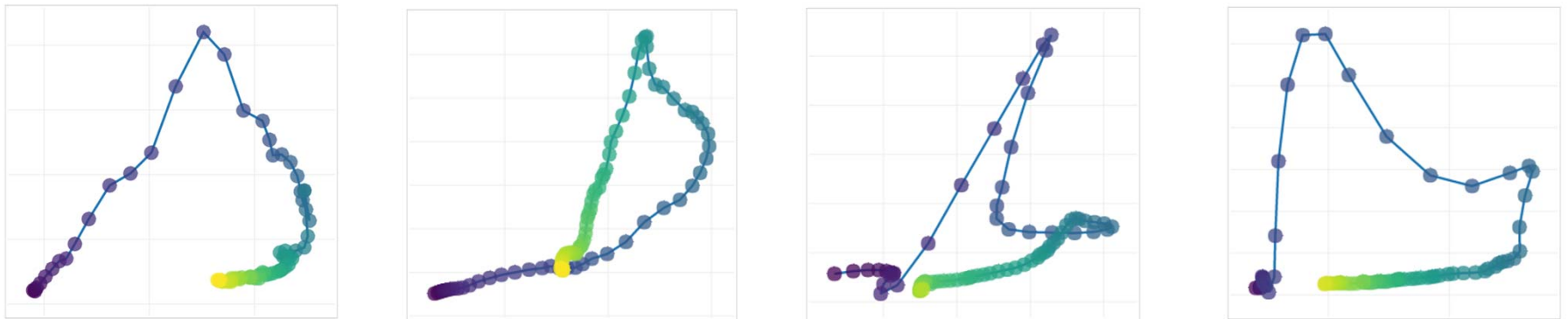


Applying Deep Learning to Hydrological Events for Watershed Modeling

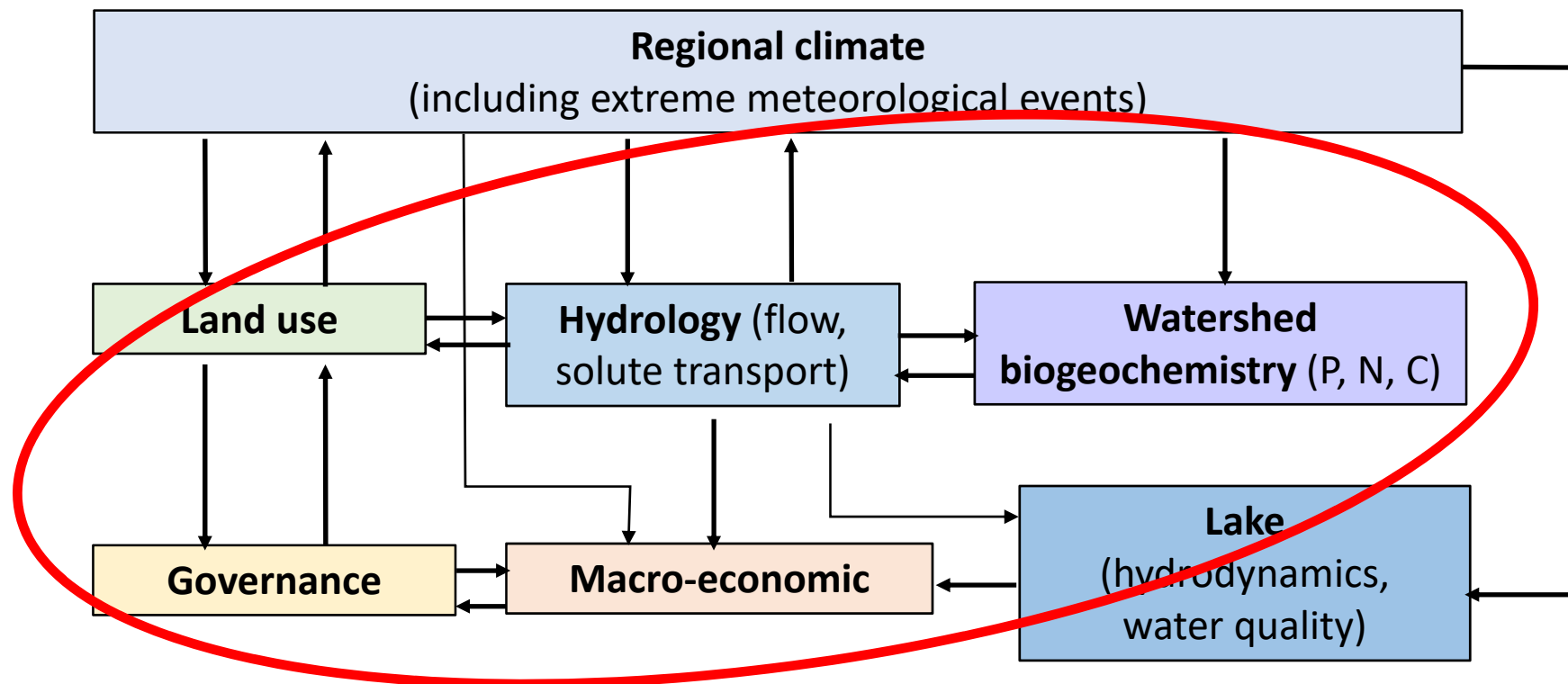


Scott Hamshaw, Donna Rizzo, Mandar Dewoolkar

Vermont EPSCoR | 06.04.2019

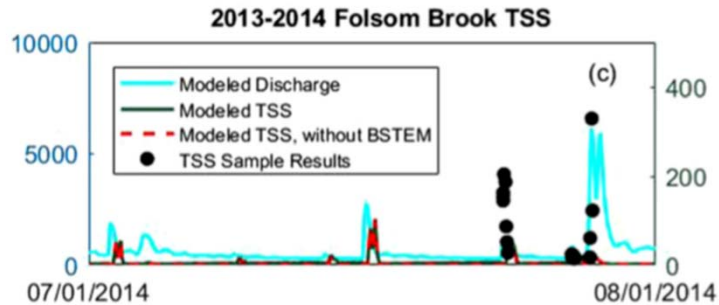


Major Features of the BREE Integrated Assessment Model

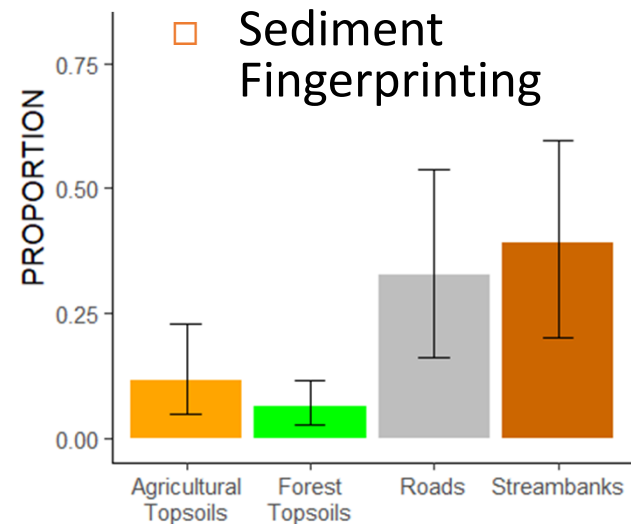


Identifying where riverine sediments originate

Watershed Modeling

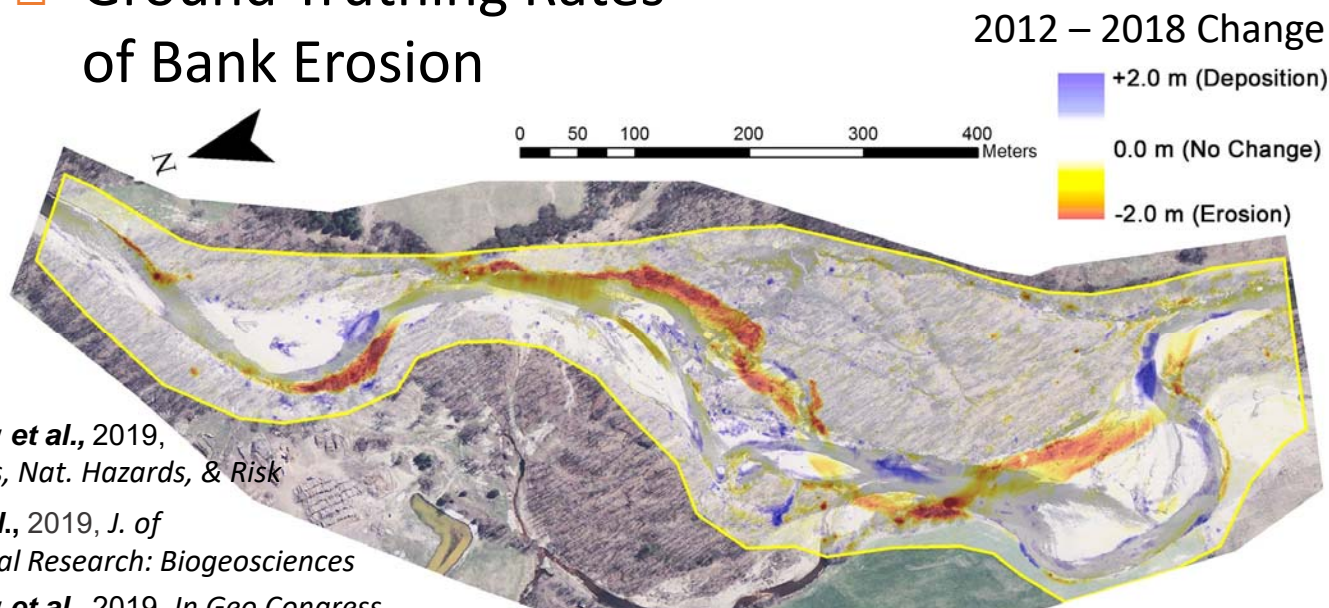


SOURCE: Stryker, et al., 2018, *Journal of Hydrology: Regional Studies*.

