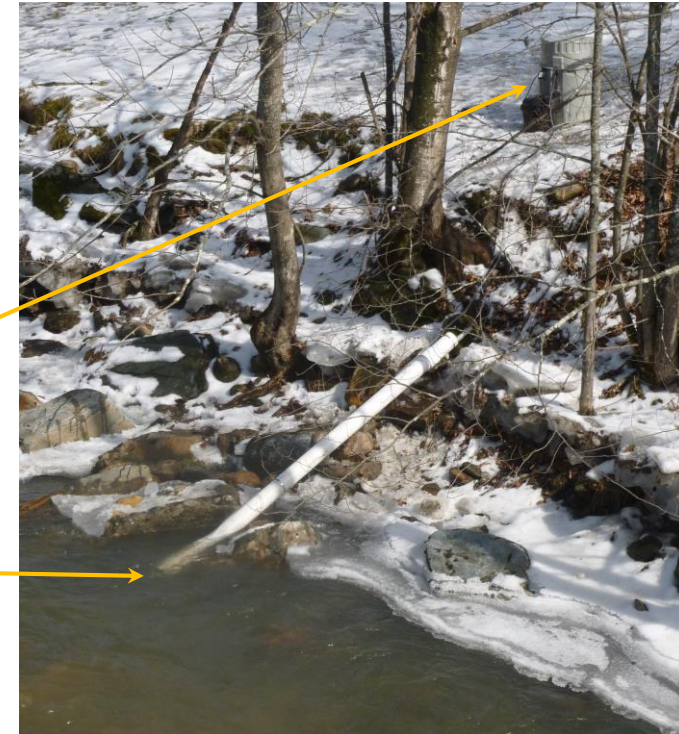
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- What if we could monitor only the outlet of the watershed and be able to infer sediment dynamics within the watershed?



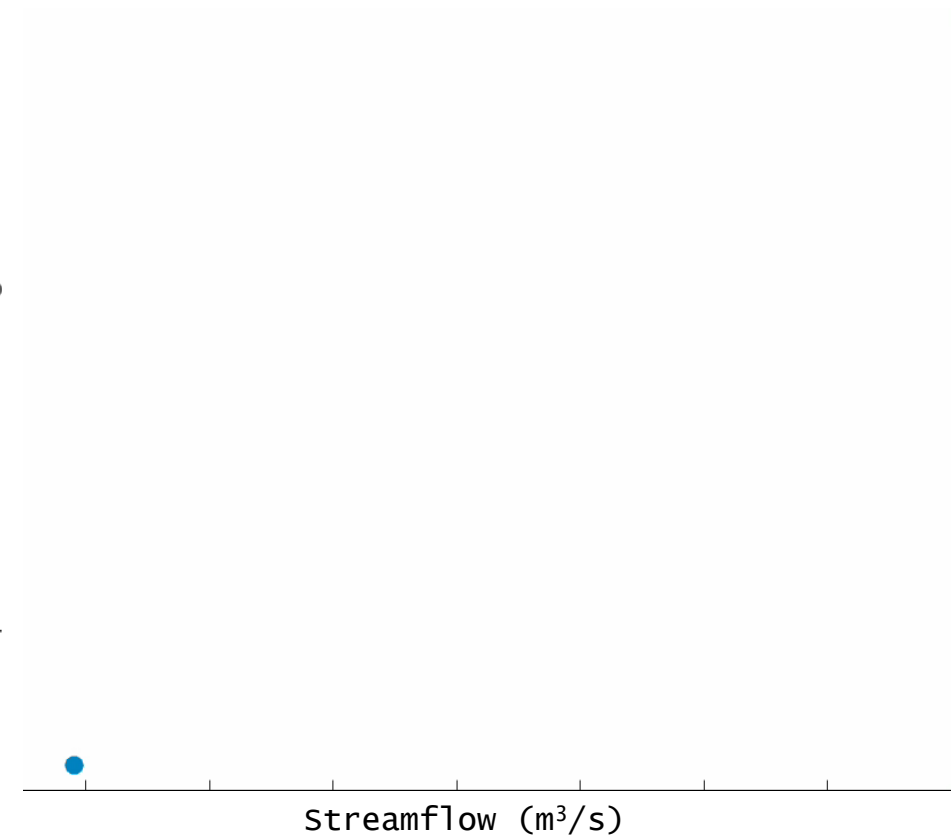
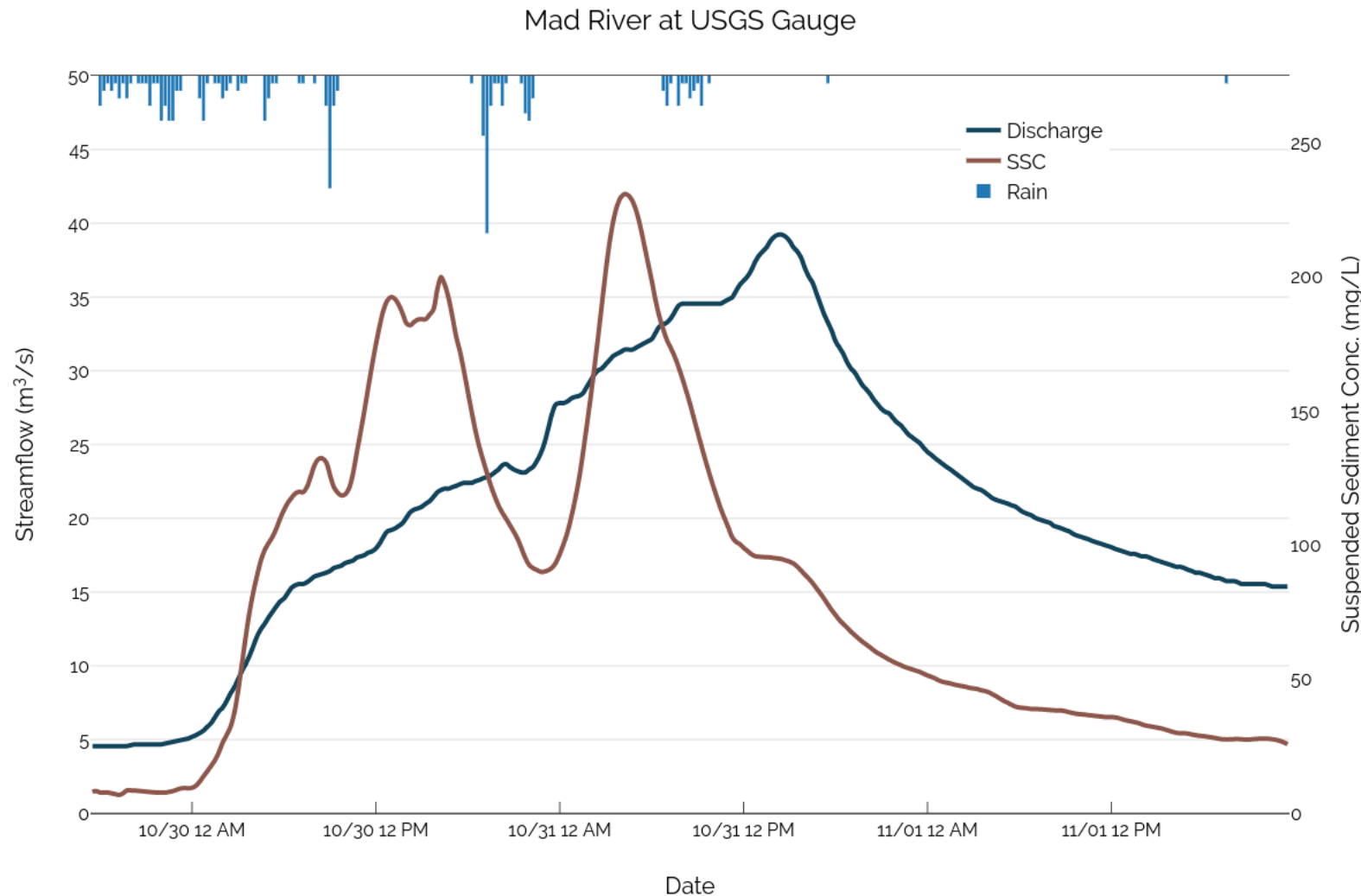
ISCO Autosampler
and Datalogger

