

Accuracy and classification of river form and extent from remote observations

Tamlin Pavelsky
University of North Carolina
Department of Geological Sciences

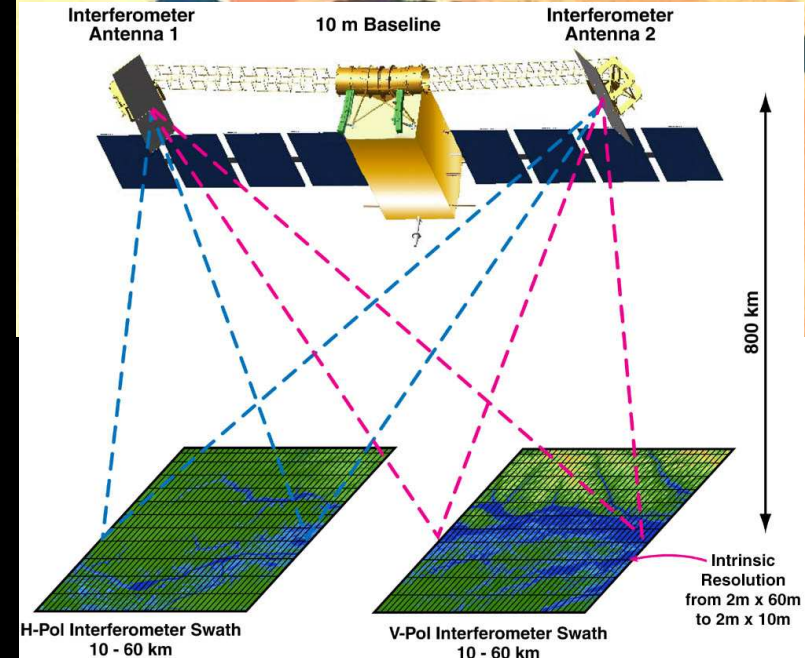
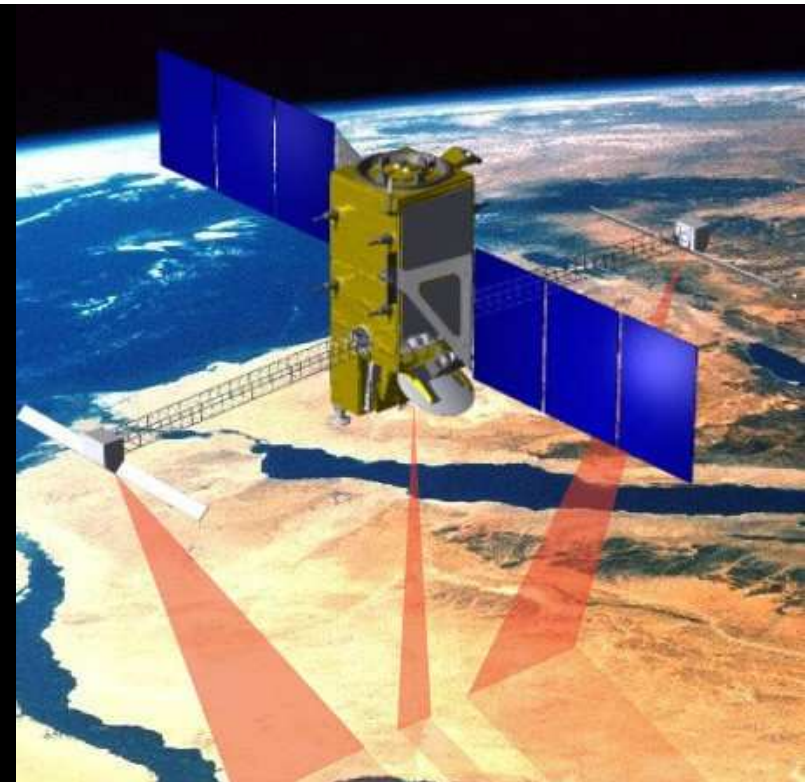
Oceans and Hydrology Applications Workshop
Lisbon, Portugal
October 21st, 2010

SWOT Measurements of Surface Water

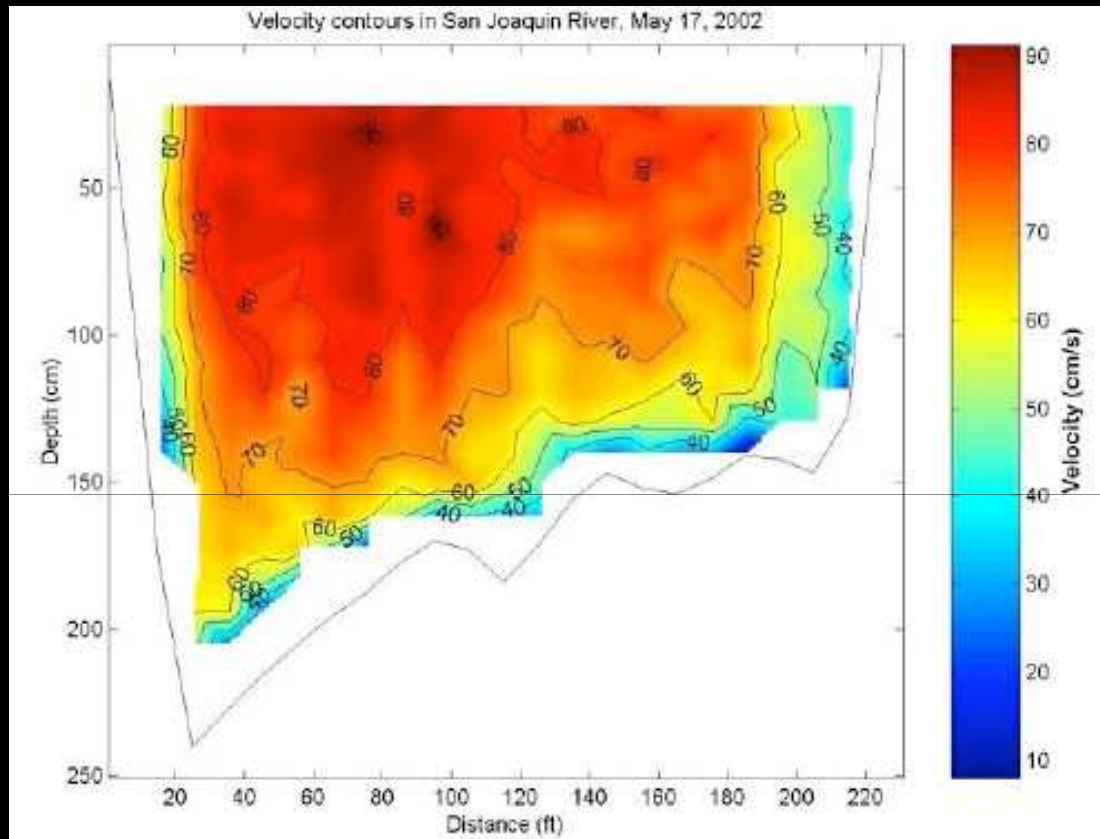
SWOT will directly measure variations in inundated area and water surface elevation.

From these quantities, we will be able to estimate both river discharge and change in water storage.

Before launch, we need to understand and quantify the sources of uncertainty in SWOT observations.



Source of Error in SWOT Discharge Measurements



$$Q = wdv$$

All three dimensions must be measured or estimated. Errors in all three will contribute to discharge error.

Algorithms to estimate depth and velocity are currently under development (Durand, Smith, Andreadis, etc).

Focus of this Presentation:

How will errors in measurement of river width affect discharge retrievals in different kinds of rivers?