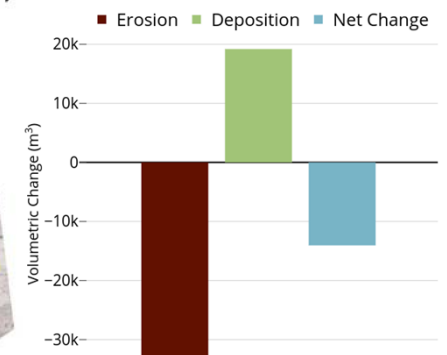


SOURCE: Underwood et al., 2018.

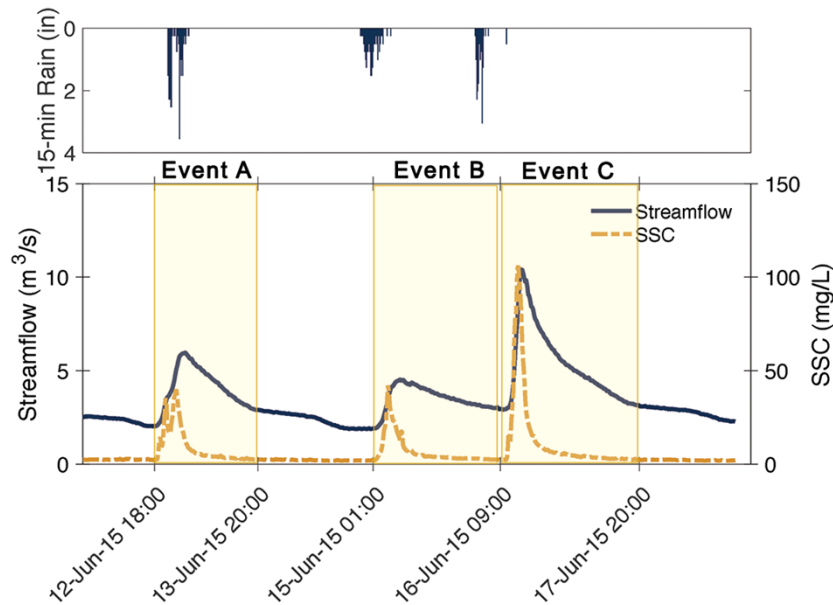
Ground Truthing Rates of Bank Erosion



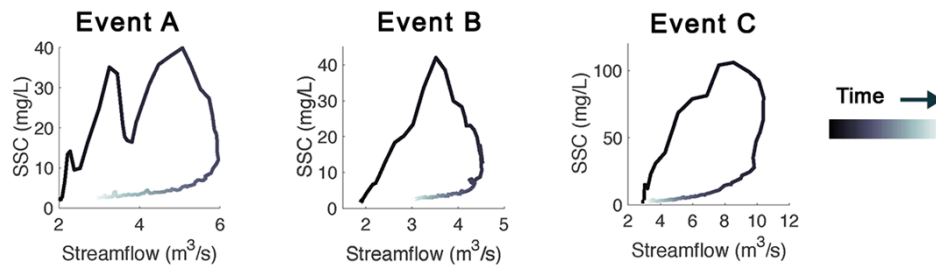
SOURCES:
Hamshaw et al., 2019, *Geomatics, Nat. Hazards, & Risk*
Ross et al., 2019, *J. of Geophysical Research: Biogeosciences*
Hamshaw et al., 2019, *In Geo Congress*.



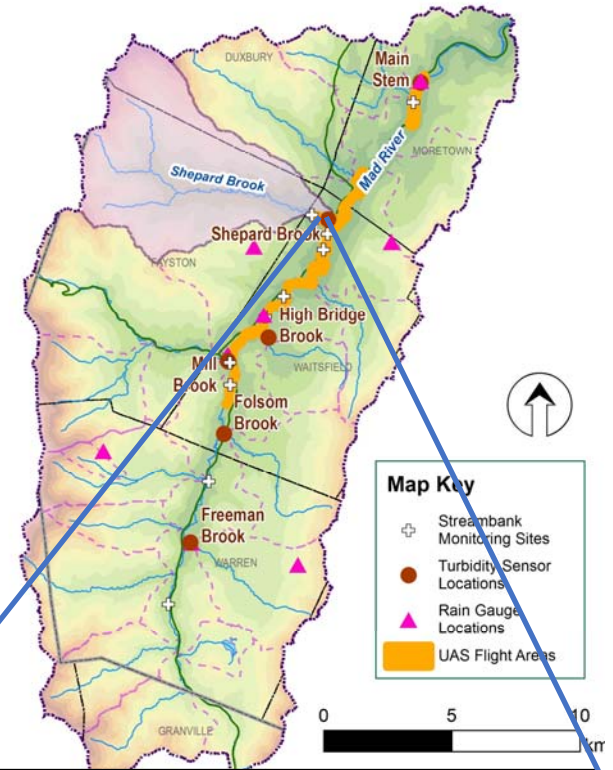
Another approach to understanding sediment dynamics in watersheds



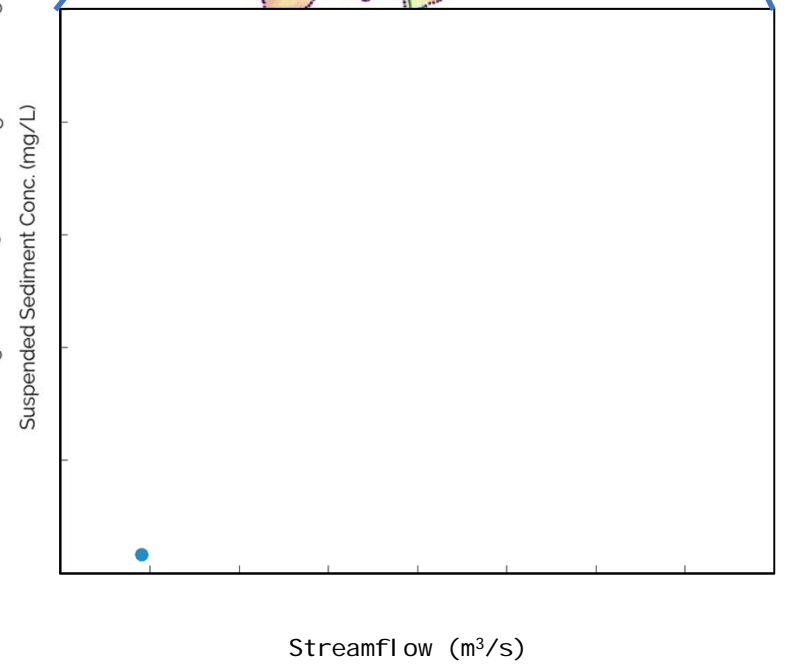
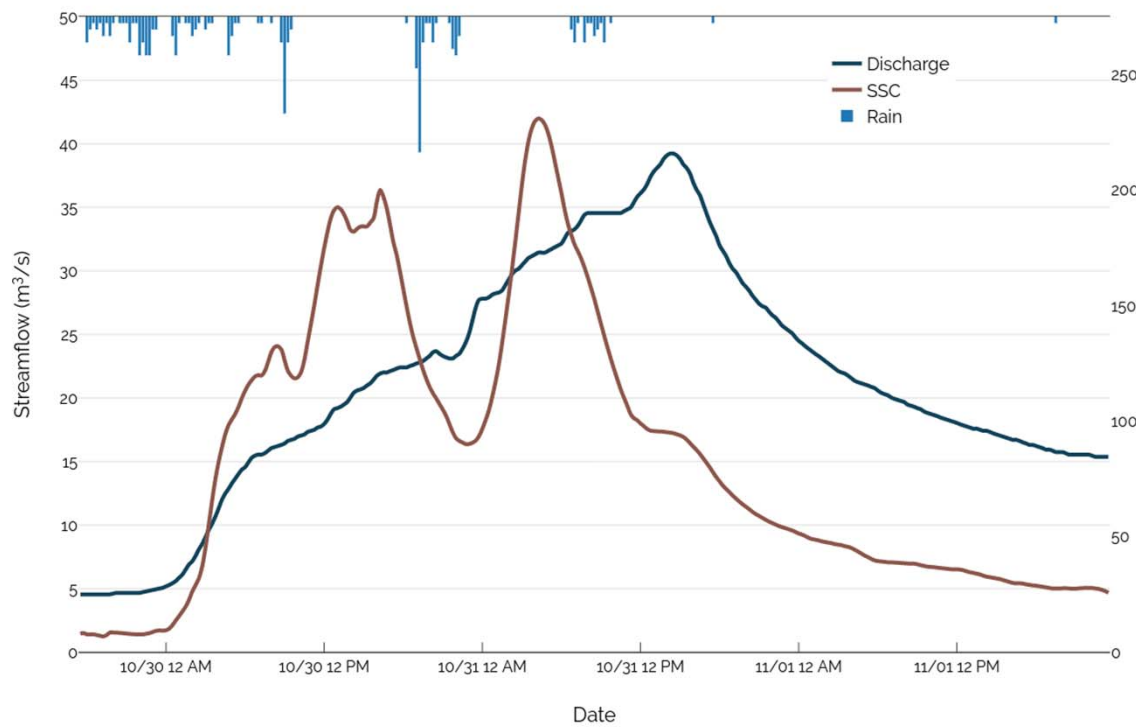
- Mine high-frequency water quality data
- Identify “types” of hydrological events