DTS-12 In-situ
Turbidity Sensor



A close look at hydrological events

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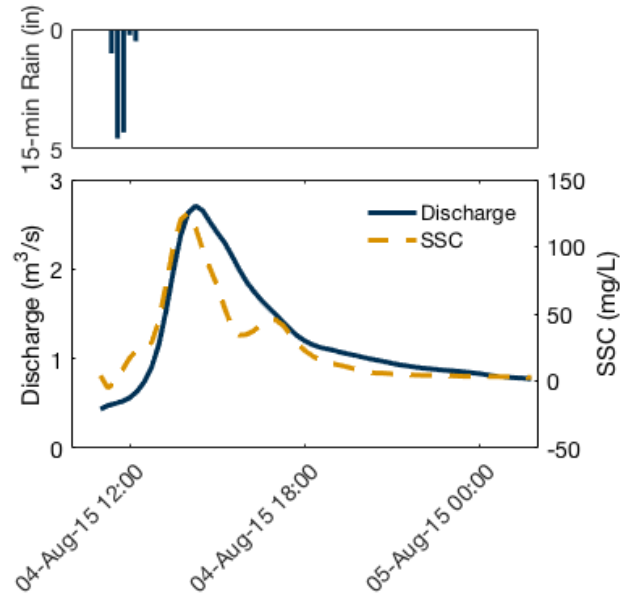


An Example: Two storm events to illustrate event sediment dynamics

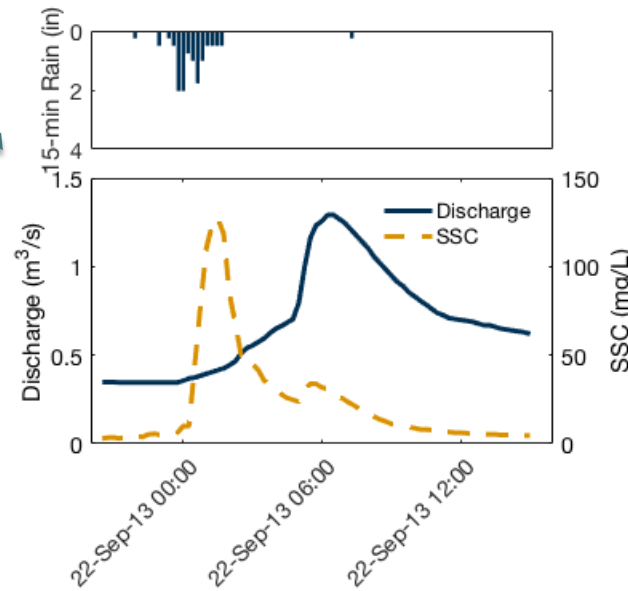
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□ Shepard Brook

■ Aug 4, 2015



■ Sep 22, 2013

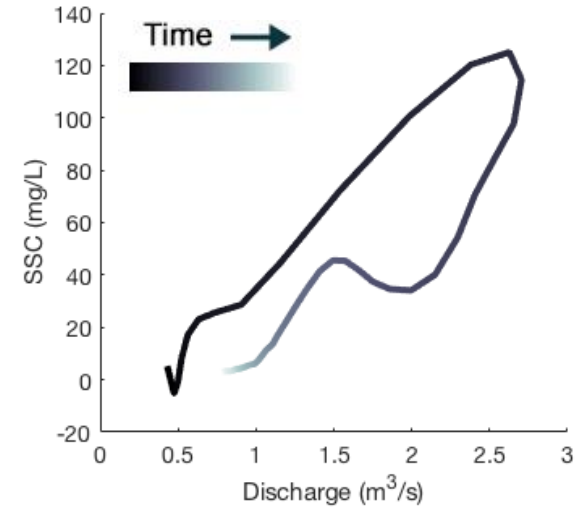
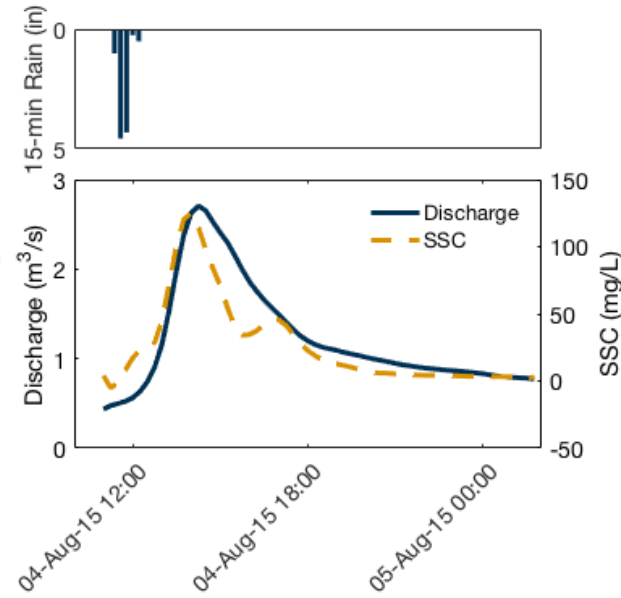


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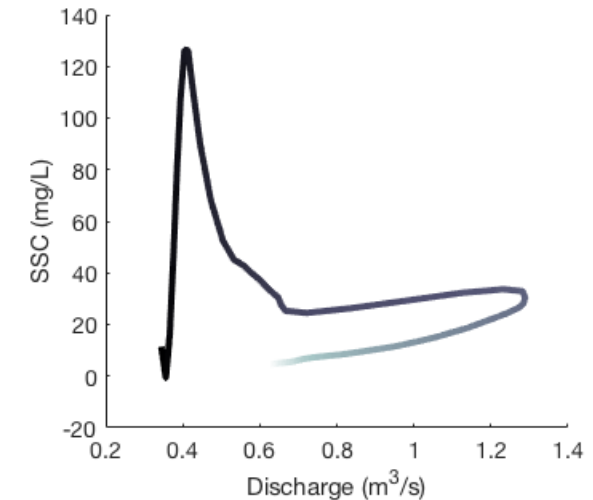
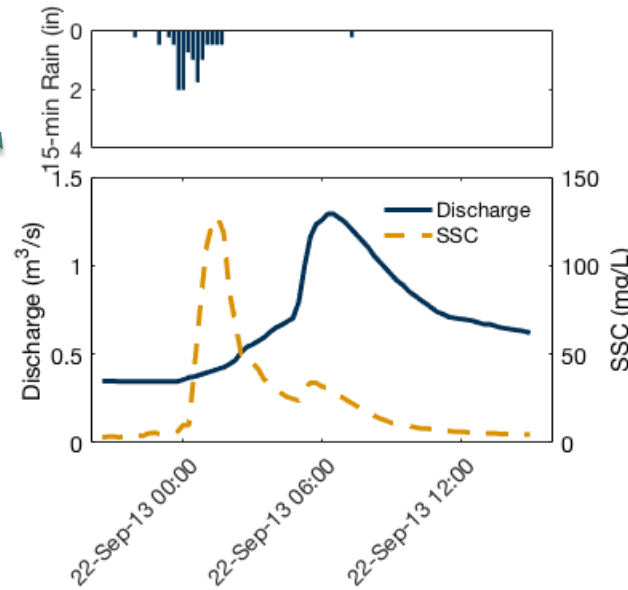
20

□ Shepard Brook

■ Aug 4, 2015



■ Sep 22, 2013



What are hysteresis patterns?

