



SWOT Global discharge accuracy

presented by Steve Coss

On Behalf of the DAWG

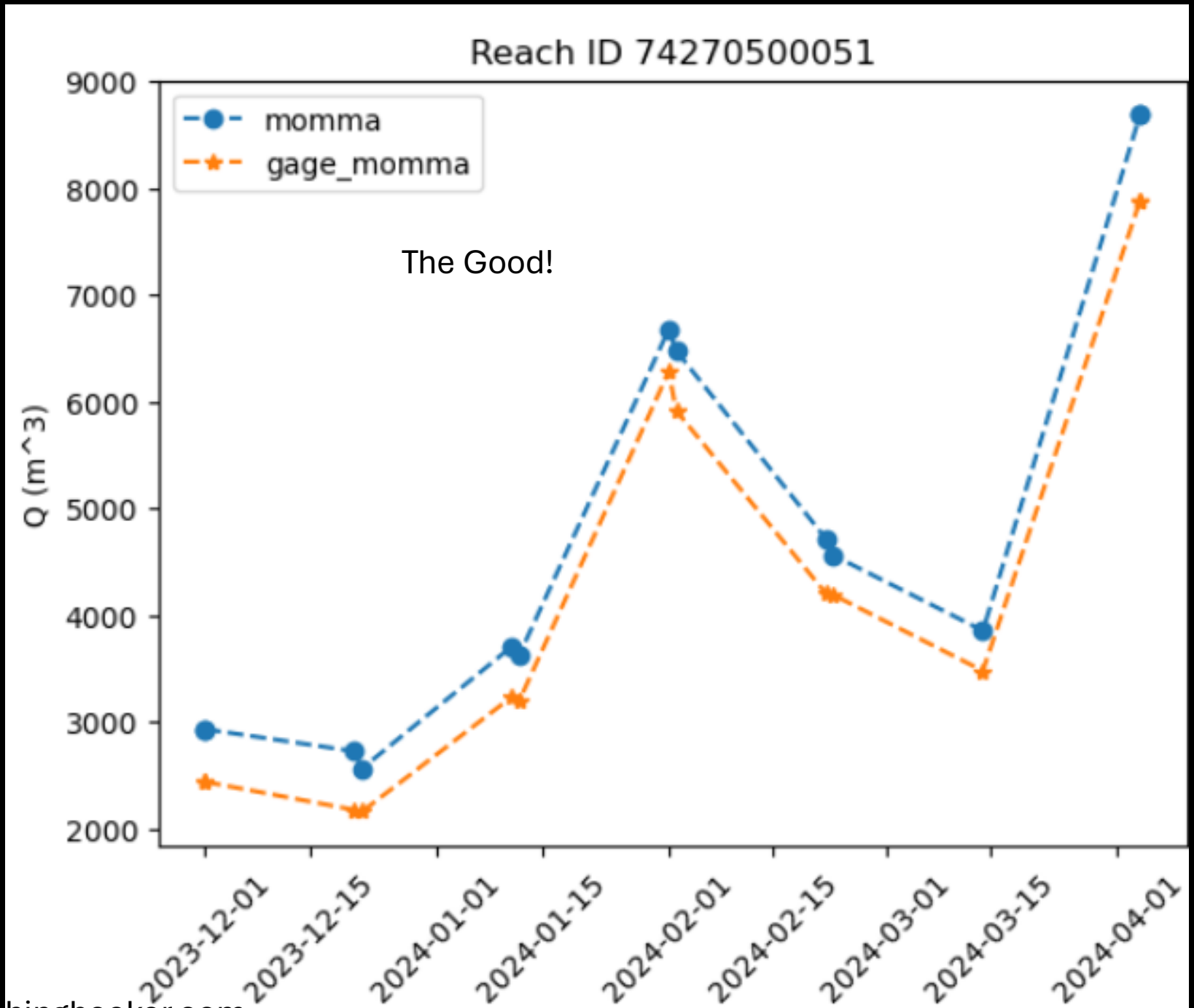
Illustration

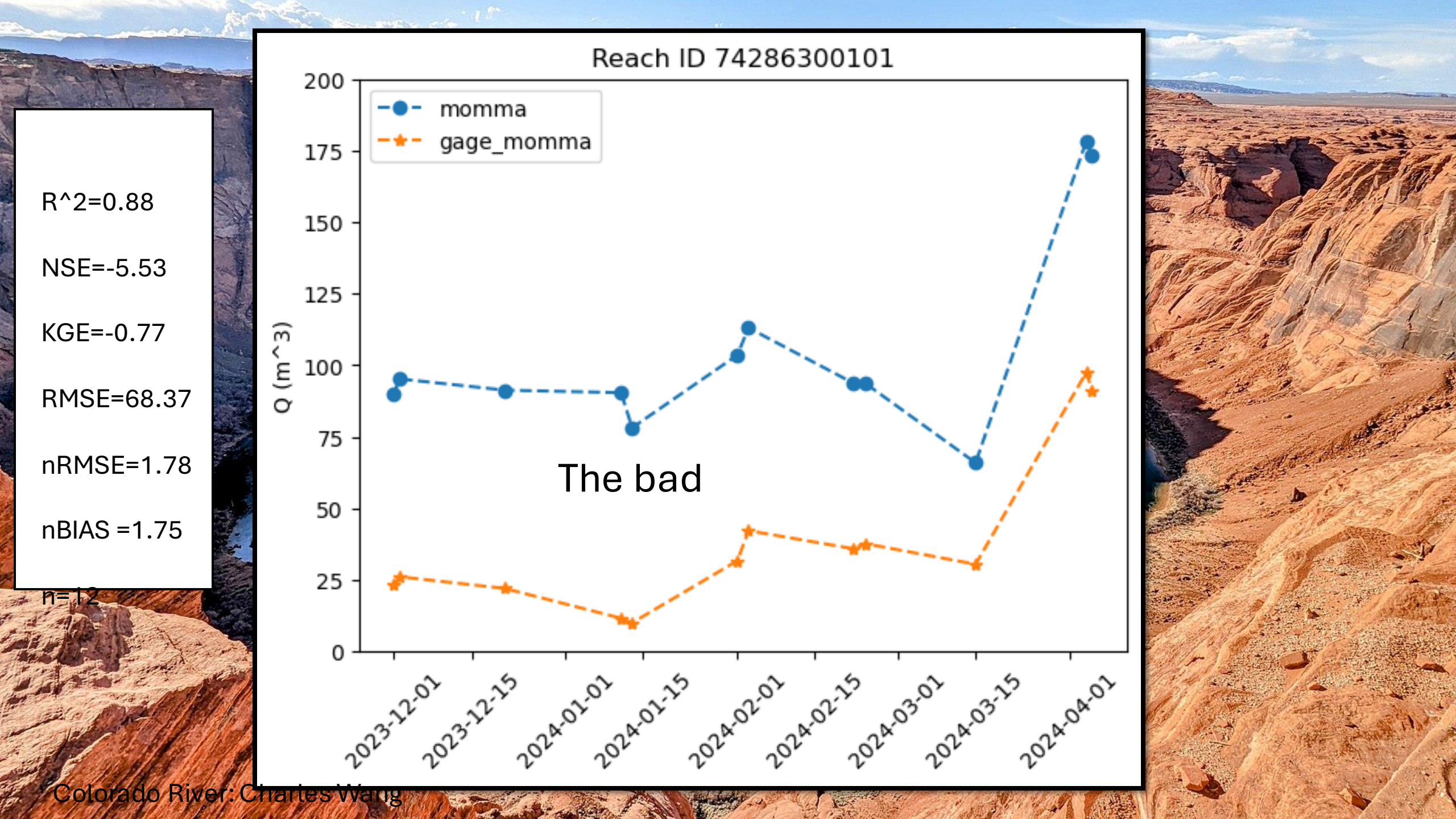
Credit: CNES

The Run

- we have an unconstrained run on **1646** gaged reaches
- we created sets at each gaged reach, then filtered by whether SWOT data was available
- gages spanned **seven** different gage agencies on **four** continents
- **1492 gaged reaches** were ultimately evaluated in the run
- This run was completed on Monday, June 10