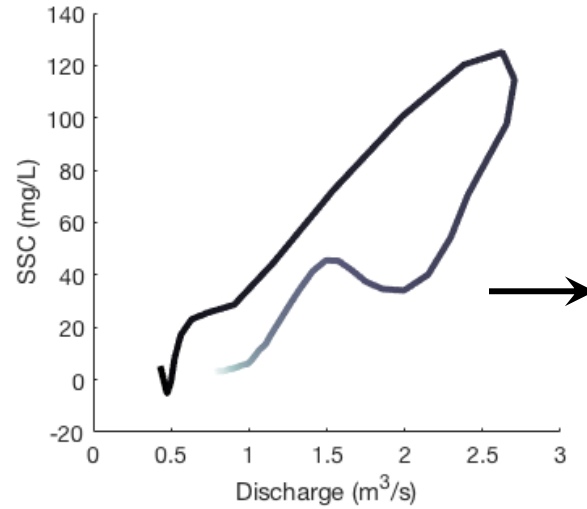
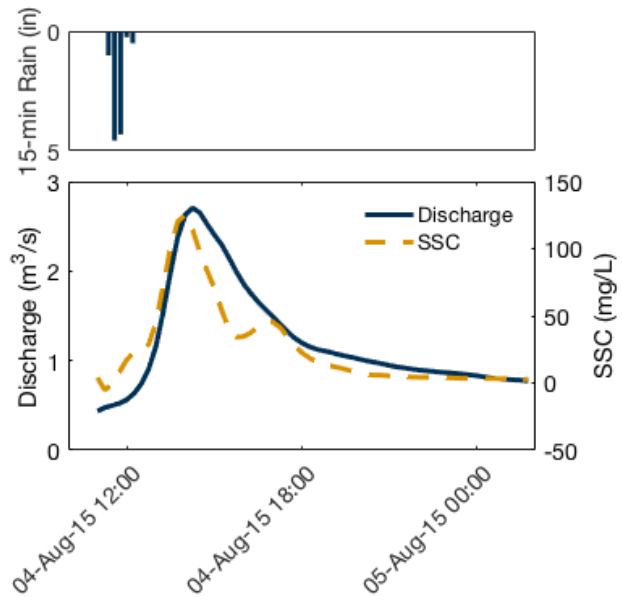
Smart Hydrological event analysis



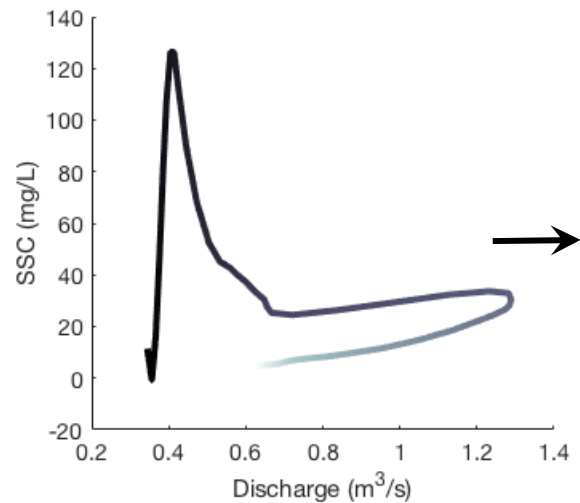
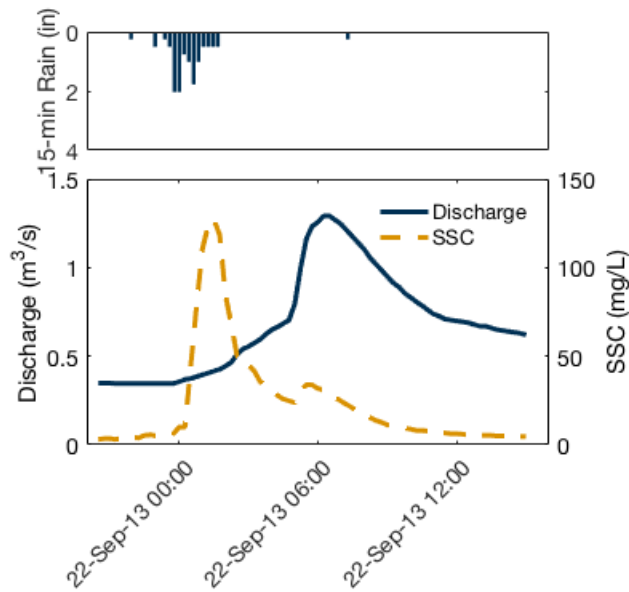
Mad River at USGS Gauge