Two methods of categorizing hysteresis

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□ Visual Patterns

Class I - Linear



Class II - Clockwise

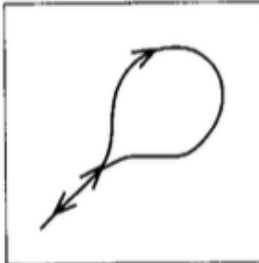


Garnett
Williams,
USGS,
1989

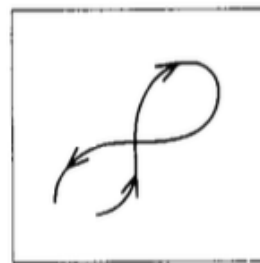
Class III -
Counterclockwise



Class IV - Linear
then Clockwise

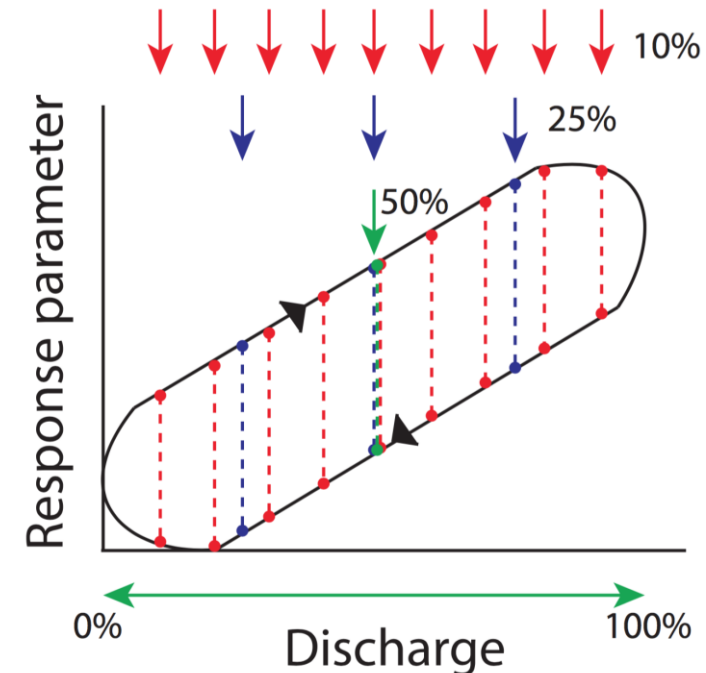


Class V -
Figure-Eight



□ Metrics (e.g. Hysteresis Index)

$$HI = T_{RL} - T_{FL}$$

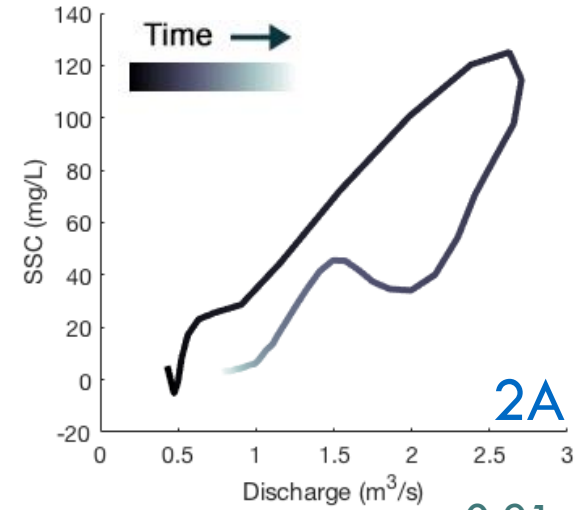
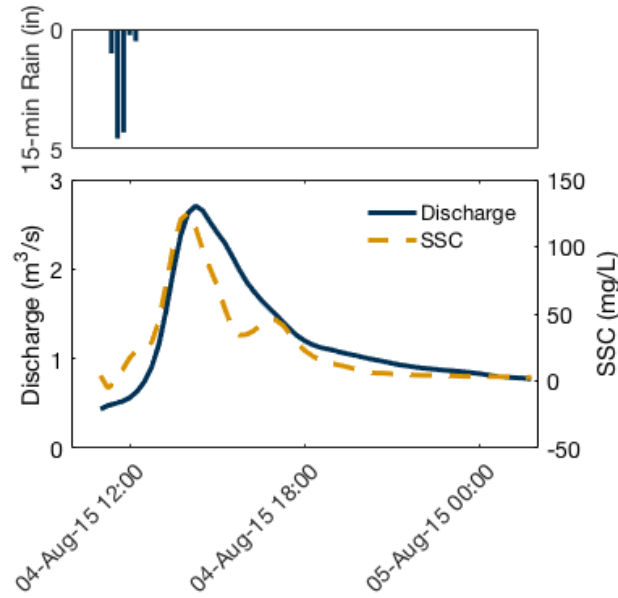


Lloyd et al. 2015

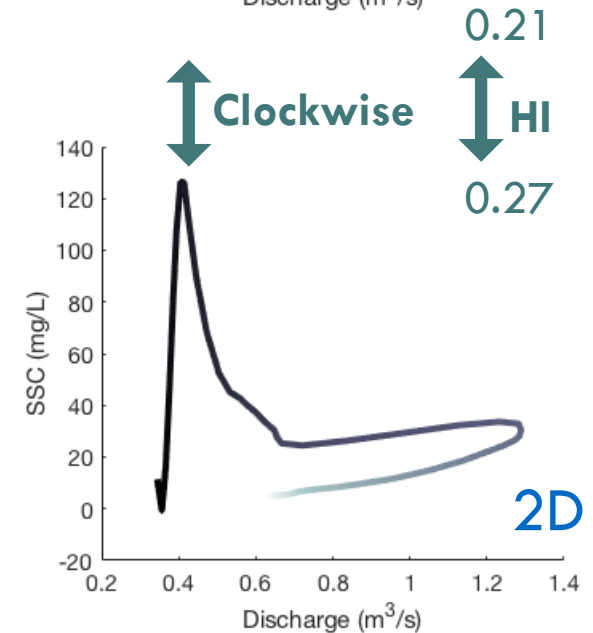
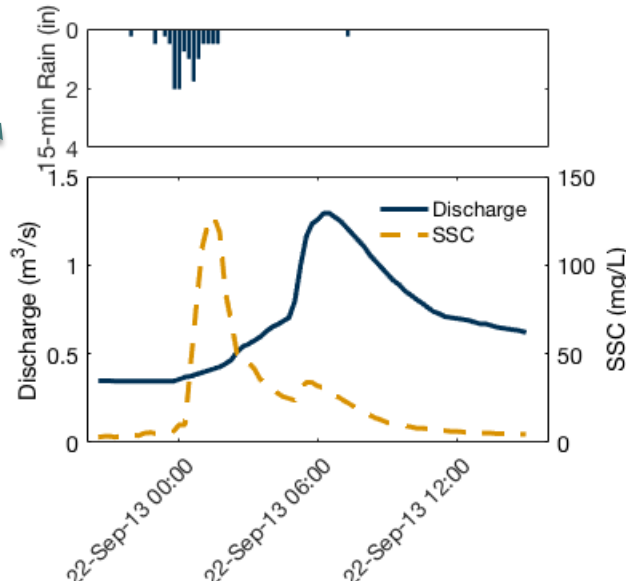
An Example: Looking back at the two storm events

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- 2 storm events
Shepard Brook
- Aug 4, 2015