Ohio

Wide, Single Channel



Widths: 300-1000 m
Mean Discharge: 8000 m³/s
Principal Adjustment:
Depth

Tanana

Wide, Braided



Widths: 300-1500 m
Mean Discharge: 1200 m³/s
Principal Adjustment:
Width

Kentucky

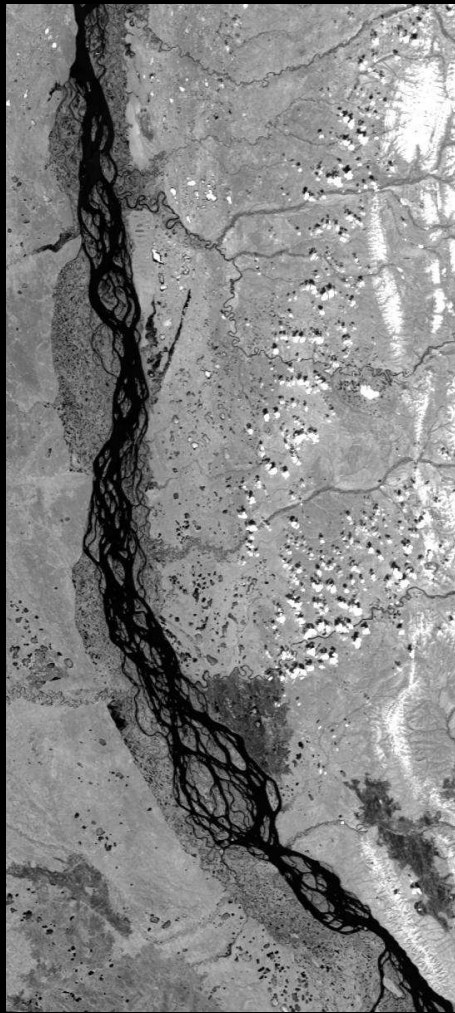
Narrow, Single Channel



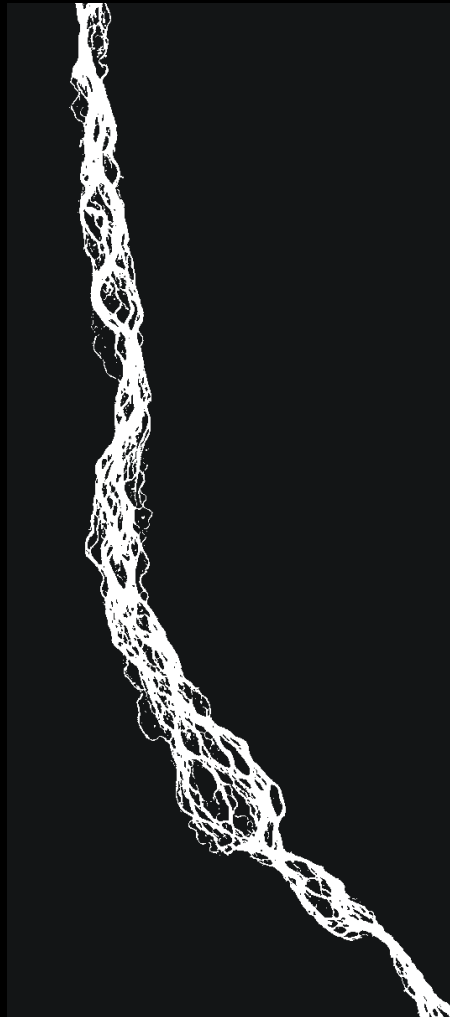
Widths: 50-200 m
Mean Discharge: 285 m³/s
Principal Adjustment:
Depth

RivWidth: Automatic River Flow Width Monitoring

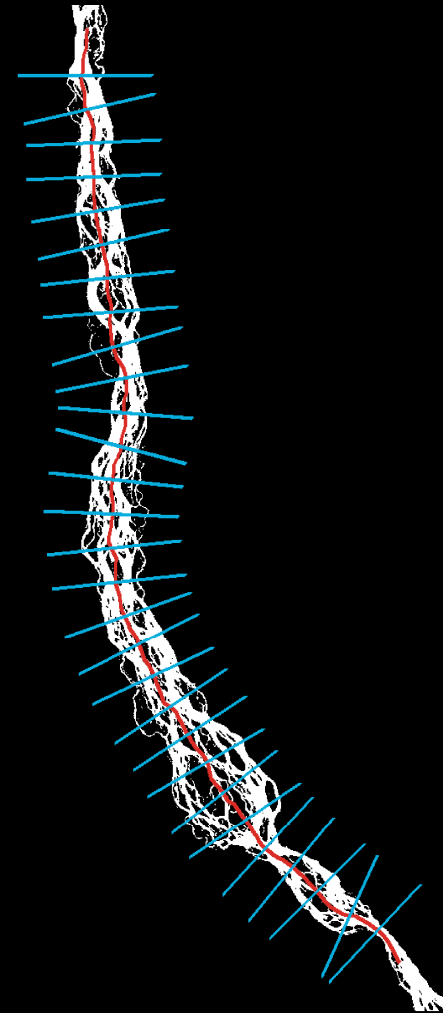
Raw Image



Channel Mask



Cross-Sections



Sources of Error in Width Classification

- Misclassification of pixels as water/nonwater
 - Sources: Emergent vegetation, wet sediment
- Error in cross-section calculation
 - Source: Complex river planforms
- Boundary/edge effect errors
 - Source: Inherent in calculation of river width from binary inundation masks
 - Defined by the number of channels in a river:

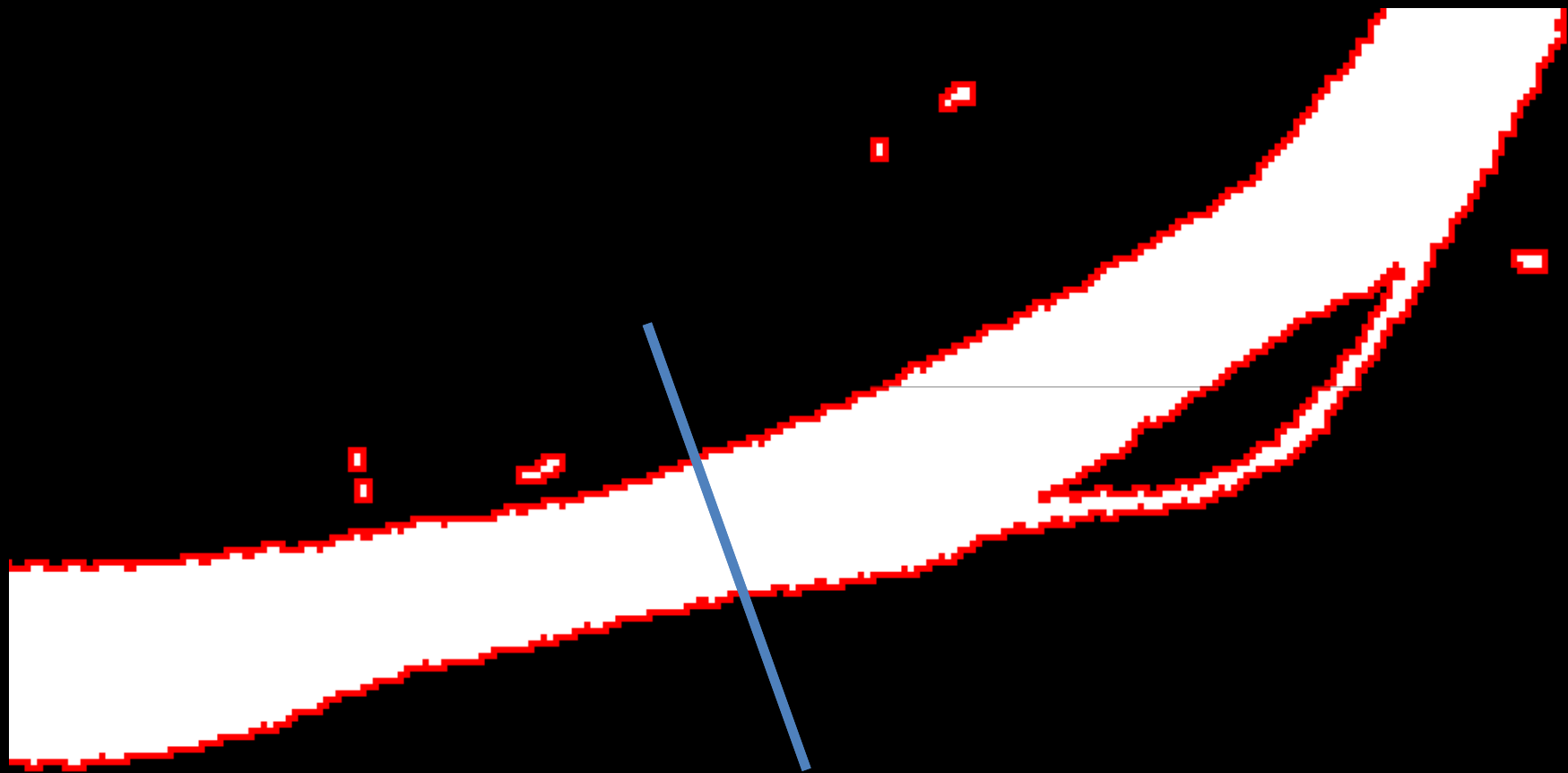
$$E = \frac{1}{2}RC$$

E: edge effect error

R: pixel resolution

C: channel crossings

Error in Cross-Section Calculation

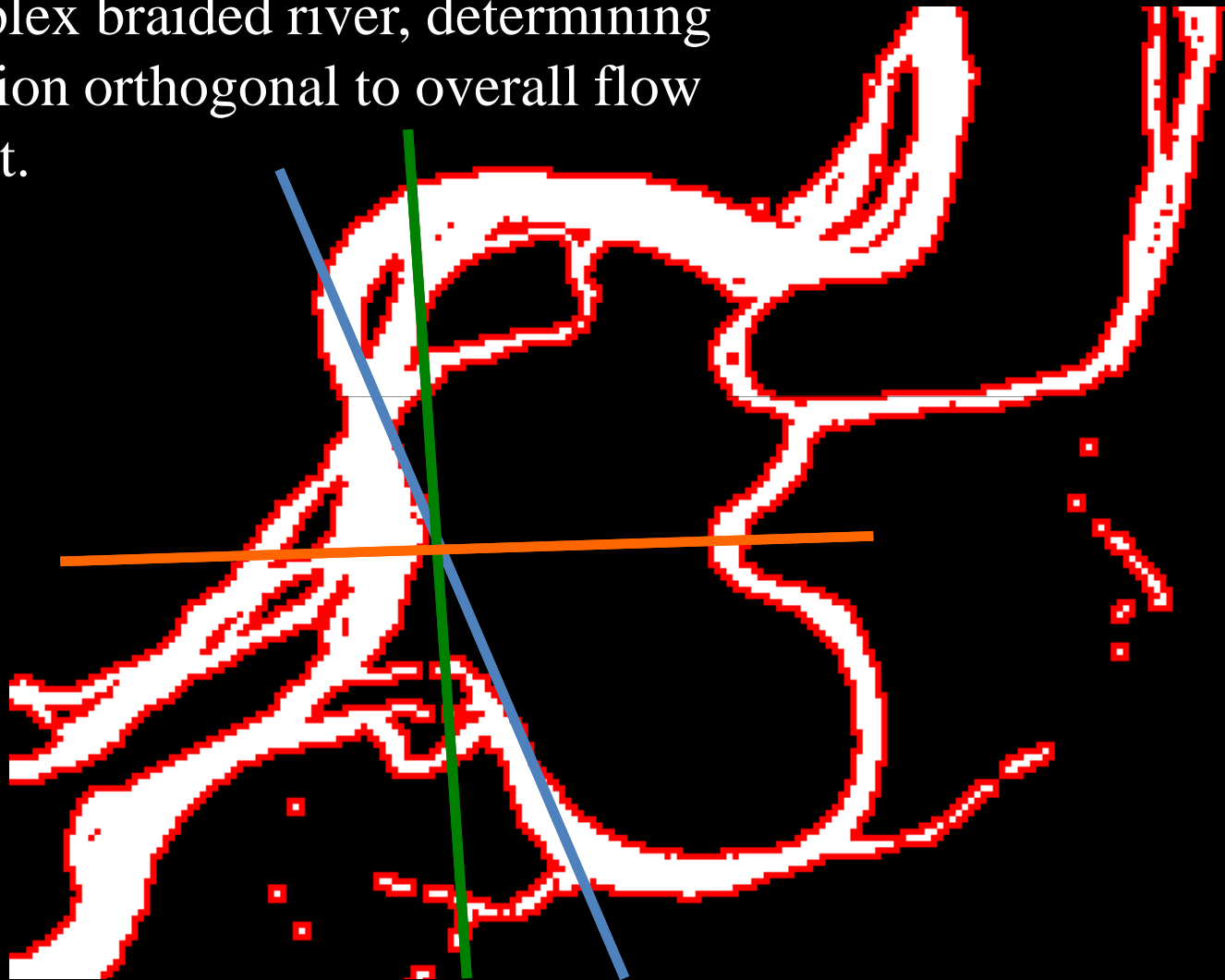


Deriving an accurate cross-section is simple in a single-channel river.

Ohio River

Error in Cross-Section Calculation

In a complex braided river, determining the direction orthogonal to overall flow is difficult.



Tanana River

Sources of Error in Width Classification

- Misclassification of pixels as water/nonwater
 - Sources: Emergent vegetation, wet sediment
- Error in cross-section calculation
 - Source: Complex river planforms
- Boundary/edge effect errors
 - Source: Inherent in calculation of river width from binary inundation masks
 - Defined by the number of channels in a river:

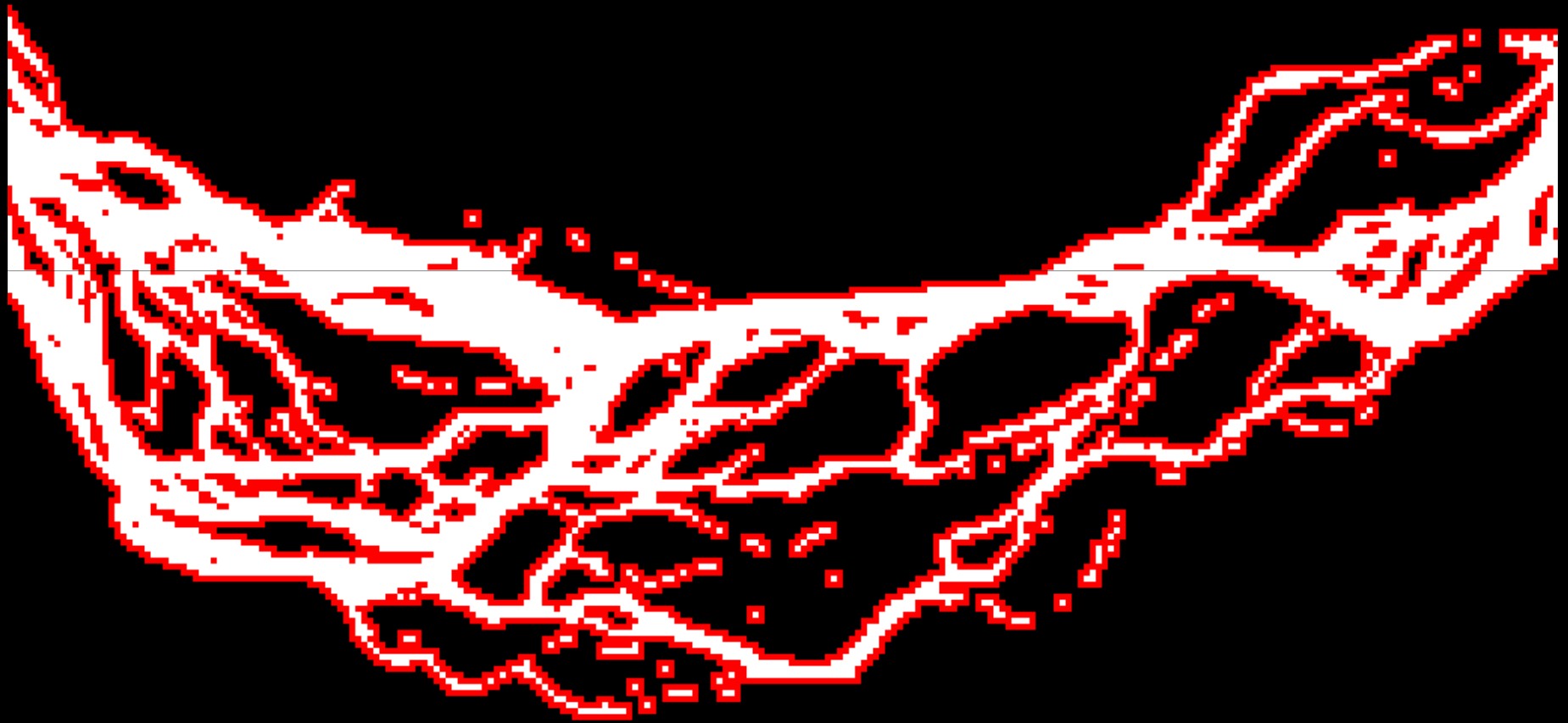
$$E = \frac{1}{2}RC$$

E: edge effect error

R: pixel resolution

C: channel crossings

Boundary/Edge Effect Error



Tanana River

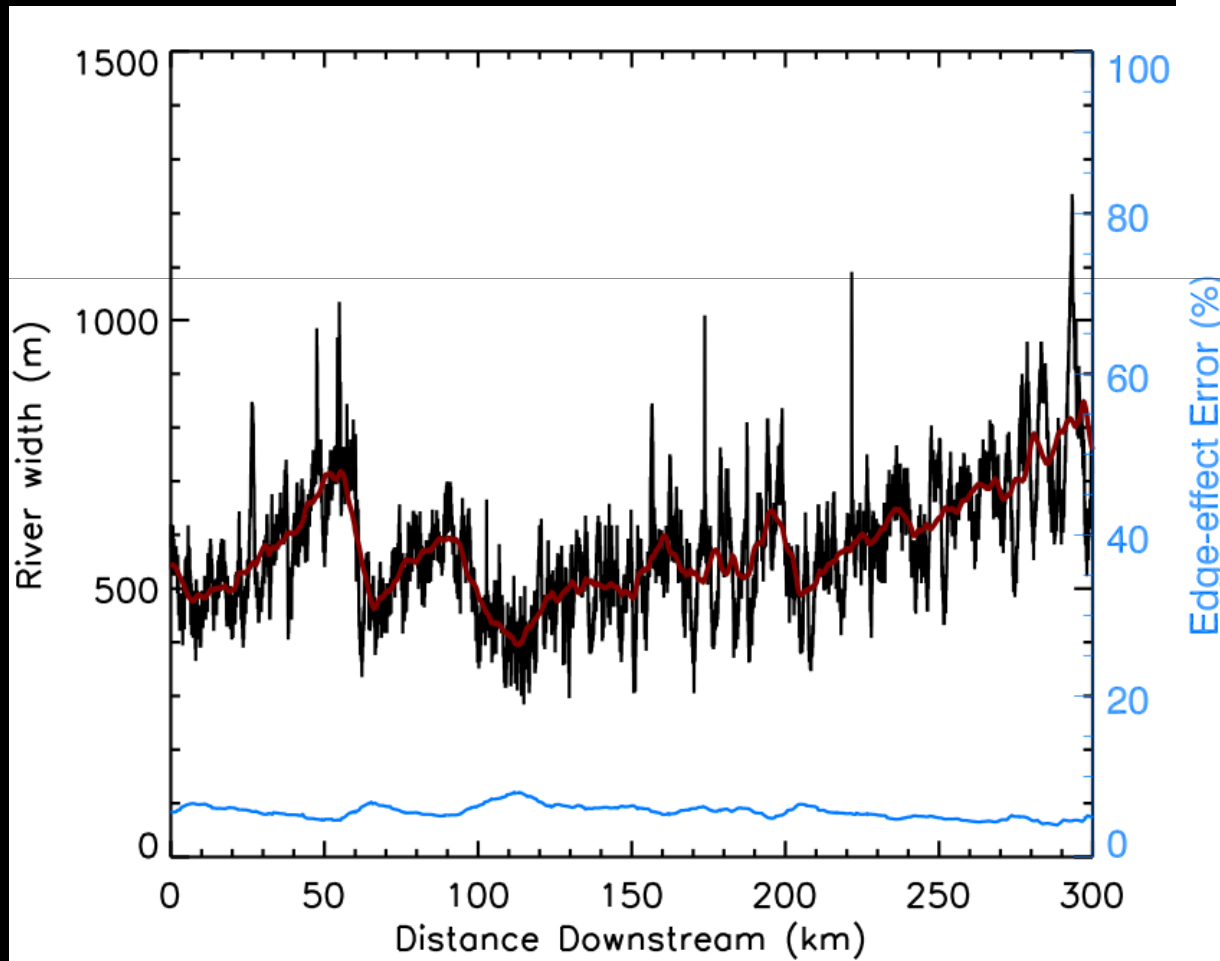
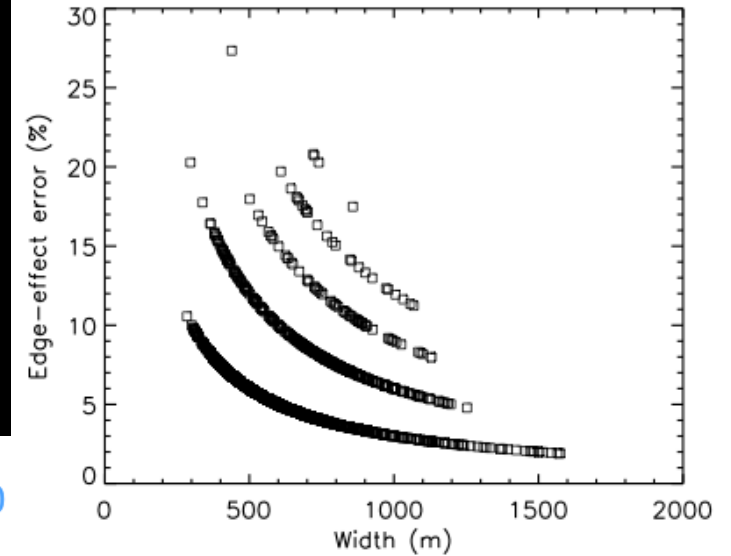
Boundary/Edge Effect Error



Kentucky River

Ohio River

On large, single channel rivers like the Ohio, edge-effect error is relatively small.