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



Streamflow activated
(channel network)
sediment sources



Rainfall activated
sediment sources

Methods of hysteresis (event) analysis



□ Garnett Williams, USGS, 1989

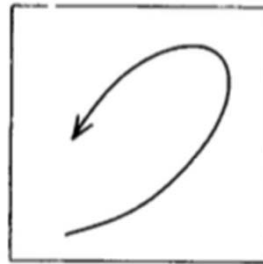
Class I –
Linear



Class II –
Clockwise



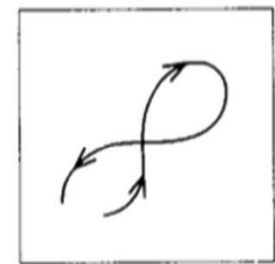
Class III -
Counterclockwise



Class IV – Linear
then Clockwise



Class V –
Figure-Eight



Methods of hysteresis (event) analysis



□ Garnett Williams, USGS, 1989

Class I –
Linear



Class II –
Clockwise



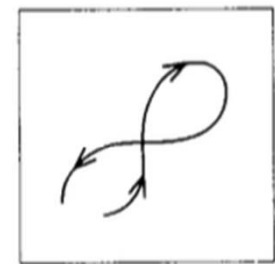
Class III -
Counterclockwise



Class IV – Linear
then Clockwise

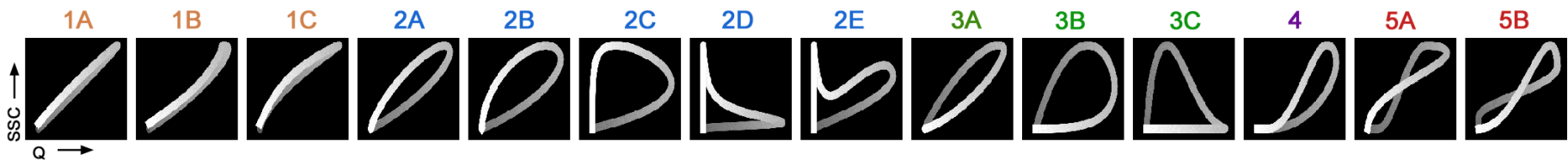


Class V –
Figure-Eight



□ Expanded Hamshaw Classes/Patterns

Hysteresis Type Key



Methods of hysteresis (event) analysis

□ Garnett Williams, USGS, 1989

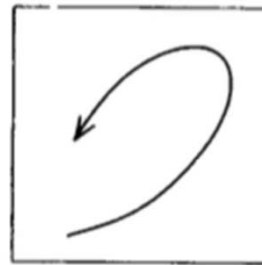
Class I –
Linear



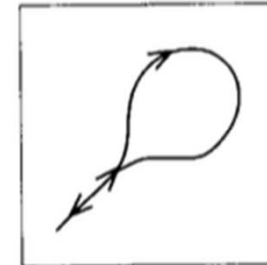
Class II –
Clockwise



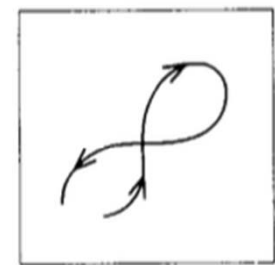
Class III -
Counterclockwise



Class IV – Linear
then Clockwise

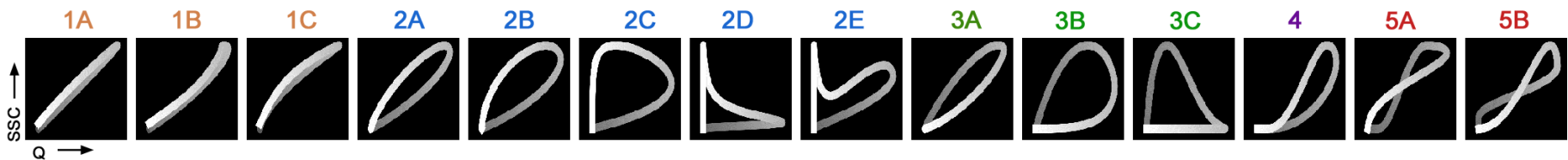


Class V –
Figure-Eight



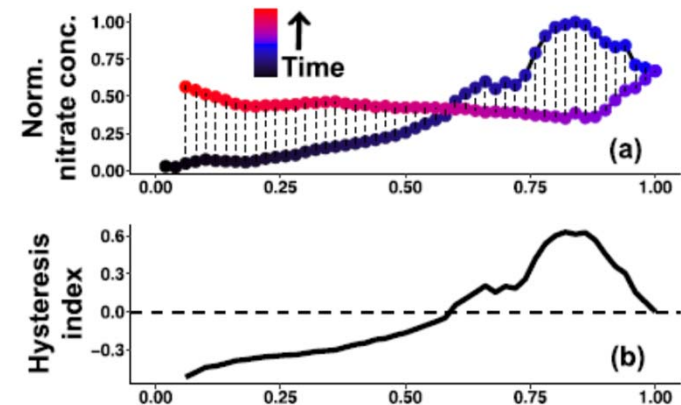
□ Expanded Hamshaw Classes/Patterns

Hysteresis Type Key



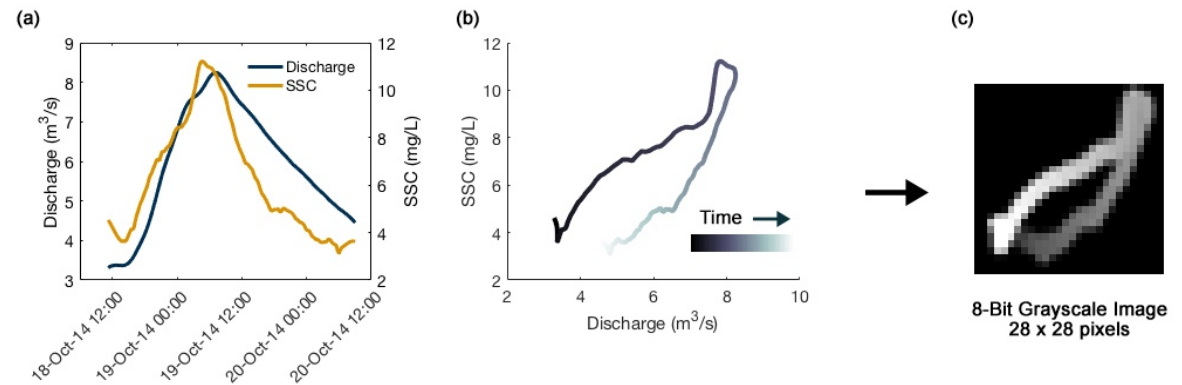
□ Hysteresis Index (non categorical)

HI



SOURCE: Vaughan, et al., 2017 *Water Resources Research*.

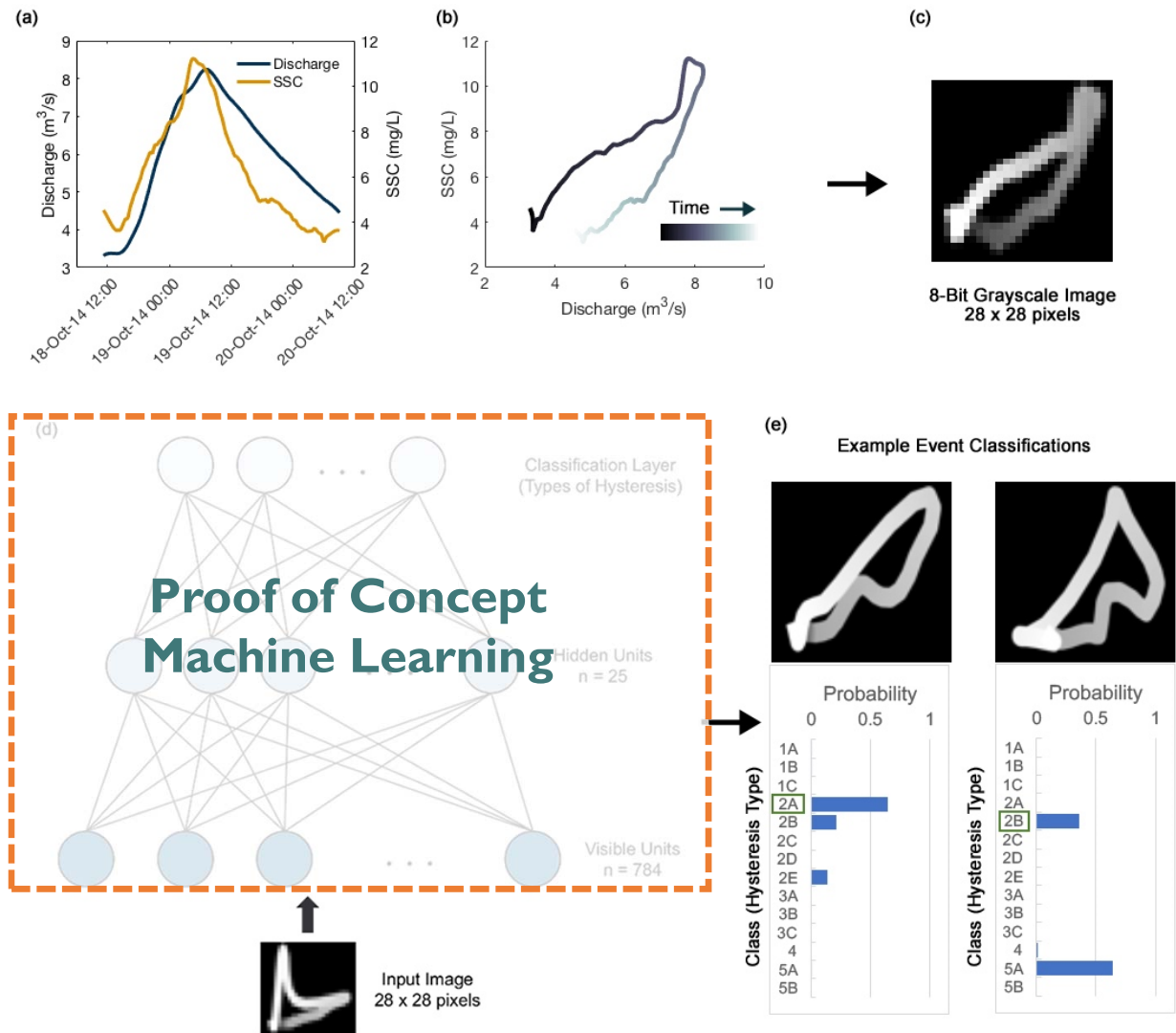
Automated event classification system



Hamshaw et al. (2018). "A new machine learning approach for classifying hysteresis in suspended sediment-discharge relationships using high-frequency monitoring data", *Water Resources Research*

Hamshaw & Rizzo. (2019). "Using Machine Learning to Leverage the Value of Big Data and High-Frequency Monitoring in Characterizing Watershed Sediment Dynamics" in *DOE Open Watershed Science Report (Invited)*

Automated event classification system

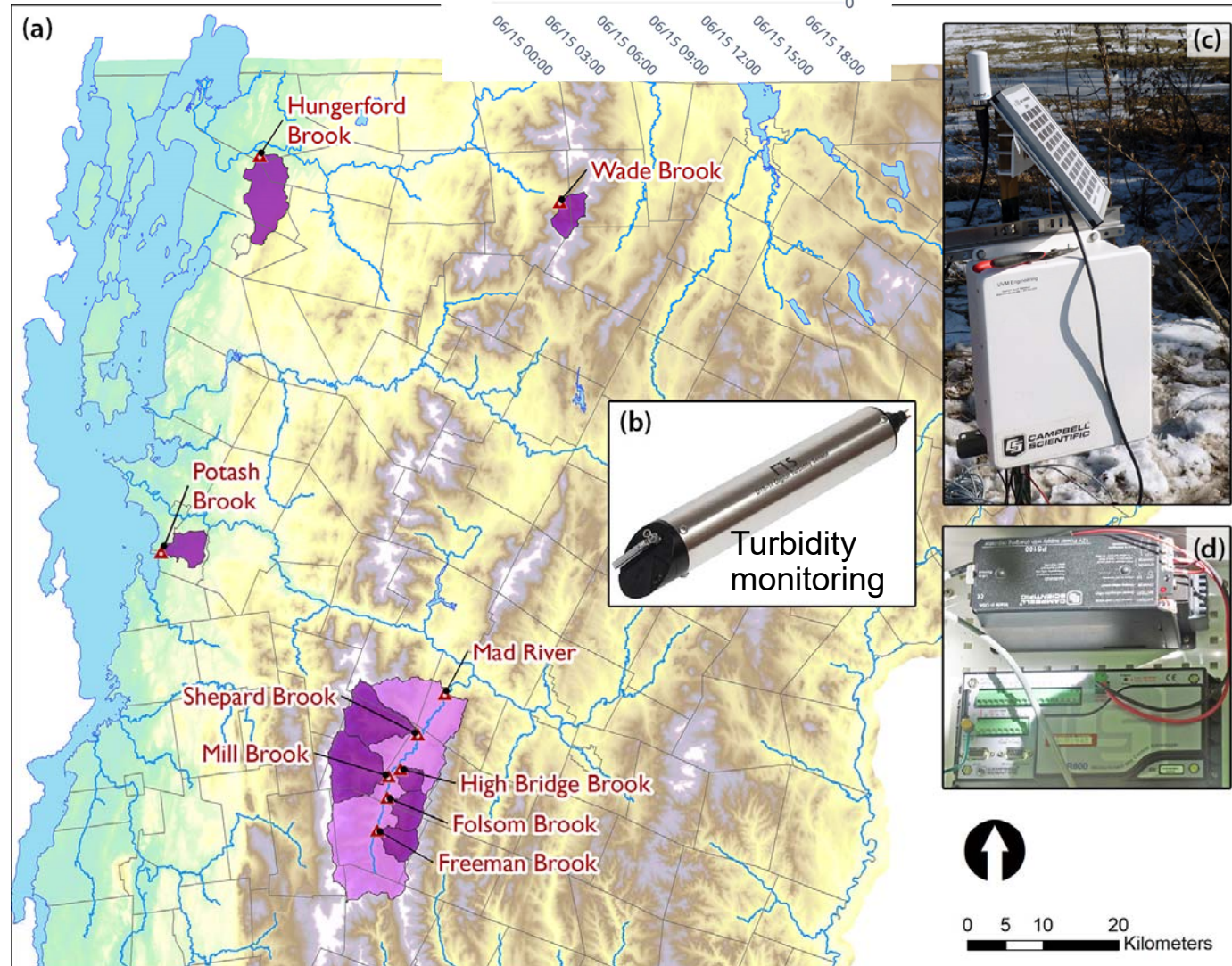
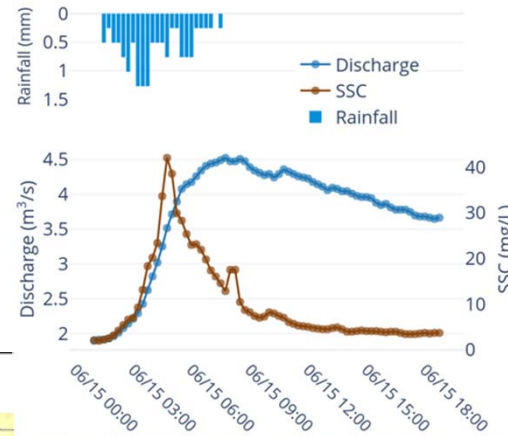