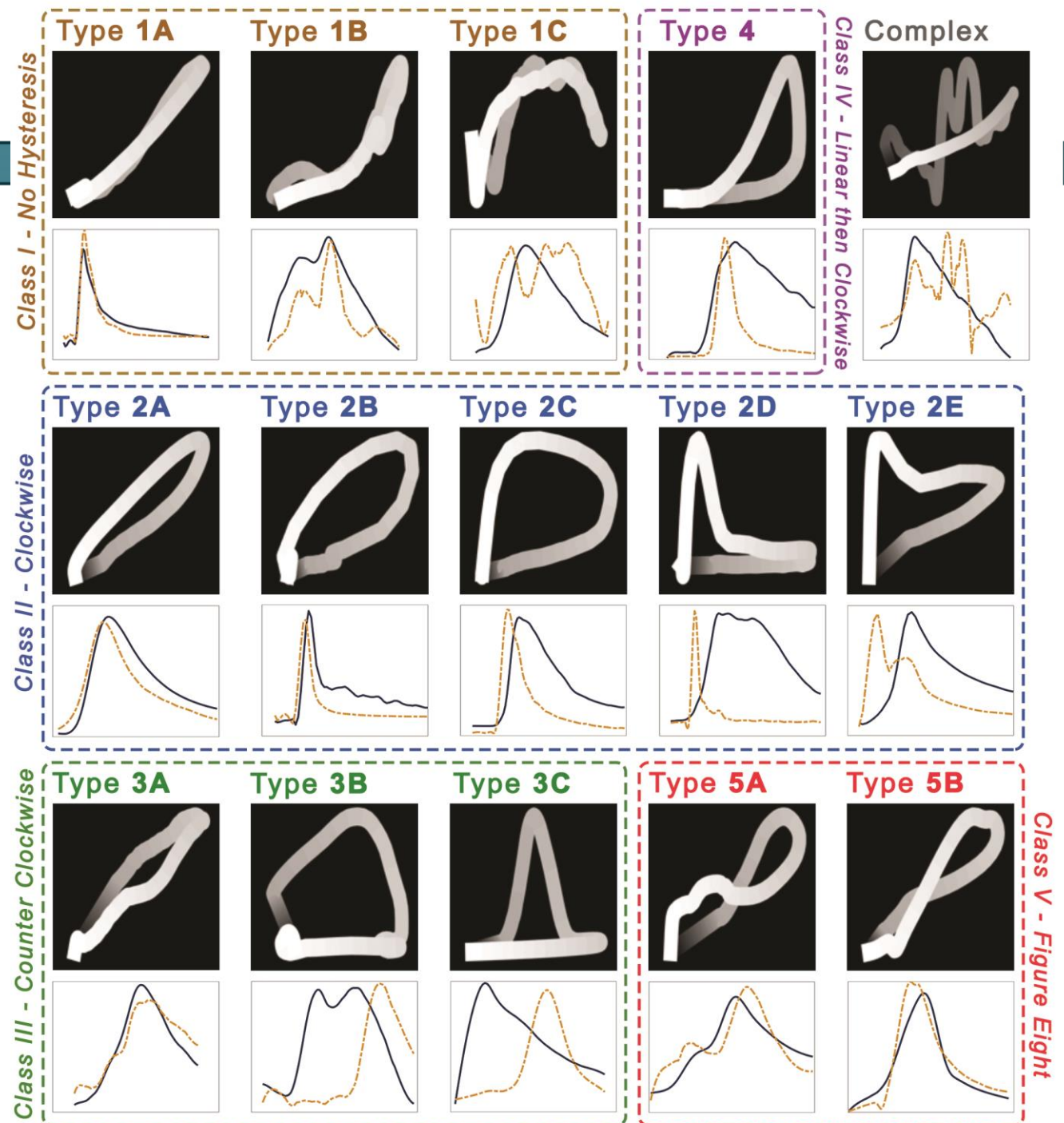
- Sep 22, 2013



Patterns of Hysteresis

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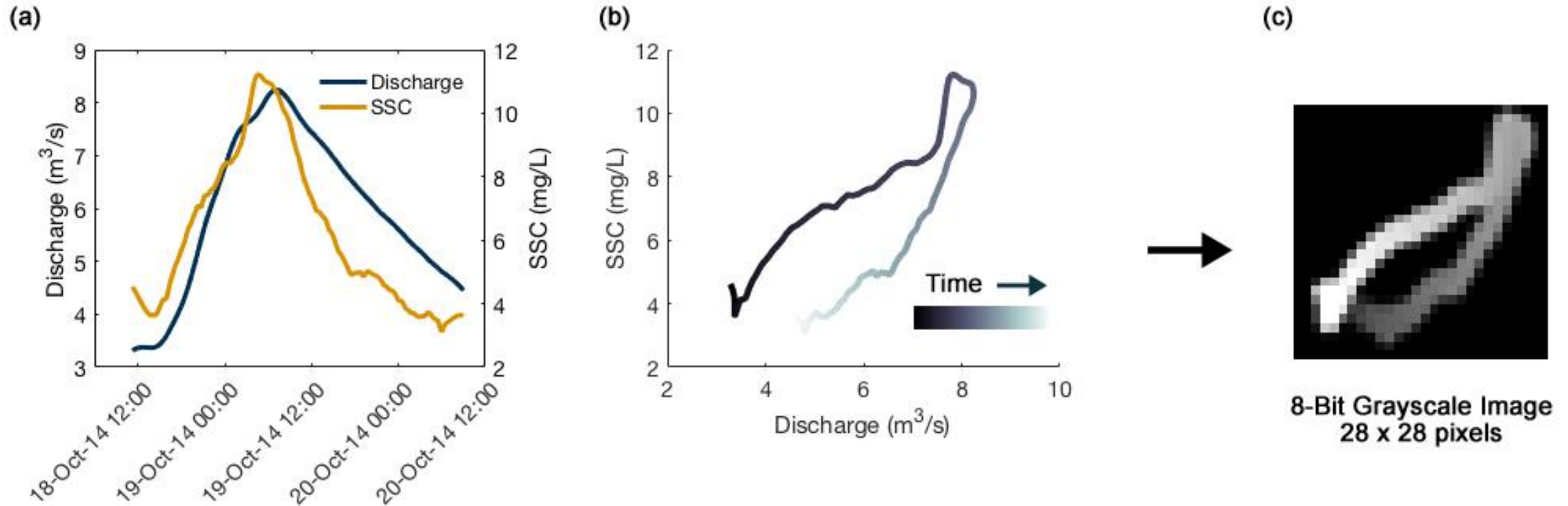
- 14 Types recognized in data from Mad River watershed
- How to automate?



An automated classification system

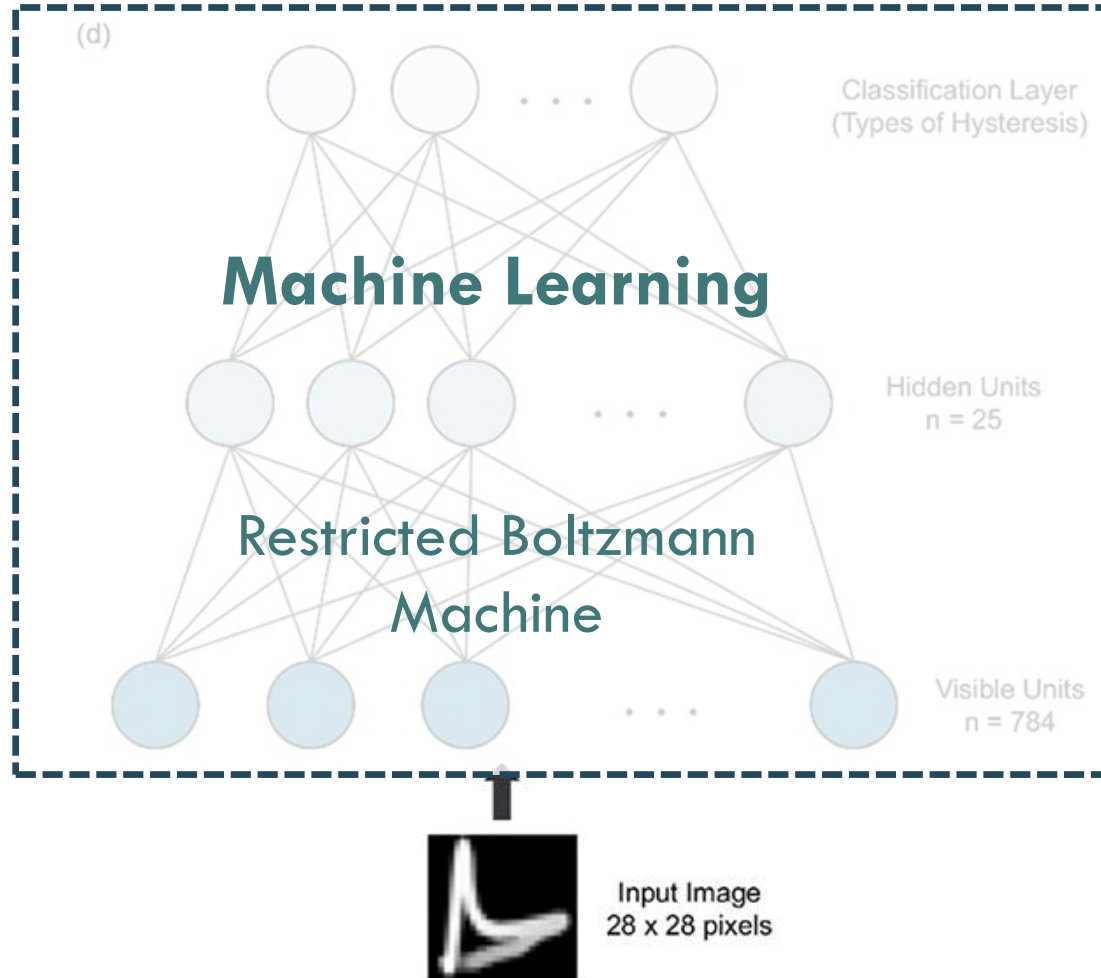
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□ Pattern recognition challenge