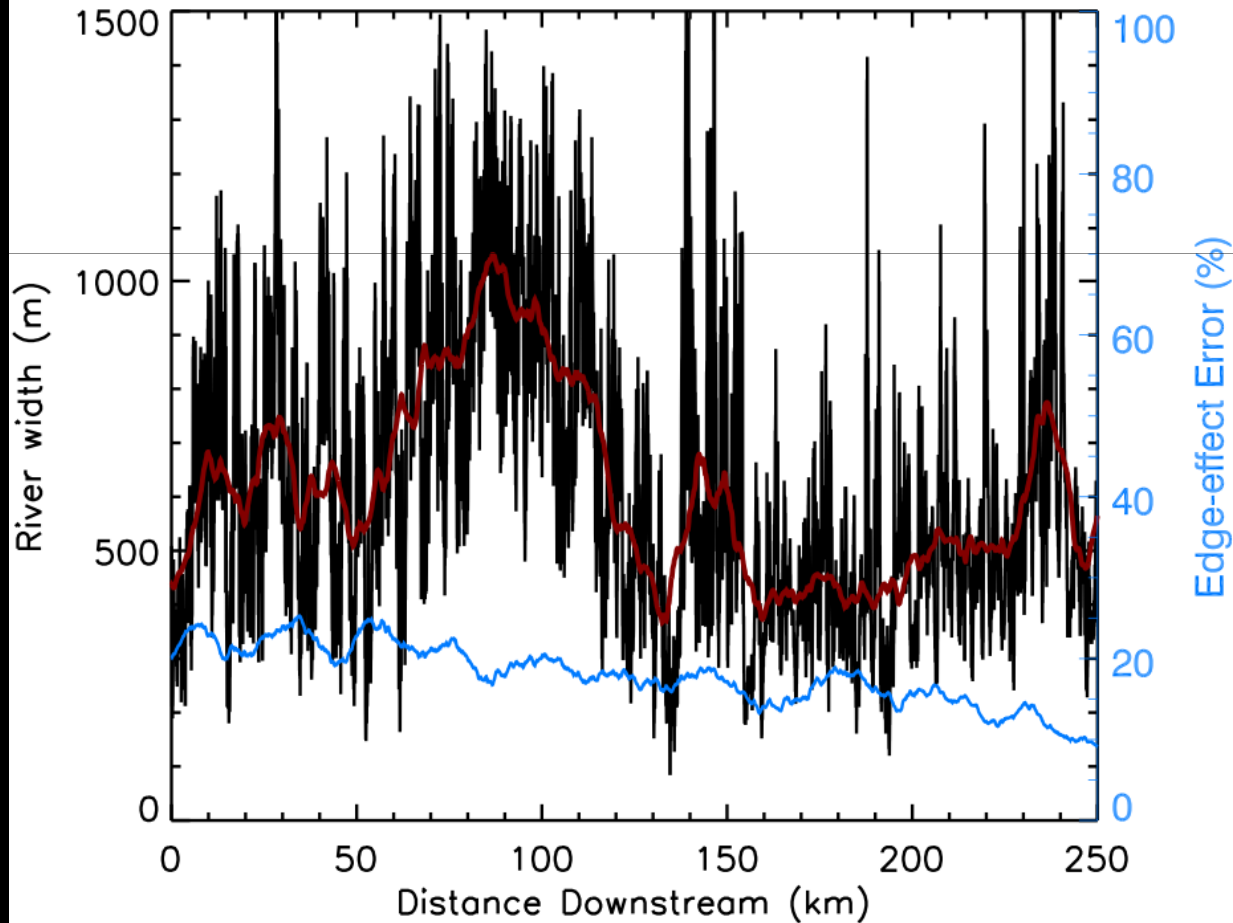
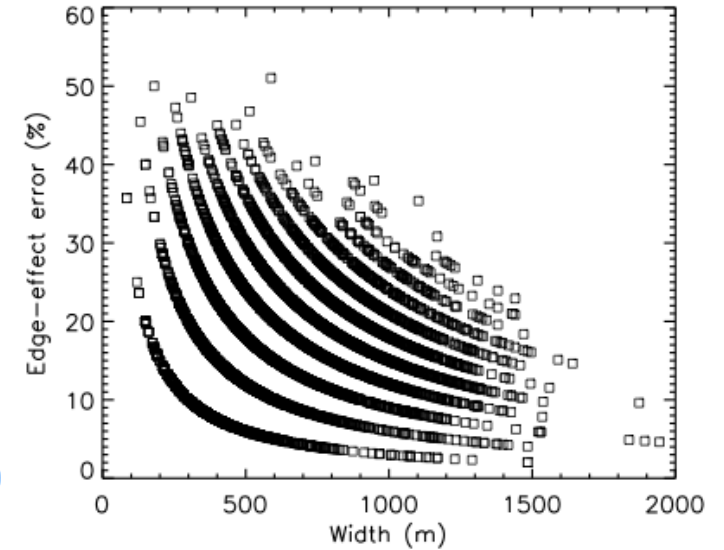


Mean Width: 585 m
Max Width: 1575 m
Min Width: 284 m

Mean Error: 6%
Max Error: 27%
Min Error: 2%

Tanana River

Increased braiding substantially increases edge-effect errors when width is constant.

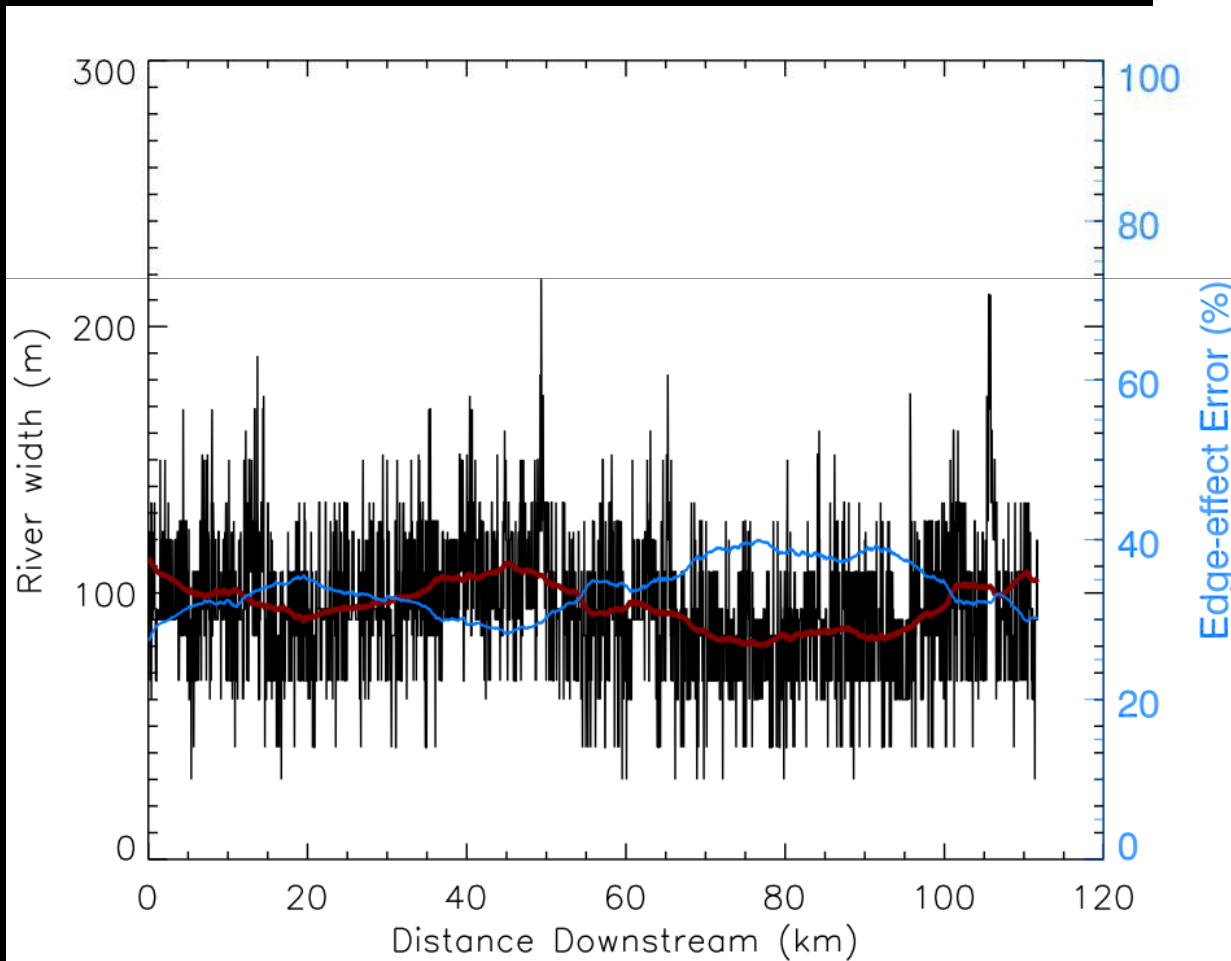
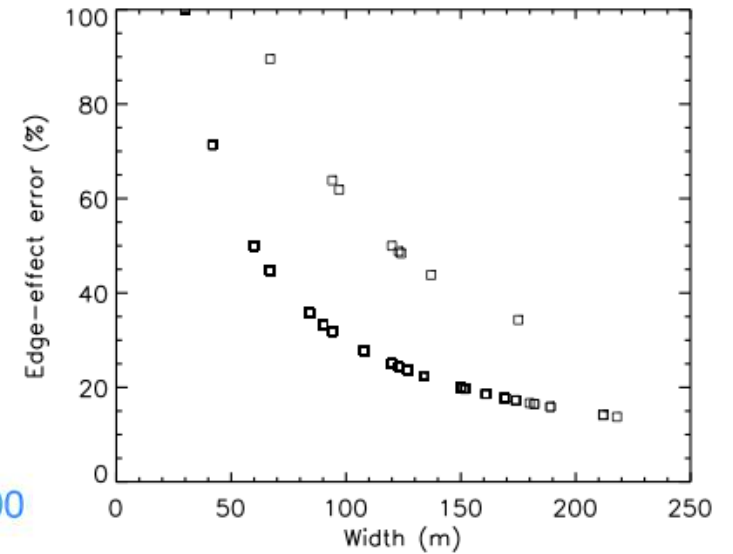