- 53% – 72% Accuracy
- Need for larger storm event data set

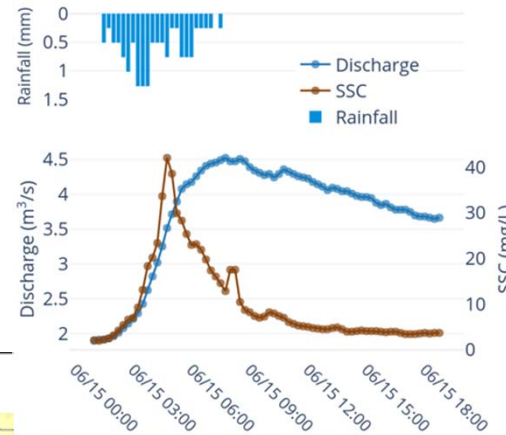
Hamshaw et al. (2018). "A new machine learning approach for classifying hysteresis in suspended sediment-discharge relationships using high-frequency monitoring data", *Water Resources Research*

Hamshaw & Rizzo. (2019). "Using Machine Learning to Leverage the Value of Big Data and High-Frequency Monitoring in Characterizing Watershed Sediment Dynamics" in *DOE Open Watershed Science Report (Invited)*

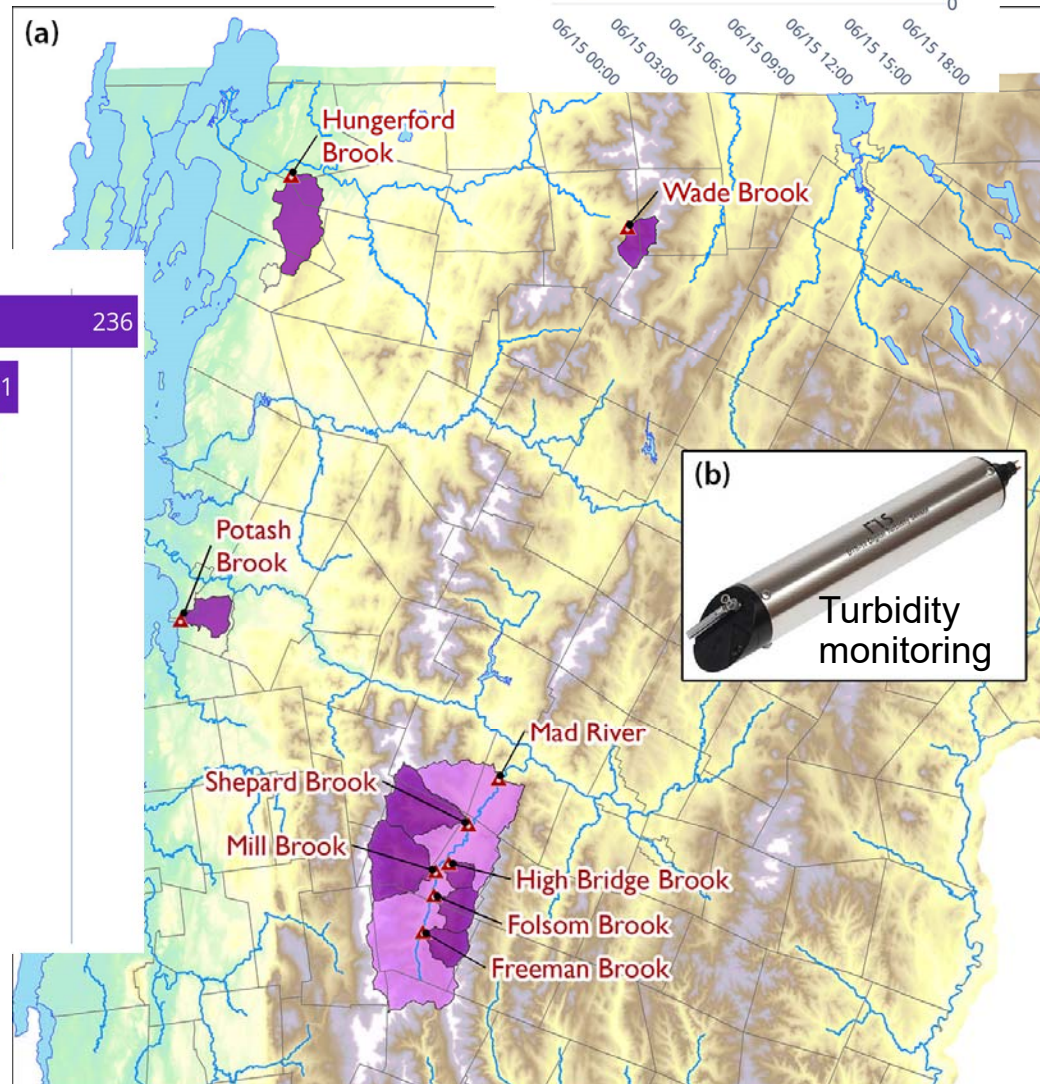
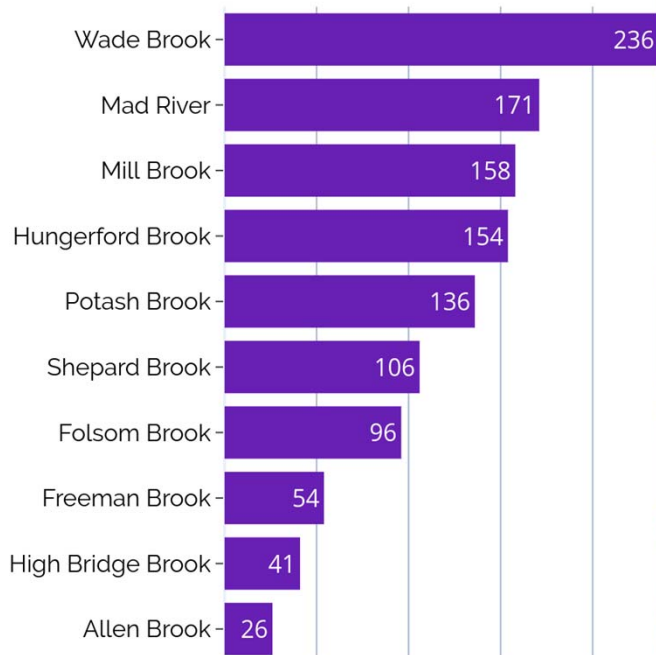
Leveraging EPSCoR in-stream water quality sensors