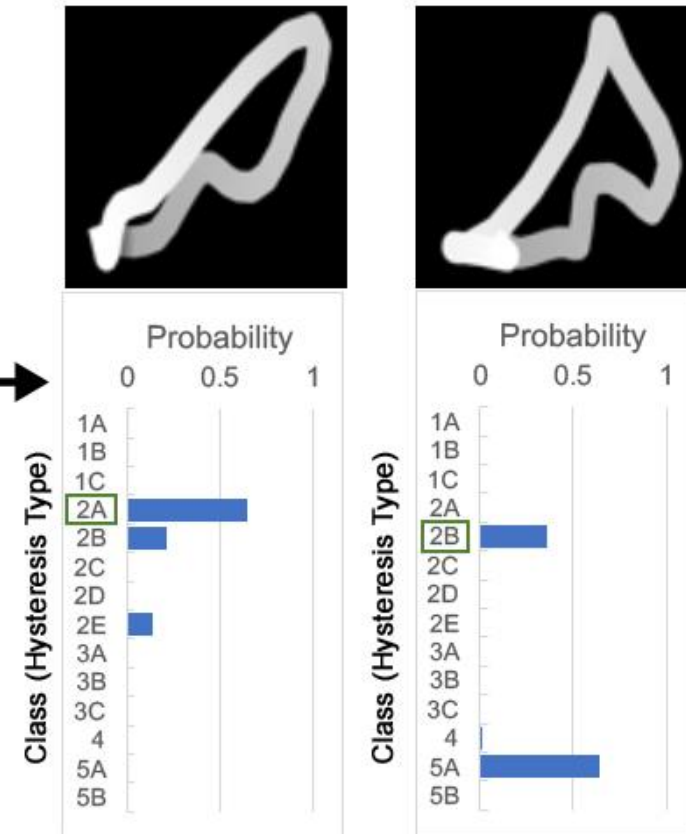
Example of classification of storm events

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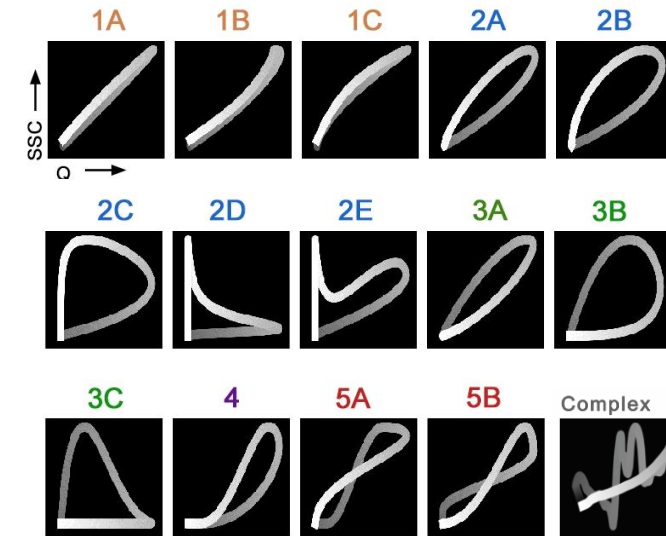
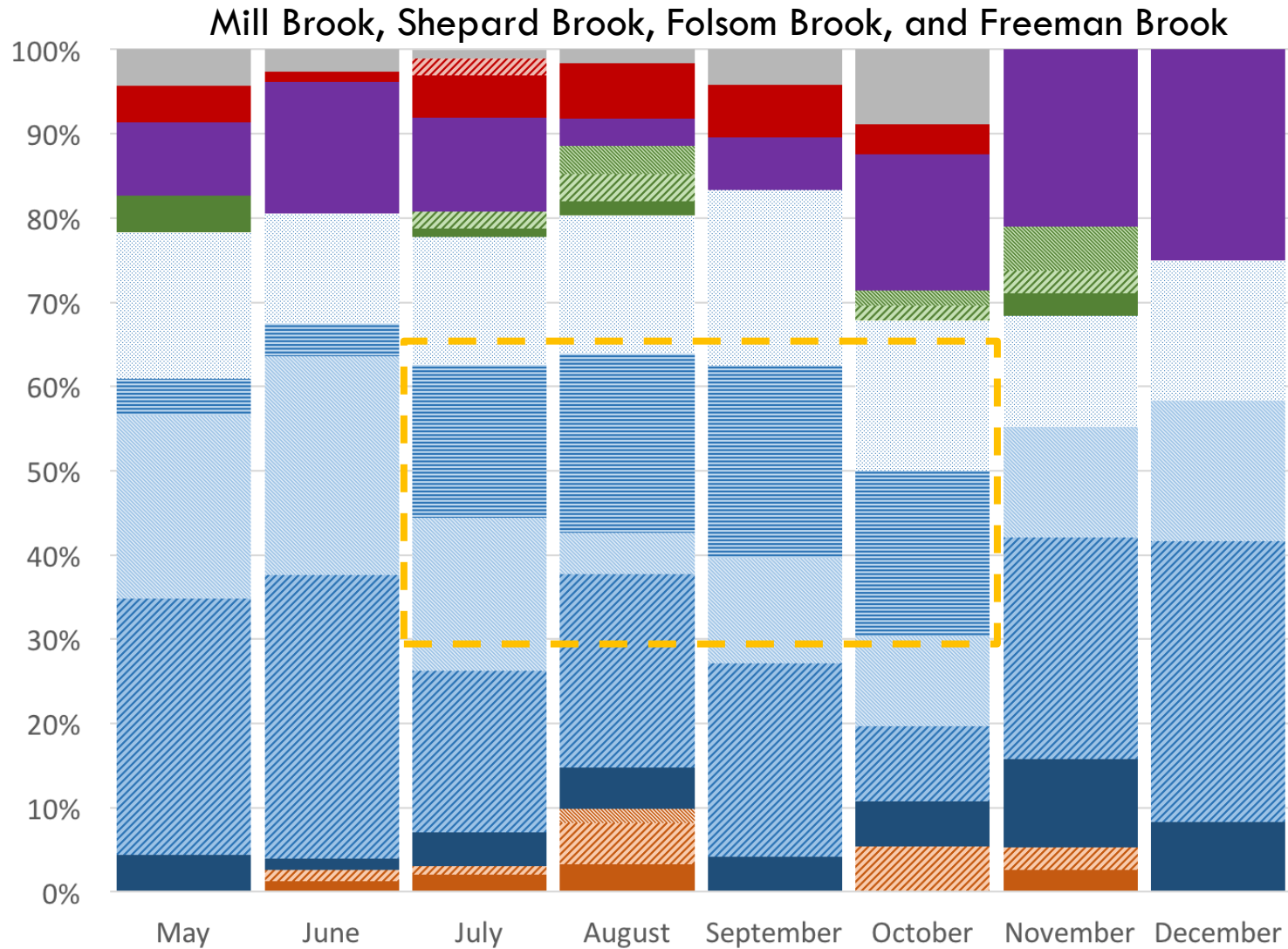
(e)