Mean Width: 611 m
Max Width: 1944 m
Min Width: 84 m

Mean Error: 17%
Max Error: 51%
Min Error: 2%

Kentucky River

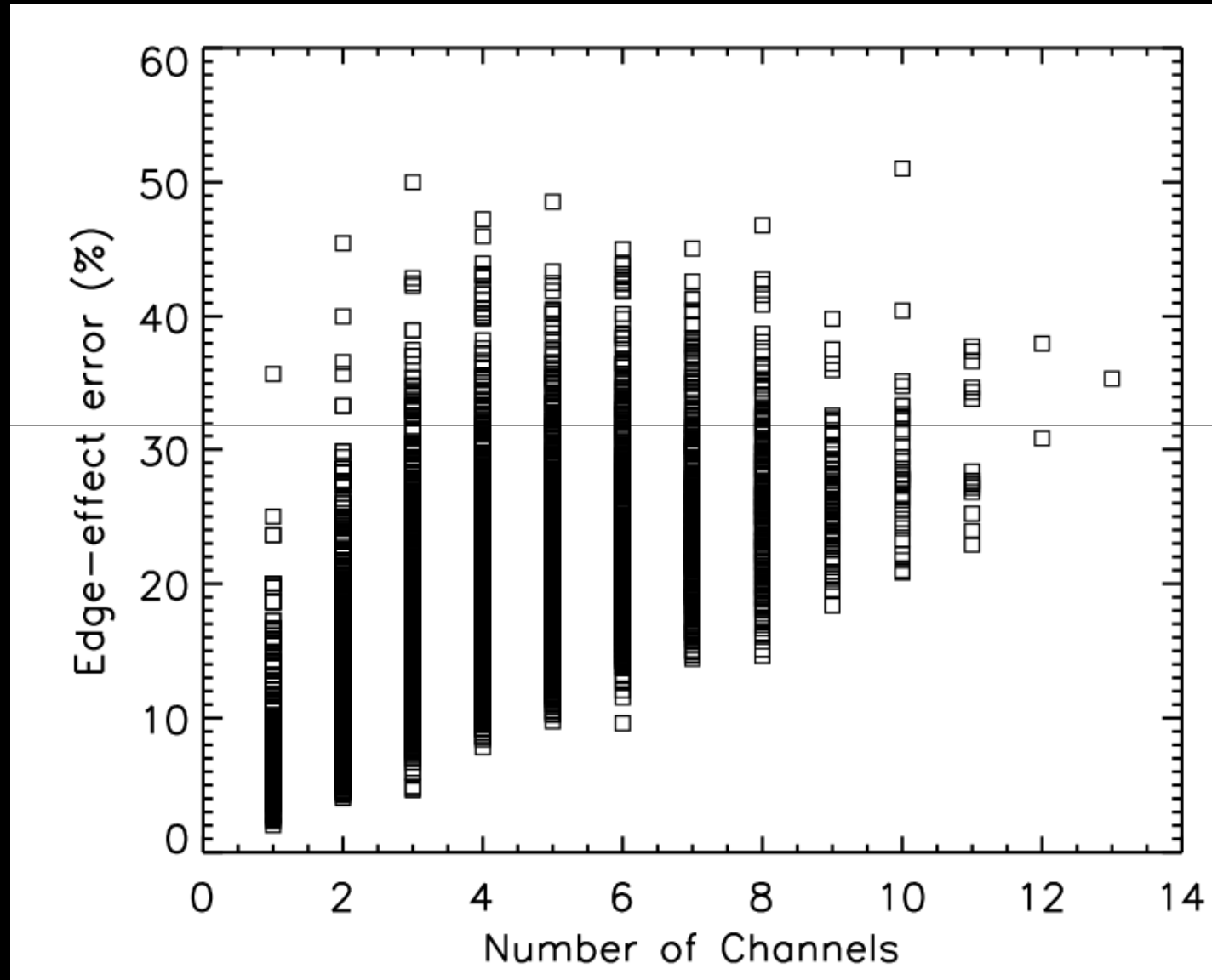
As a river narrows, edge-related uncertainty becomes increasingly important.



Mean Width: 95 m
Max Width: 218 m
Min Width: 30 m

Mean Error: 35%
Max Error: 100%
Min Error: 13%

Edge-Effect Error vs. Number of Channels, Tanana River



Examples of Discharge Errors

Assuming constant width and depth, how will width errors affect discharge calculations?

Ohio

Mean width/error:

4413-4946 m³/s

Maximum error:

2563-4460 m³/s

Minimum error:

6176-6418 m³/s

Tanana

Mean width/error:

1004-1442 m³/s

Maximum error:

576-1775 m³/s

Minimum error:

1454-1513 m³/s

Kentucky

Mean width/error:

125-254 m³/s

Maximum error:

0-300 m³/s

Minimum error:

210-270 m³/s

Ways to Minimize Width Error

- Limit cross-sections used to those below a given error threshold.
- Develop a method to reliably extract subpixel inundation information
- Average widths across multiple adjacent cross-sections or switch from width to area.



Ways to Minimize Width Error

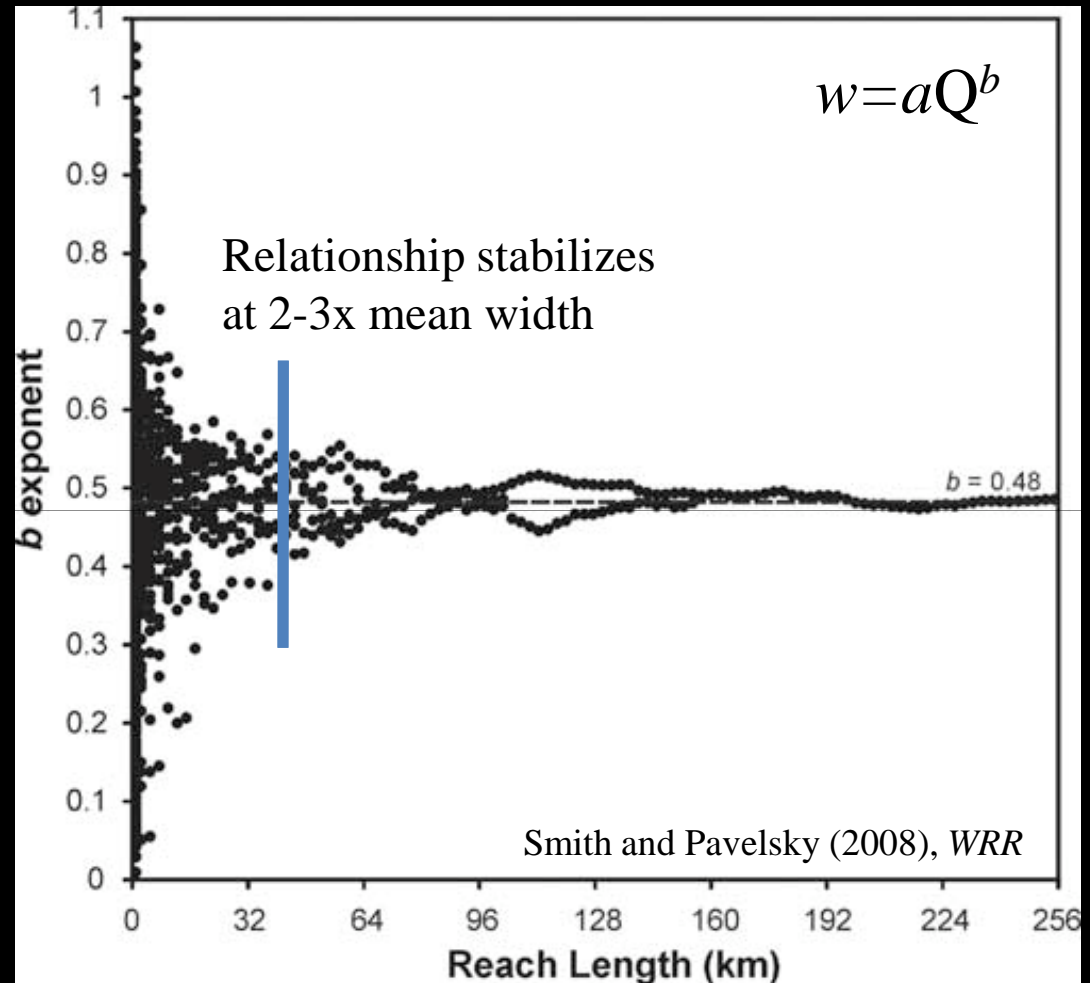
- Limit cross-sections used to those below a given error threshold.
- Develop a method to reliably extract subpixel inundation information
- Average widths across multiple adjacent cross-sections or switch from width to area.