Leveraging EPSCoR in-stream water quality sensors



- 2012 - 2018
- **1,178 Events**



Automated, supervised event extraction



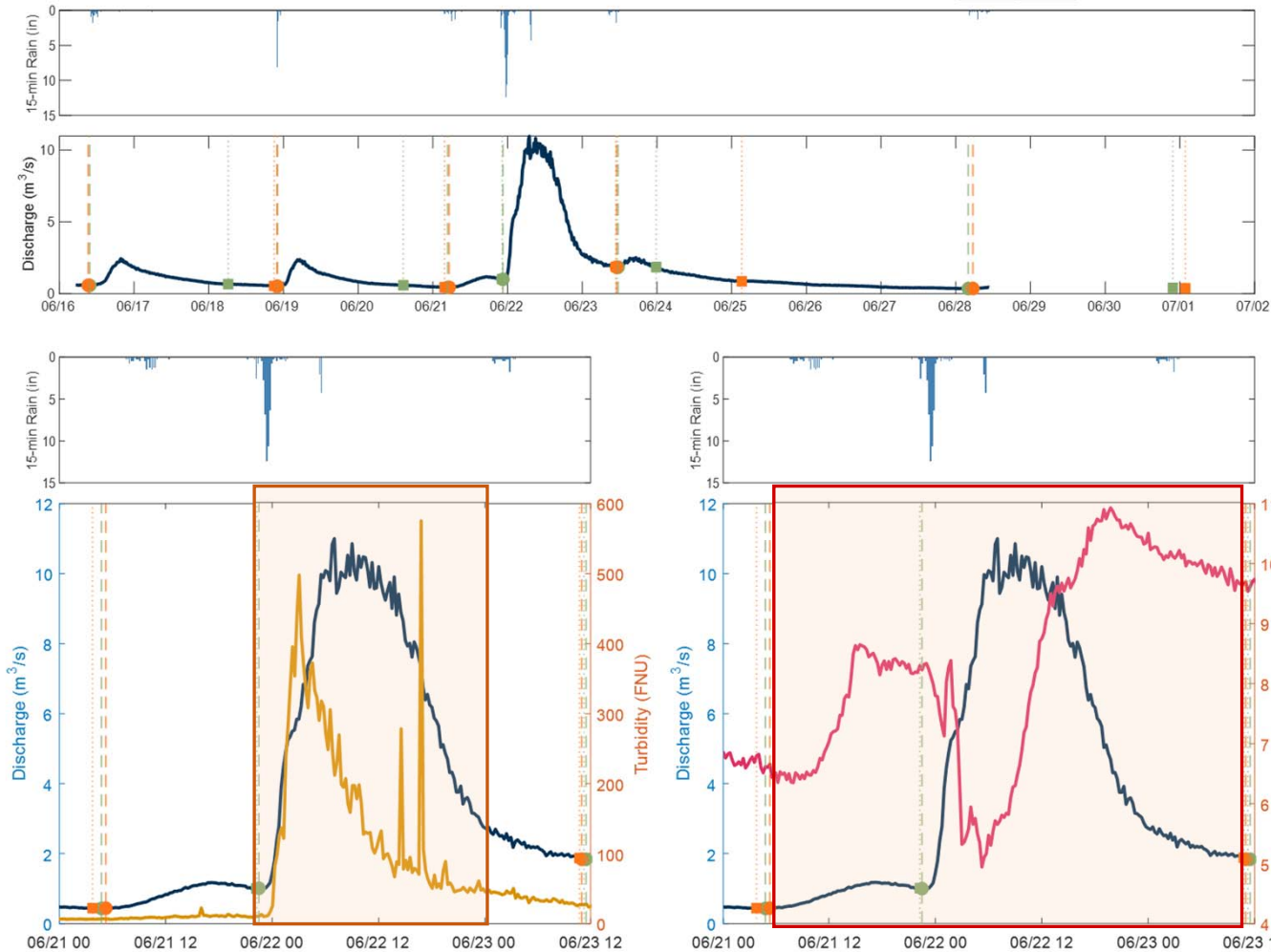
HGF 42

Hungerford Brook

Storm Number: 8.15062105

Storm Date: 21-Jun-2015 05:15:00

Prev Next

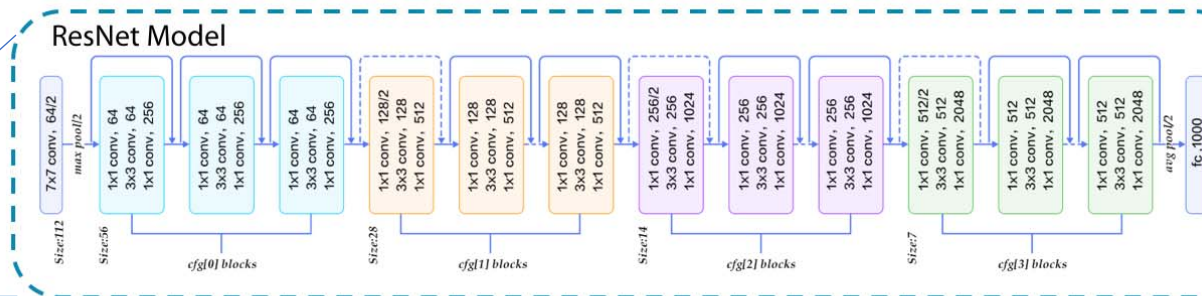
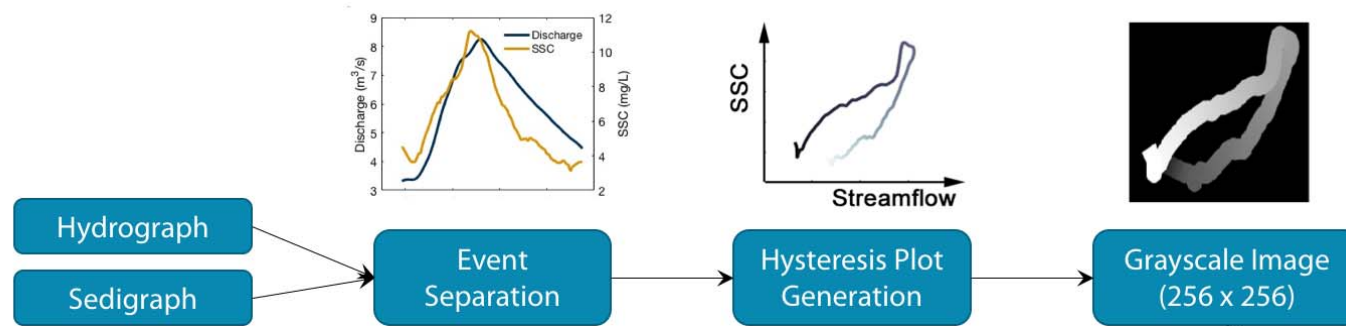


Simultaneous water quality (turbidity, DOC) and streamflow event analysis

Hamshaw & Javed, 2019. Improvements to Event-based Analysis of High-Frequency Turbidity and Suspended Sediment Monitoring Data.

CUAHSI Hydroinformatics Innovation Fellowship

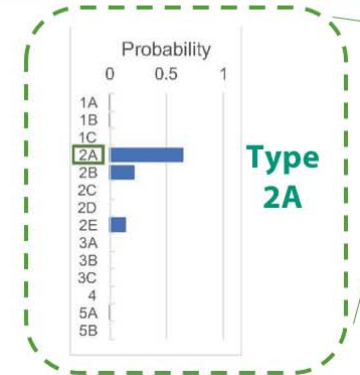
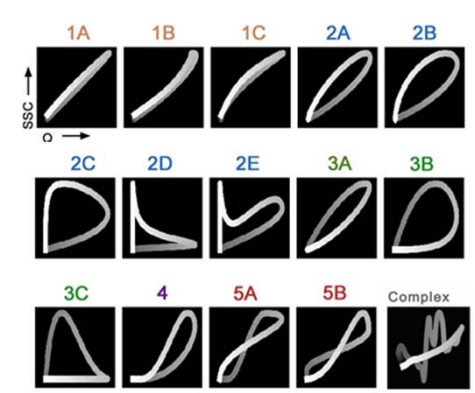
Expansion of event classification to state of the art deep learning algorithms



Deep Learning Classifier (Convolutional Neural Network)

Event Hysteresis Type

**EPSCoR
Nvidia DGX-1
Deep Learning
Server**

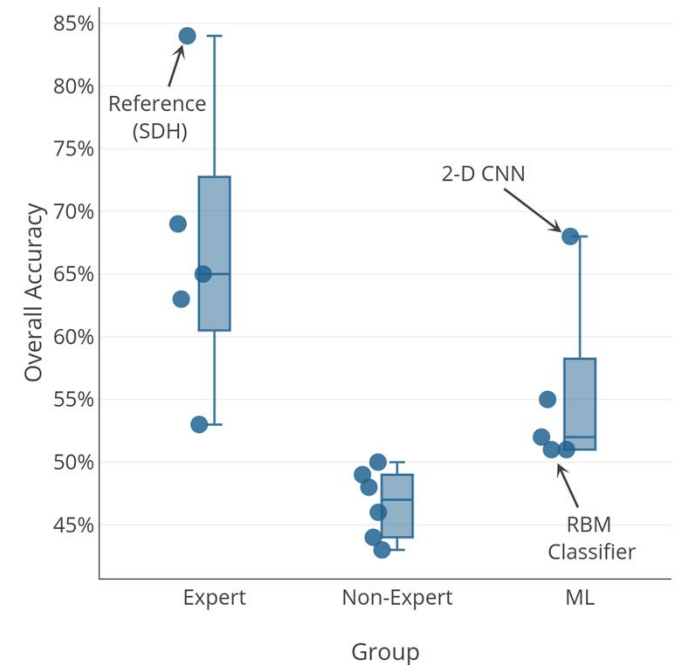
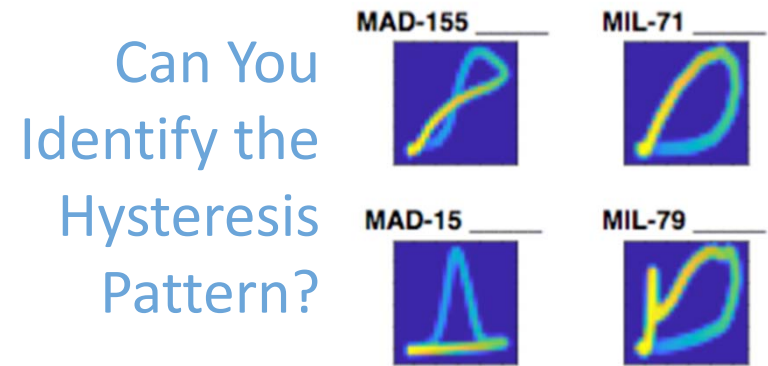
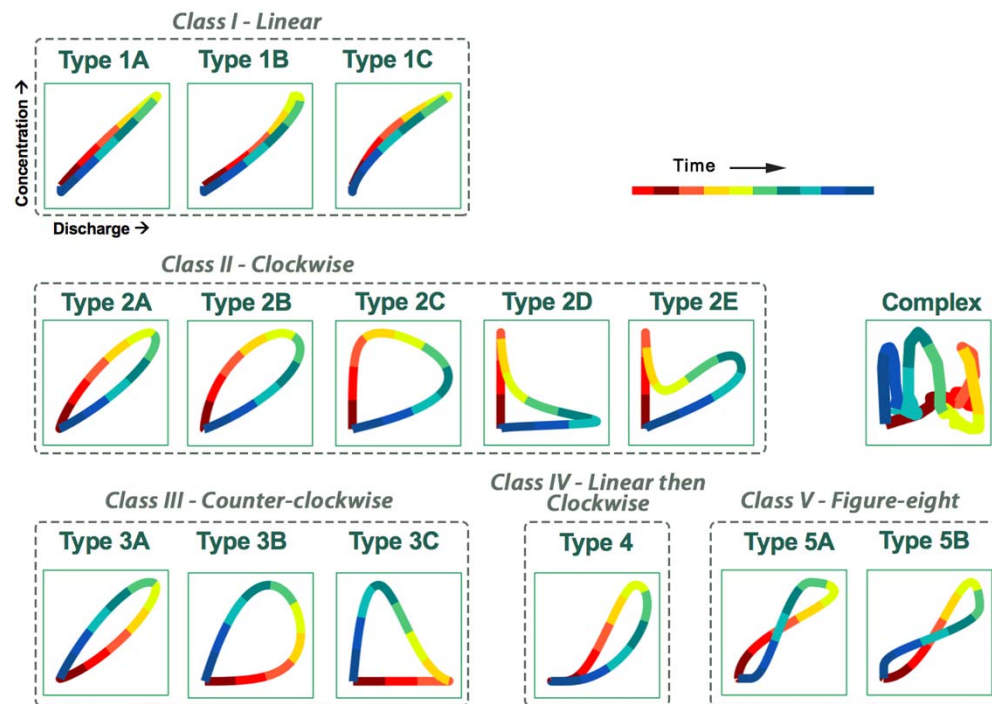