Example Event Classifications



Seasonal trends in hysteresis types

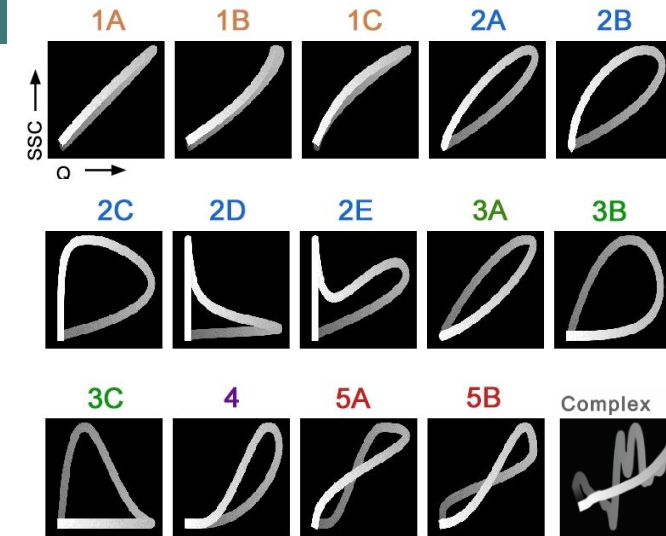
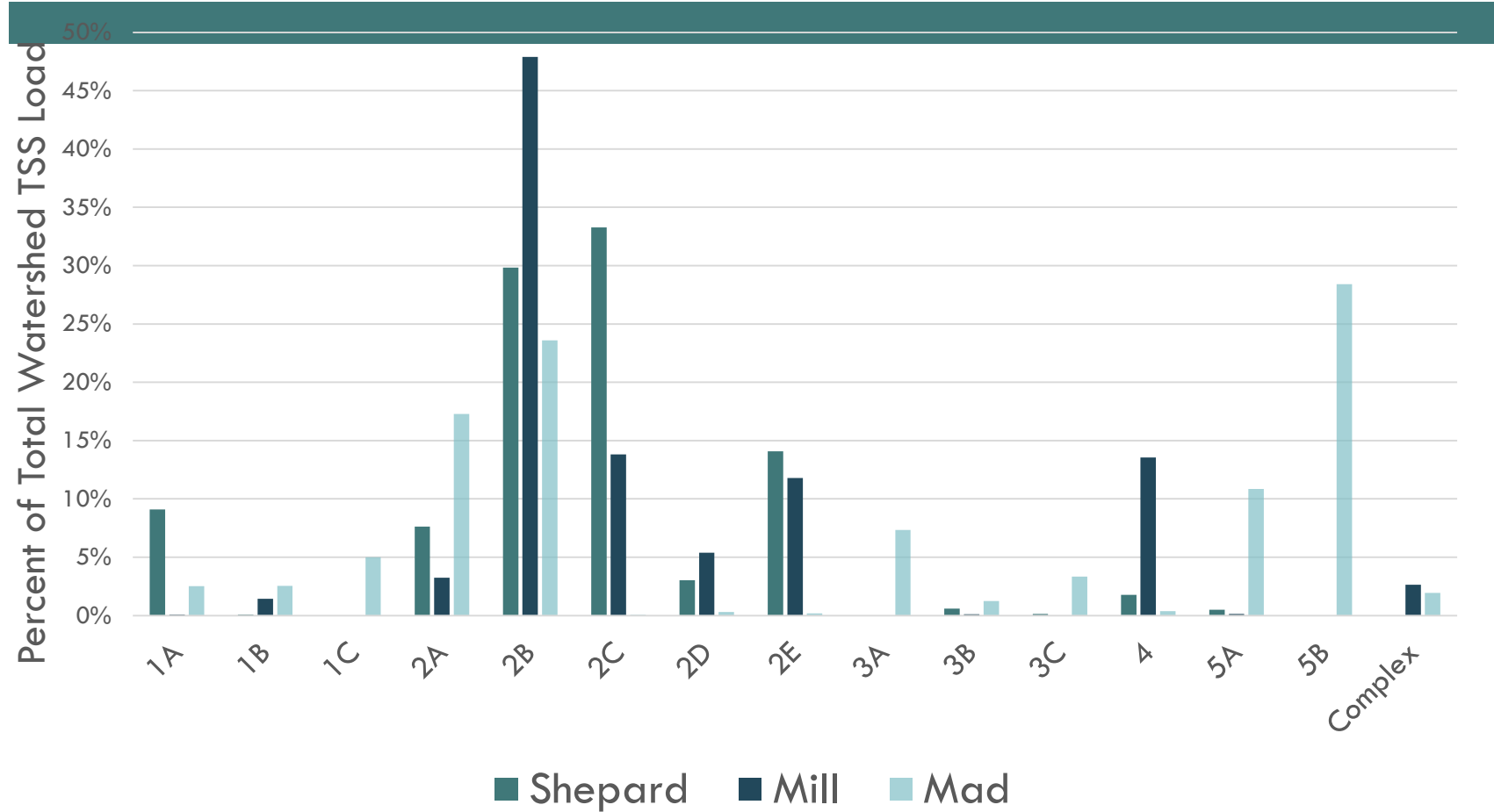
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Also identified trends in hysteresis patterns by:

- Site
- Drainage Area Size
- Sediment Load

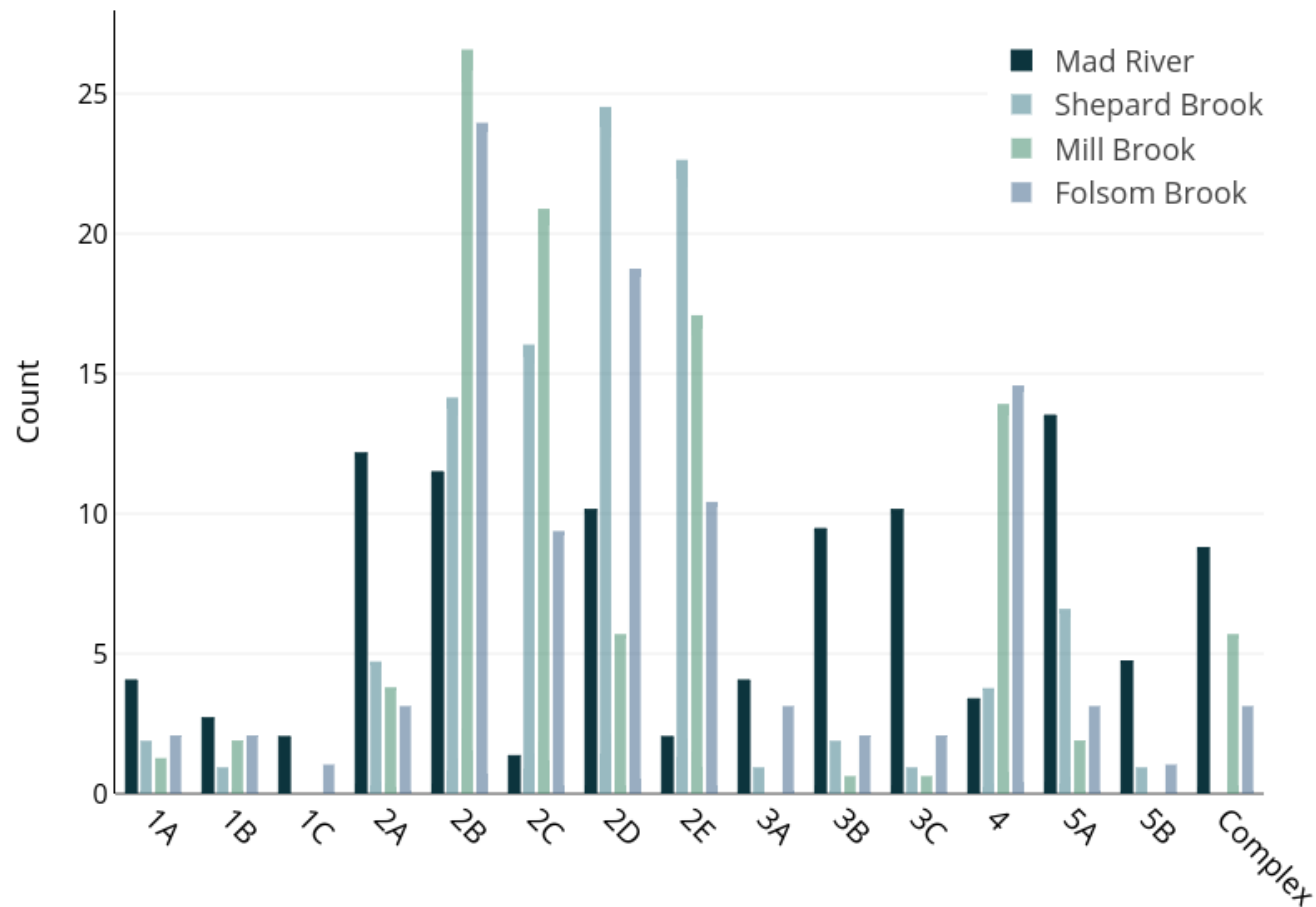
Sediment load by hysteresis type



Effects of spatial scale on hysteresis type

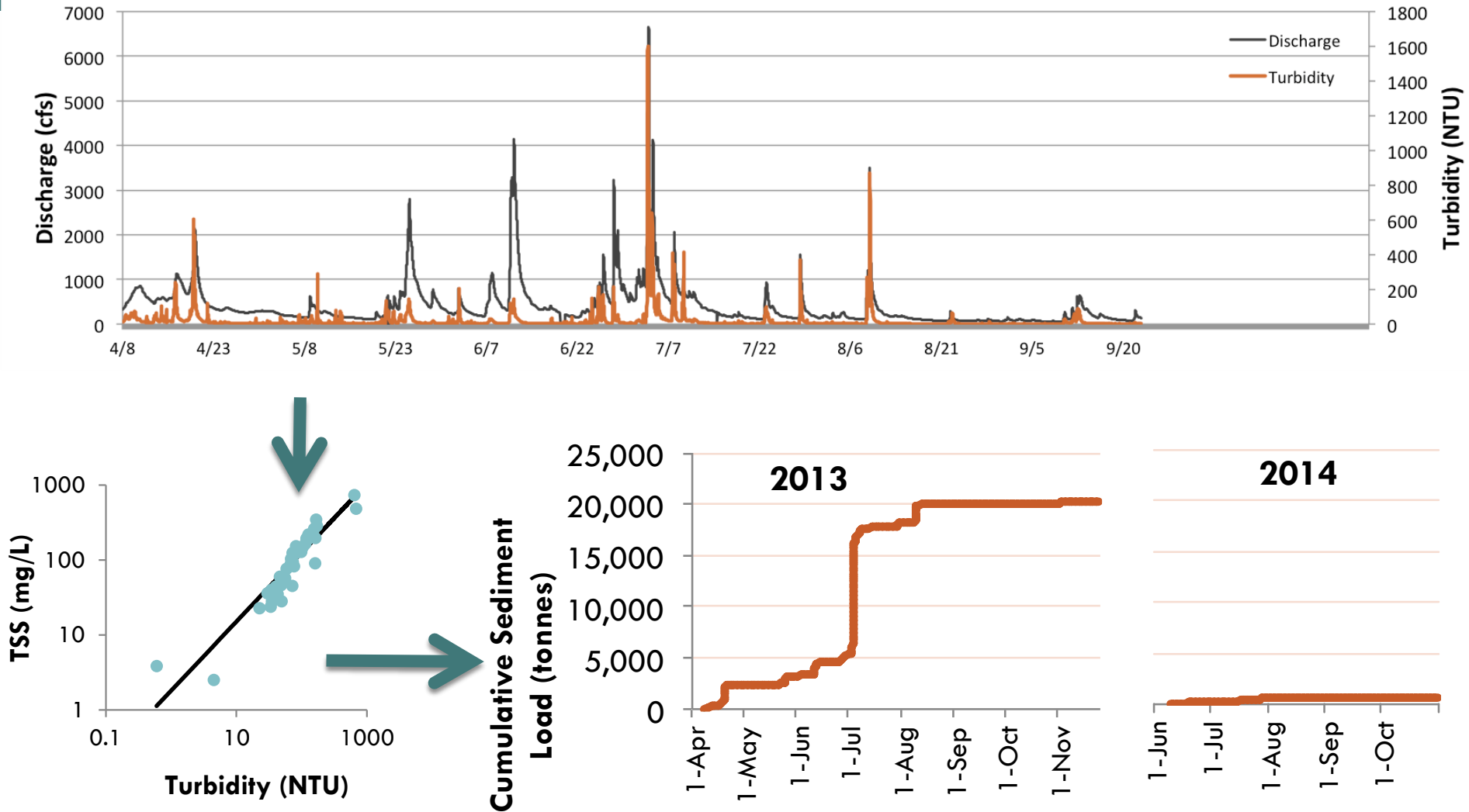
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- Clockwise types (Class II) most common in tributaries
- Mad River more varied in hysteresis types observed



Sediment Load Estimation

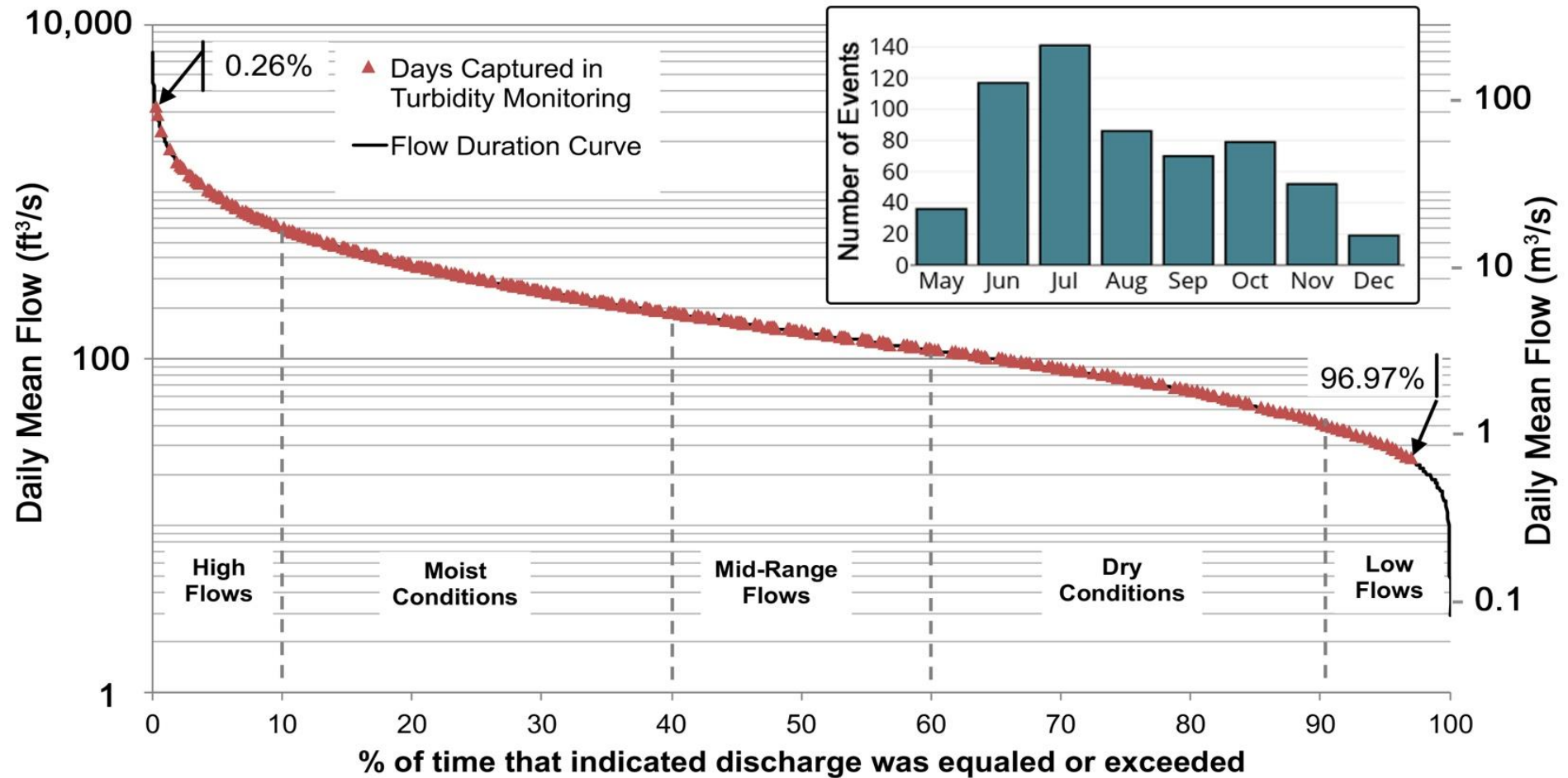
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Hydrology of monitoring period