Most methods for extracting subpixel information on inundation use either multiple optical bands or a combination of optical and microwave sensors. Can a method be developed to extract subpixel inundation information from SWOT-like data alone?

Ways to Minimize Width Error

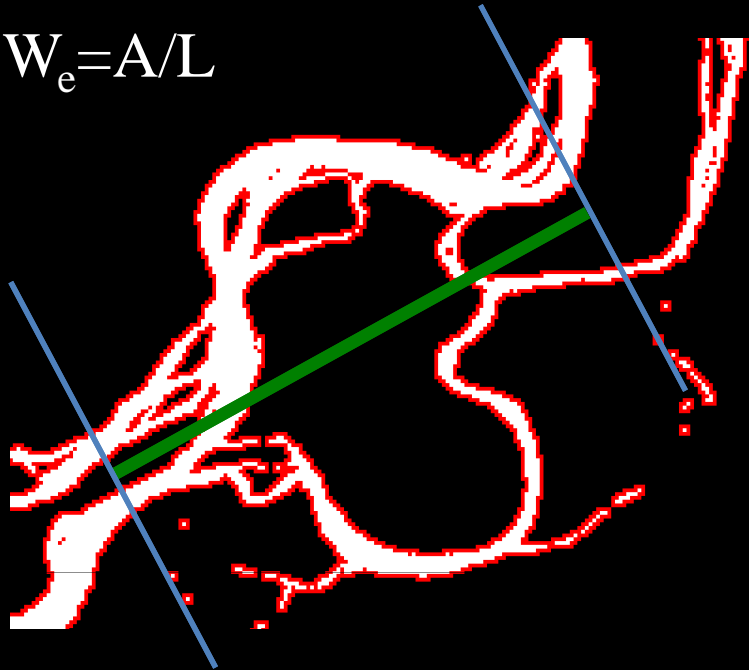
- Limit reaches used to those below a given error threshold.
- Develop a method to reliably extract subpixel inundation information
- Average widths across multiple adjacent cross-sections or switch from width to area.



Relationship between reach length and the coefficient in the width-discharge power law equation for the Lena River, Siberia.

Effective width vs. Averaged width

$$W_e = A/L$$



Advantages:

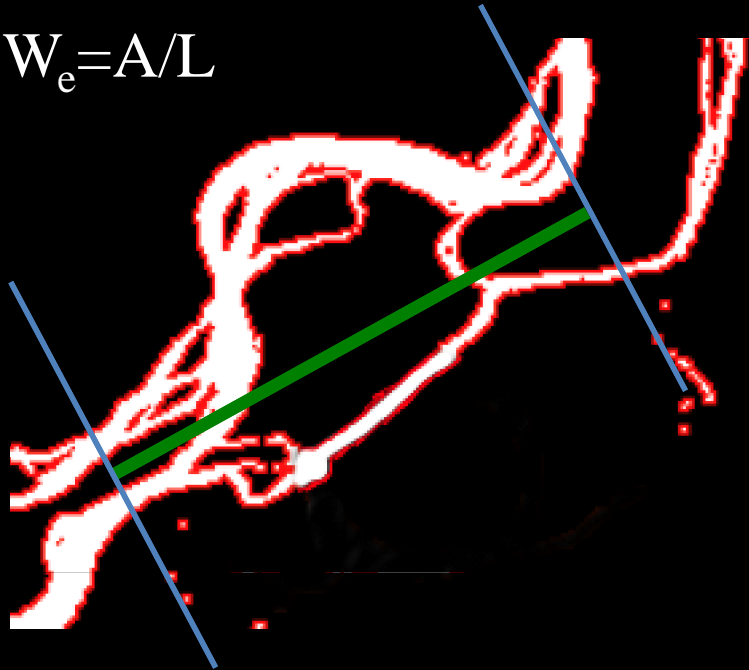
Fast, easy to compute

Disadvantages:

Varies depending on
the sinuosity of
channels within reach

Effective width vs. Averaged width

$$W_e = A/L$$



Advantages:

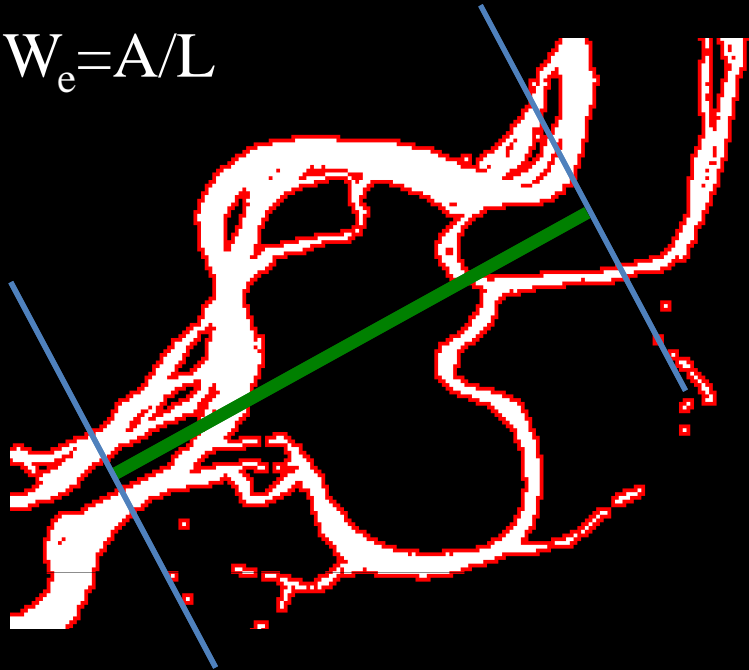
Fast, easy to compute

Disadvantages:

Varies depending on
the sinuosity of
channels within reach

Effective width vs. Averaged width

$$W_e = A/L$$



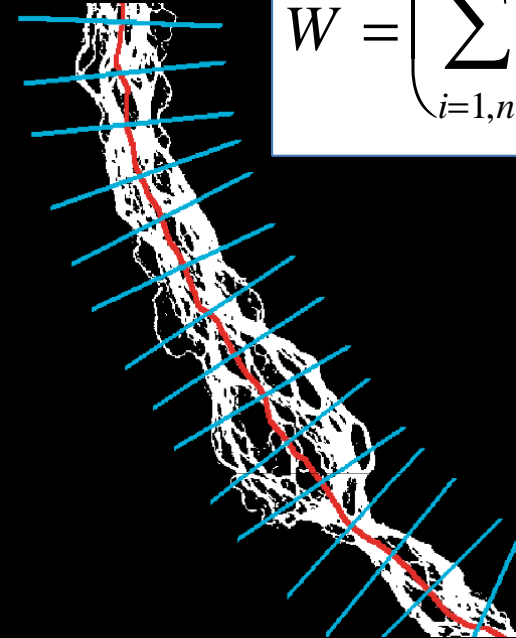
Advantages:

Fast, easy to compute

Disadvantages:

Varies depending on the sinuosity of channels within reach

$$\bar{W} = \left(\sum_{i=1,n} w_i \right) / n$$



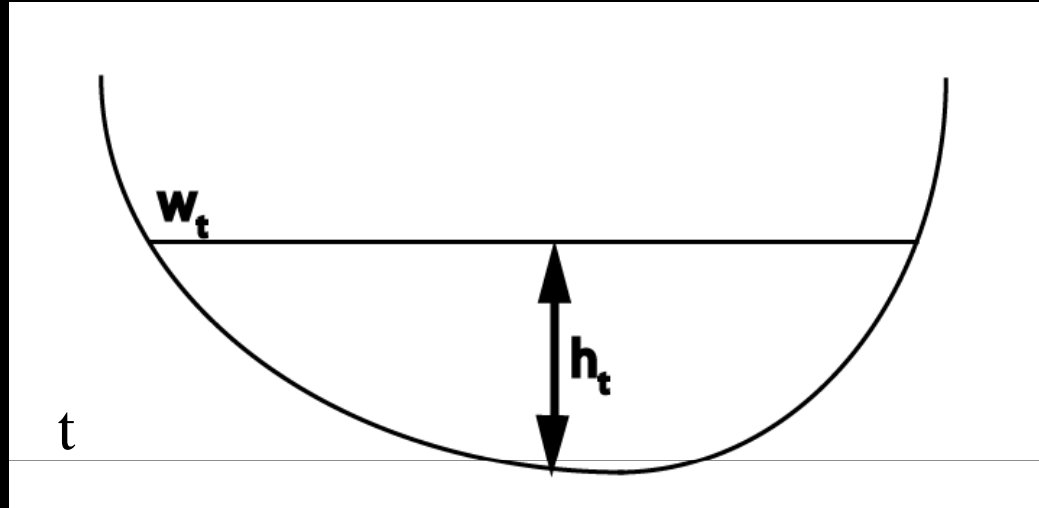
Advantages:

More precise widths, easier to compare to gauge data

Disadvantages:

More difficult to calculate, problems with orthogonals.

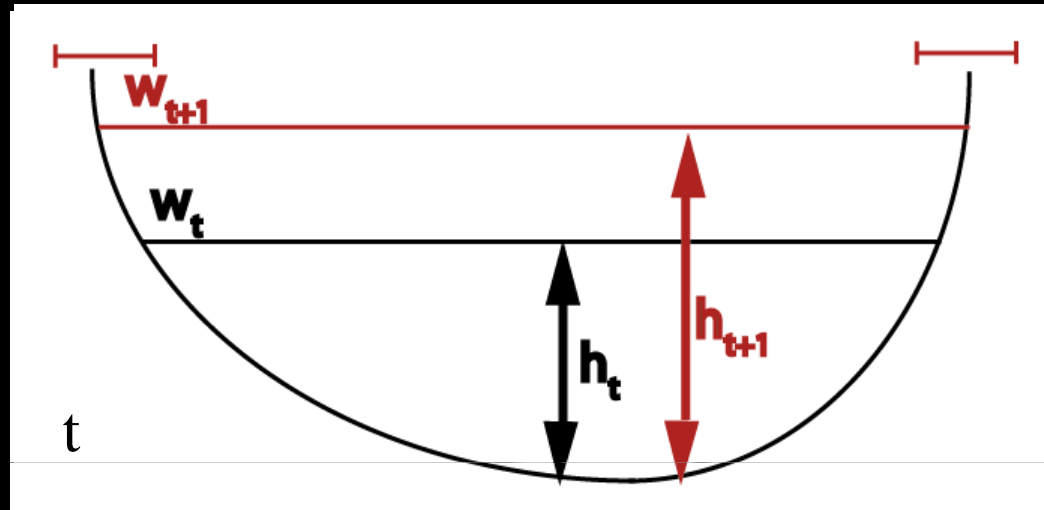
Constraining Widths with Multitemporal Data



Multitemporal SWOT data will allow us to constrain width errors further.

Given two overpasses at times t and $t+1$, if $h_{t+1} > h_t$ then w_{t+1} also must be greater than w_t .

Constraining Widths with Multitemporal Data

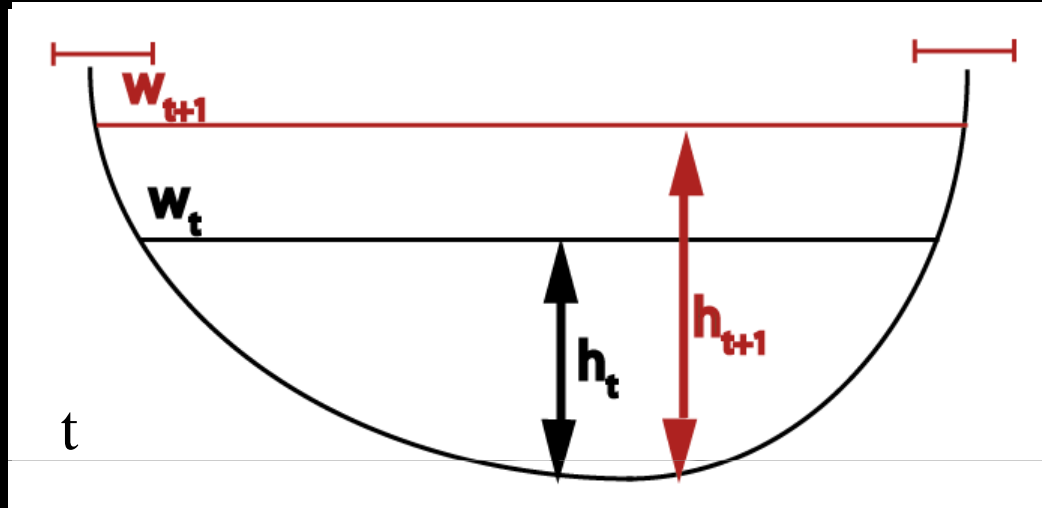


Multitemporal SWOT data will allow us to constrain width errors further.

Given two overpasses at times t and $t+1$, if $h_{t+1} > h_t$ then w_{t+1} also must be greater than w_t .

This method will be particularly useful because high-resolution overpasses can be used to constrain low-resolution overpass widths.

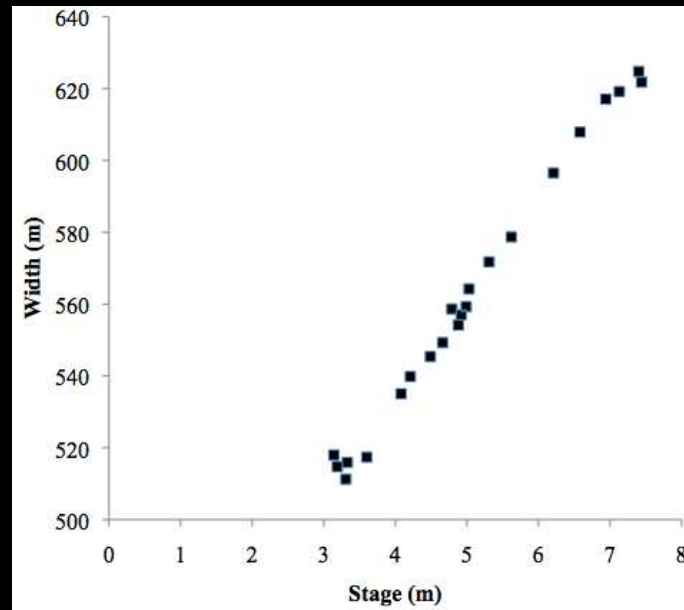
Constraining Widths with Multitemporal Data



Multitemporal SWOT data will allow us to constrain width errors further.

Given two overpasses at times t and $t+1$, if $h_{t+1} > h_t$ then w_{t+1} also must be greater than w_t .

Using many time steps, we can build up SWOT-based width-depth rating curves



This method will be particularly useful because high-resolution overpasses can be used to constrain low-resolution overpass widths.

Next Steps

- Use AirSWOT to validate and calibrate inundation retrievals
- Develop a system for classification of river reaches for SWOT discharge retrieval
 - Map likely sources of uncertainty
 - Adjustment based on Width vs. Depth
- Release an improved version of RivWidth