Hamshaw, et al. (2019). "Automating the Classification of Hysteresis in Event Concentration-Discharge Relationships." *In Proceedings of SedHyd 2019 Conference*

Visual representation of hydrological event data

- Color as Time

- Evaluation of human visual interpretation

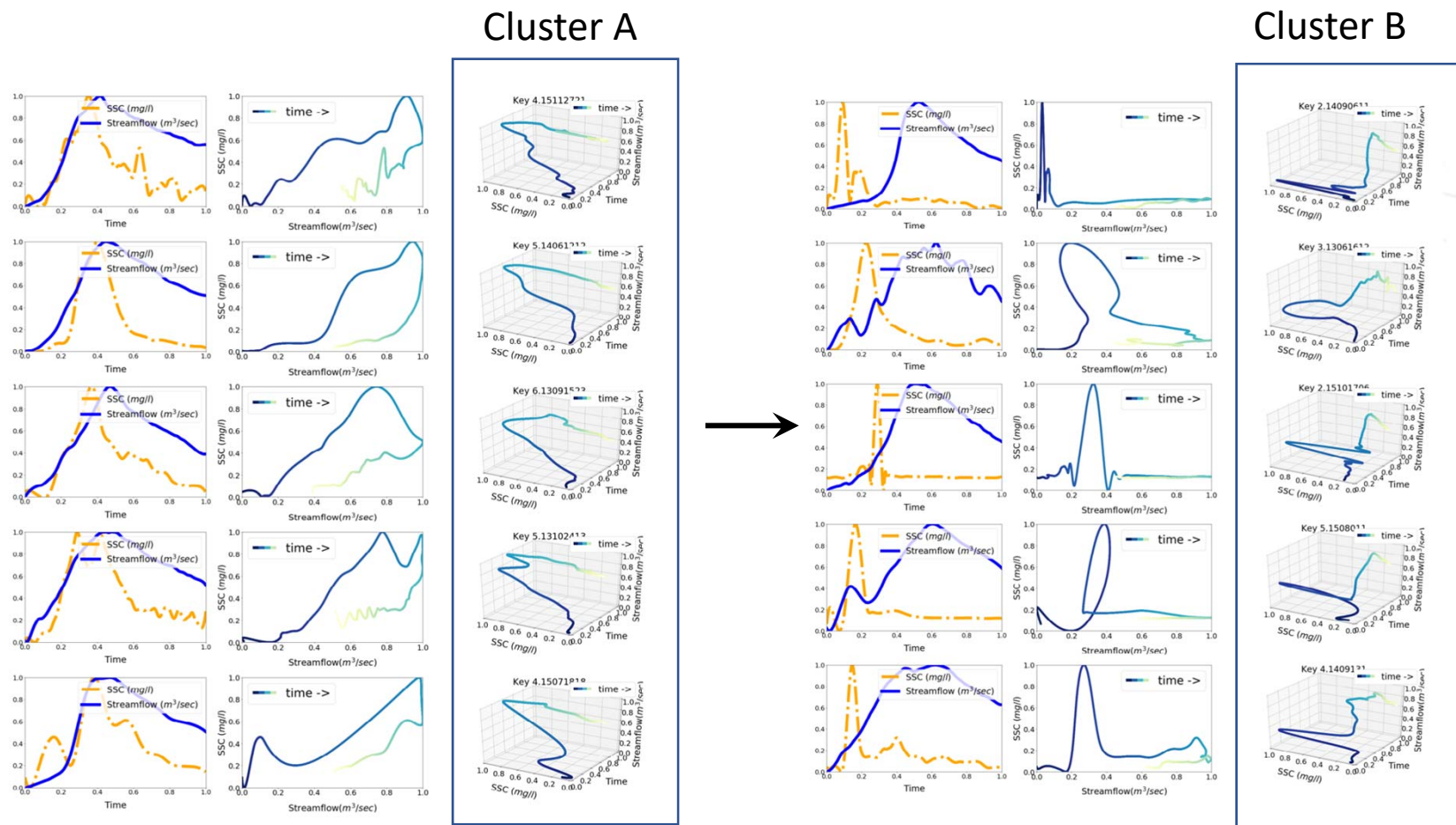


BREE Undergrad Interns Eric Romero & Nicole Dávila

Romero, *et al.* (2018). "Automating the Classification of Hysteresis in Event Concentration-Discharge Relationships." In *AGU Fall Meeting 2018*

Data-driven identification of new categories of events

- Leverage temporal information in sensor signal “3-D trajectories”

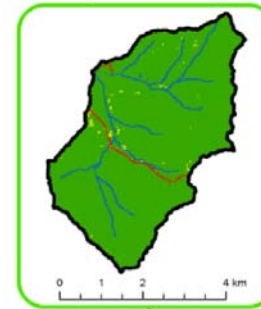


Connection to LULC/BMP

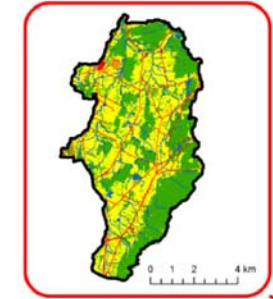
- Watersheds have existing characteristic distribution of event types

- Affected by changes in
 - Climate & extreme events
 - LULC & BMP adoption

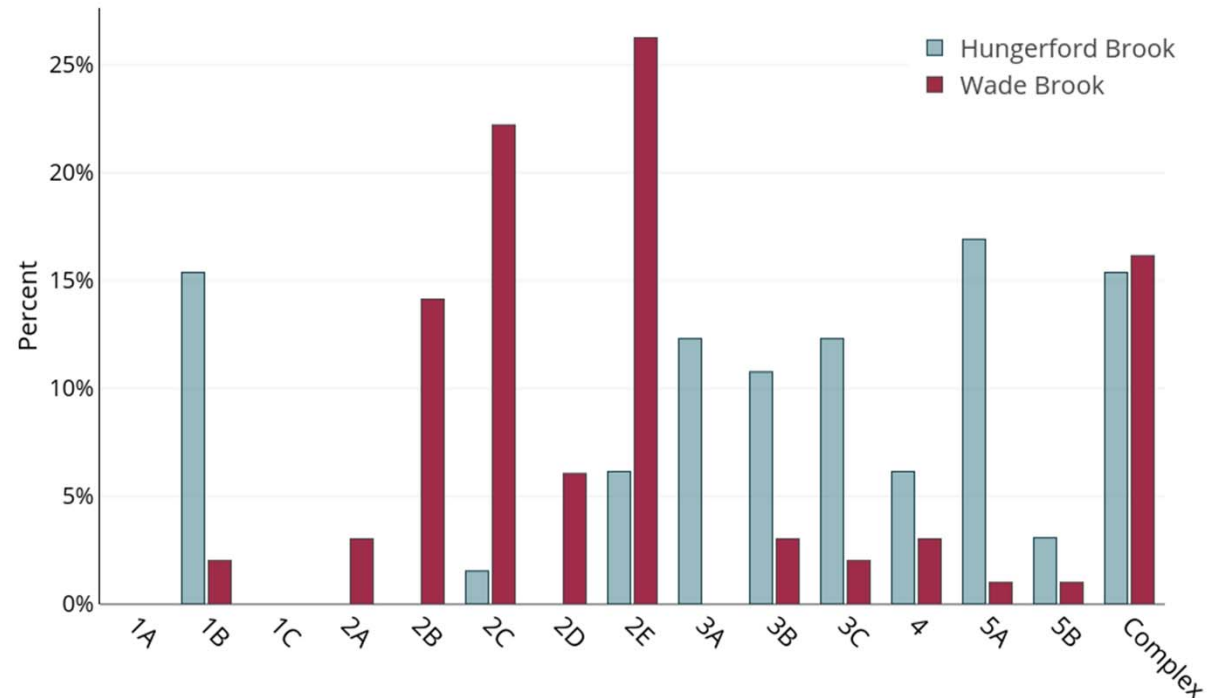
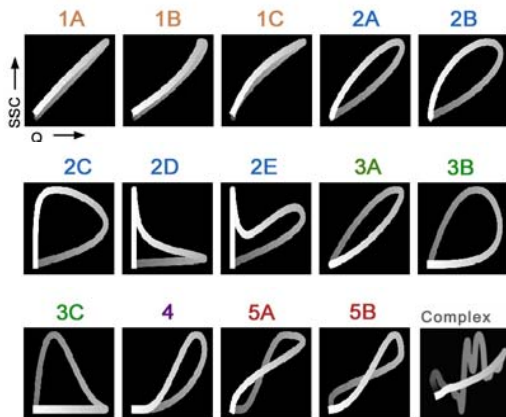
Wade Brook