30

600+ events identified



Hydrological event analysis

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Storm
Metrics

SHP 36

Storm Number: 2.14062422

Storm Date: 24-Jun-2014 22:45:00

Rainfall Station: **Fayston Weather (JFR)**

Total Rainfall (mm): **41.40**

Time Since Last Event (hr): **157.75**

Rainfall-Runoff Ratio: **7.35 %**

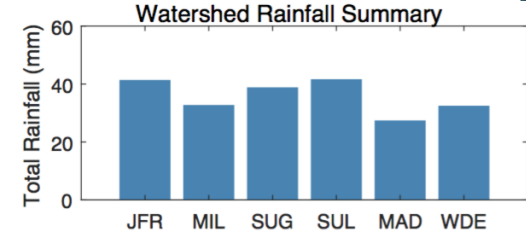
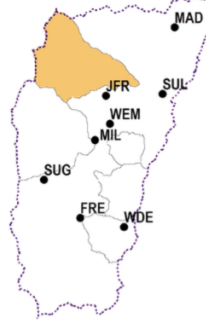
Turbidity Based Load (kg): **22815**

Flood Intensity: **0.505**

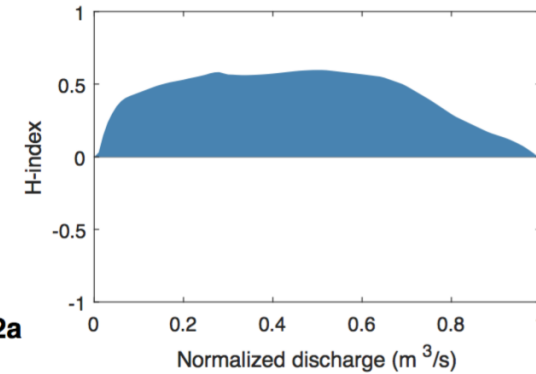
Antecedent Soil Moisture (%): **29.71 %**

Antecedent 7-Day Rainfall (mm): **37.85**

Shepard Brook

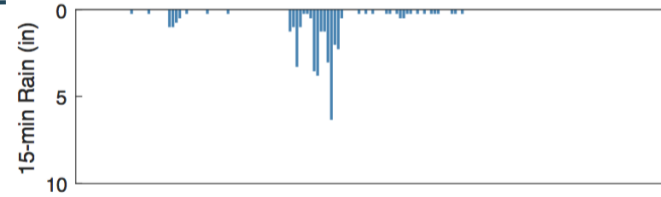


Rainfall
Summary

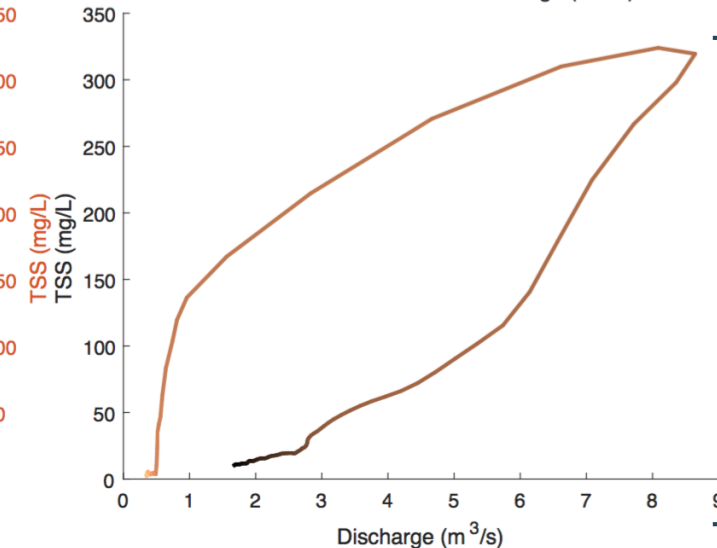
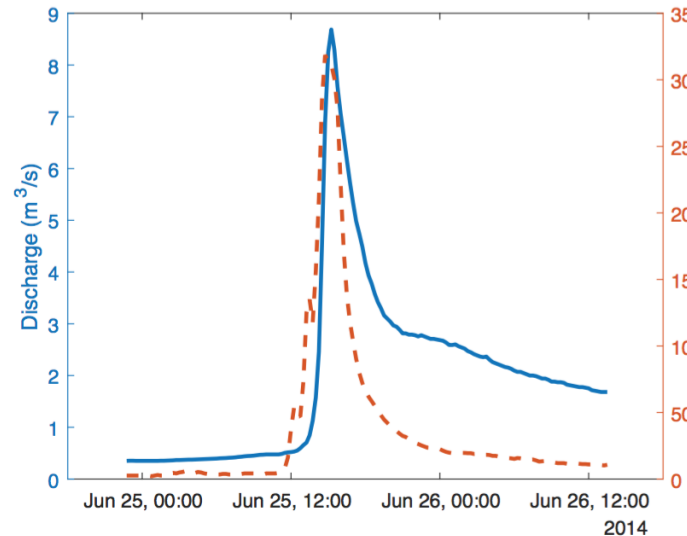


Hysteresis
Index

Hyetograph



Hydrograph &
Sedigraph



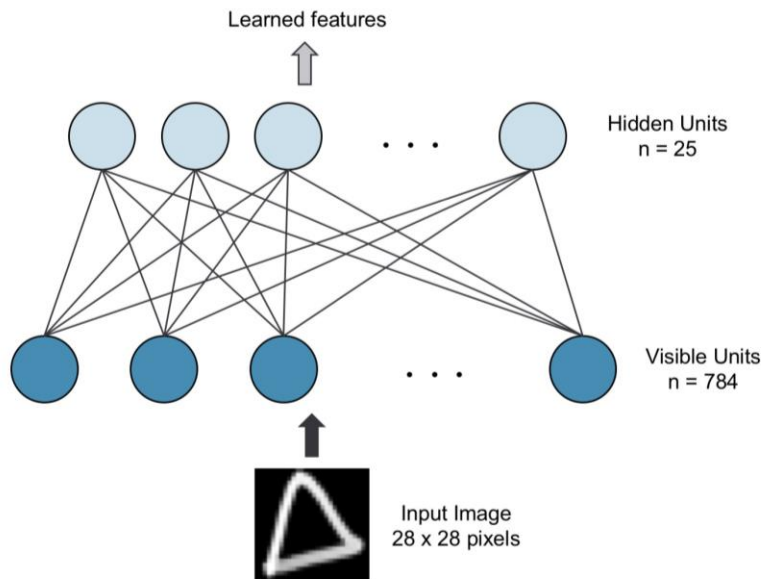
SSC-
Discharge
Plot

Automated Classification using a RBM

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- RBM application
 - ▣ Training: 210 events
 - ▣ Testing: 306 events

Restricted Boltzmann Machine (RBM)



Restricted Boltzmann Machine (RBM)
with Classifier Layer