Hungerford Brook

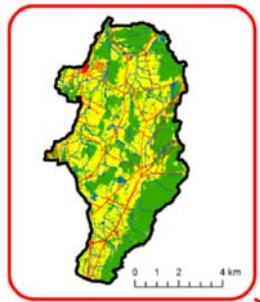
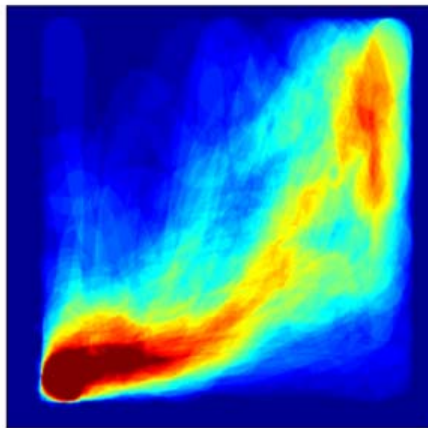


Event Types

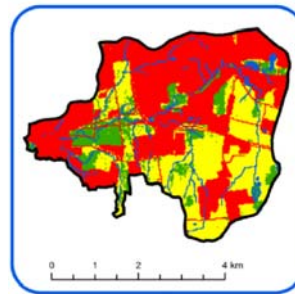
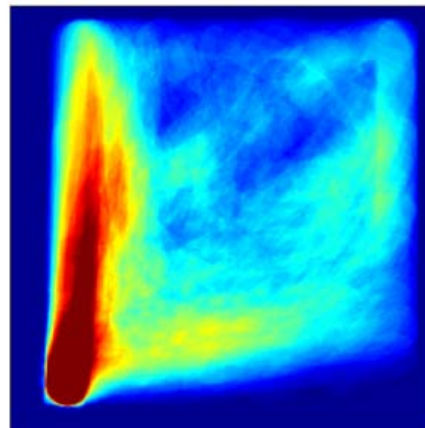


Smart detection of shifts in storm event behavior

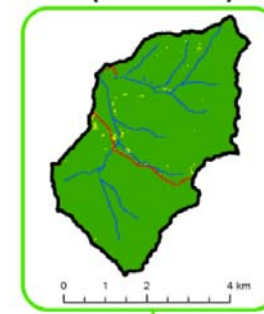
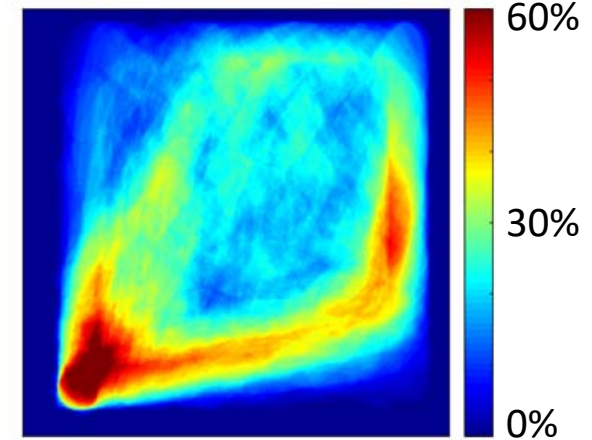
Watershed 3:
Hungerford Brook



Watershed 2:
Potash Brook



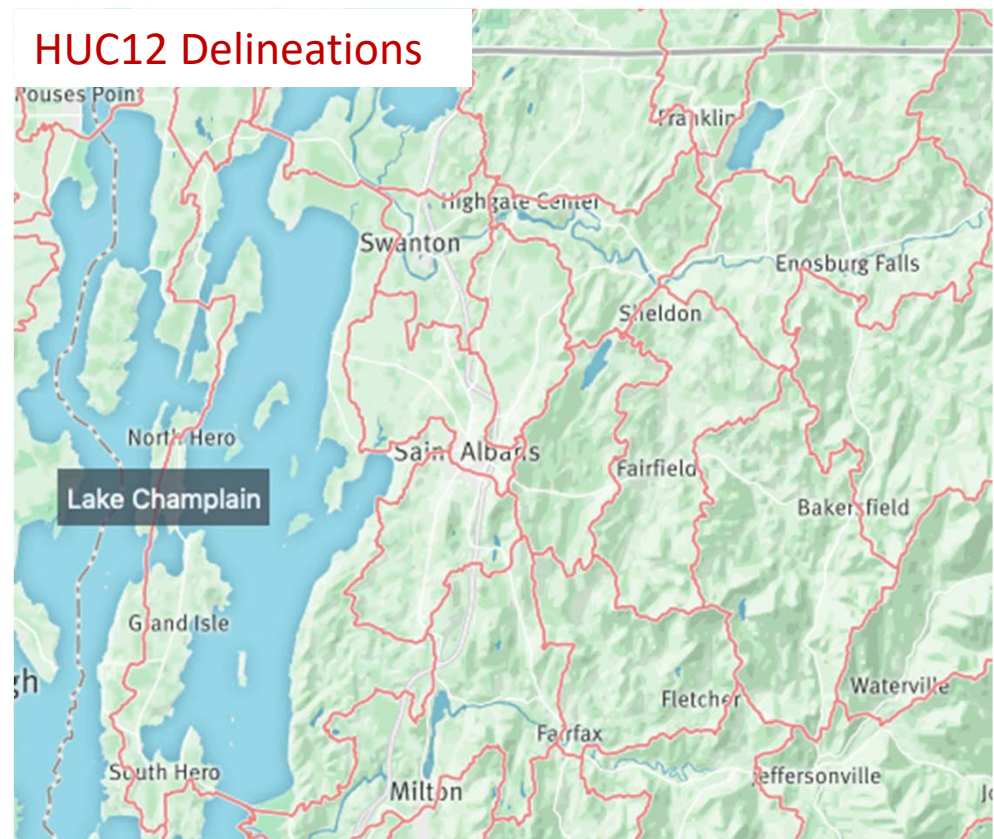
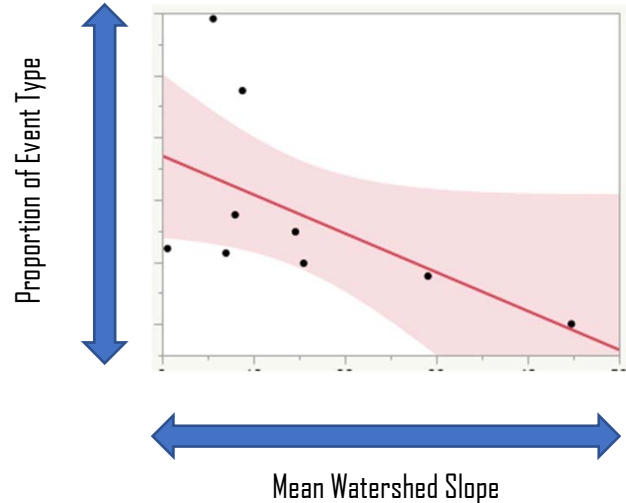
Watershed 1:
Wade Brook



Maps courtesy
Matt Vaughan

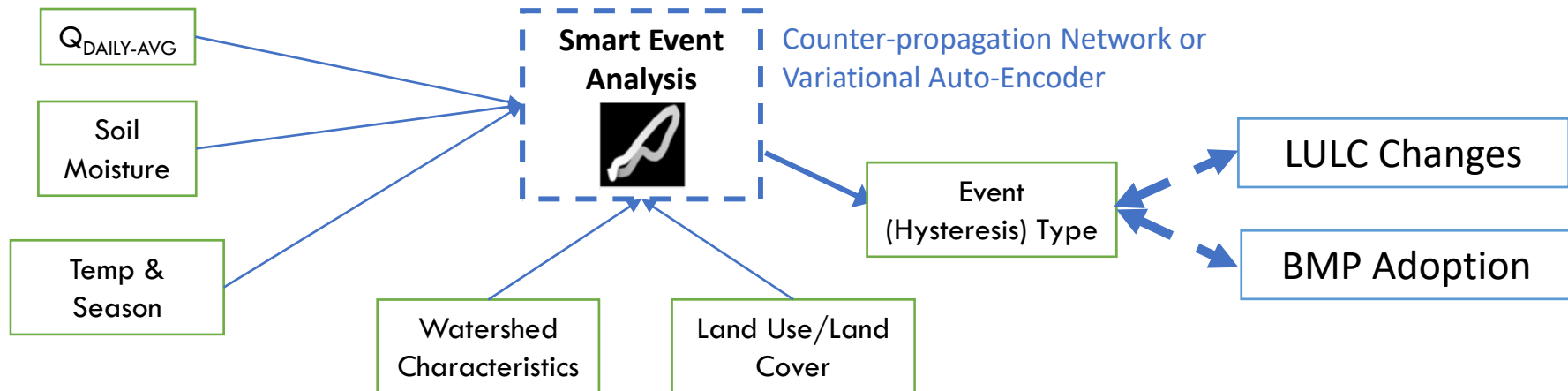
Scaling event analysis to additional watersheds in the IAM

- Regionalization approach



Scaling event analysis to additional watersheds in the IAM

- Predict hysteresis type occurring at subwatershed outlet



Output → Informs BMP Implementation and Adoption:

- RHESys (e.g., alter manure application at subwatershed level)
- ABM (e.g., alter intent to adopt)