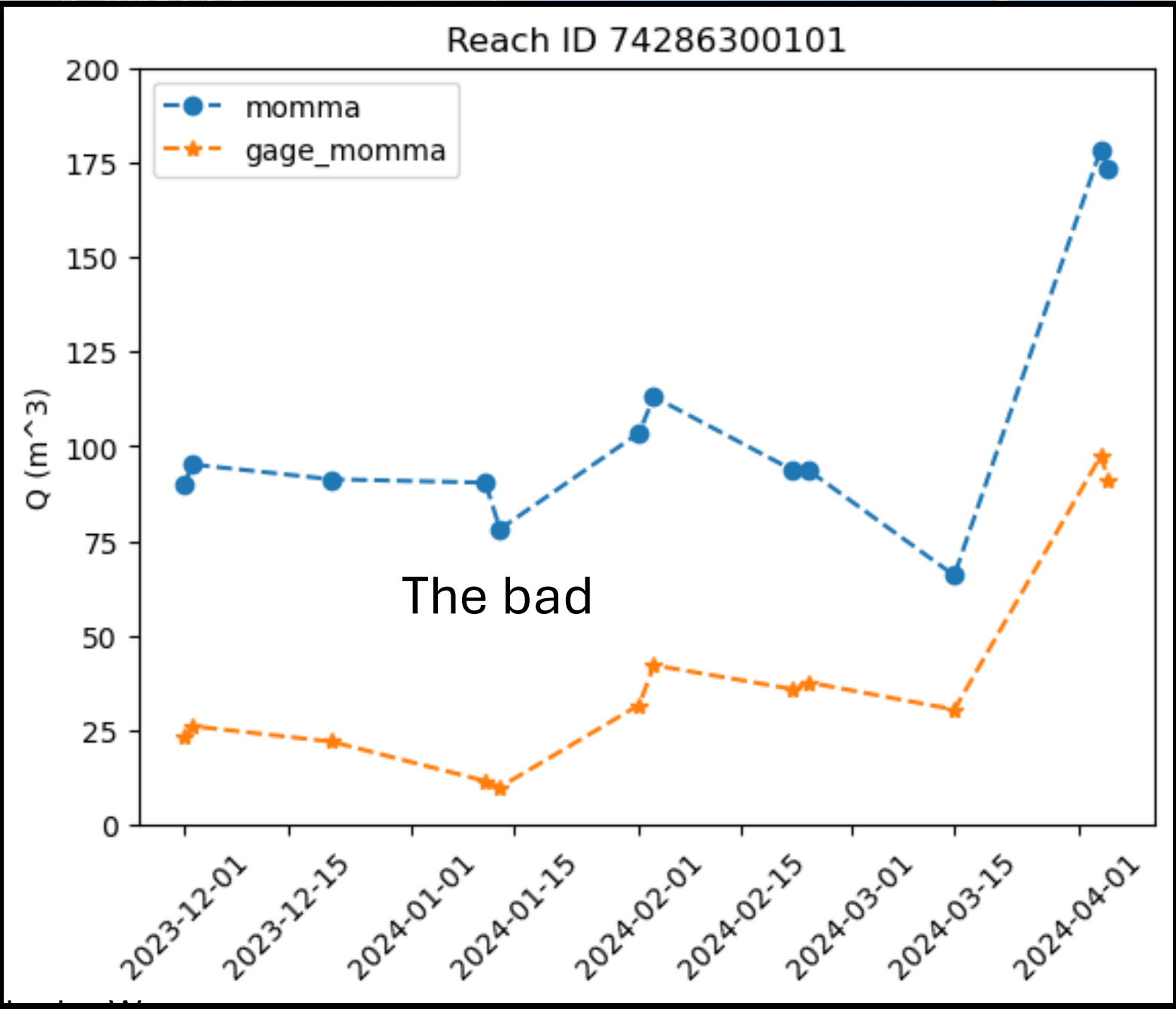
R²=0.99
NSE=0.93
KGE=0.87
RMSE=531.01
nRMSE=0.12
nBIAS =0.12
n=13





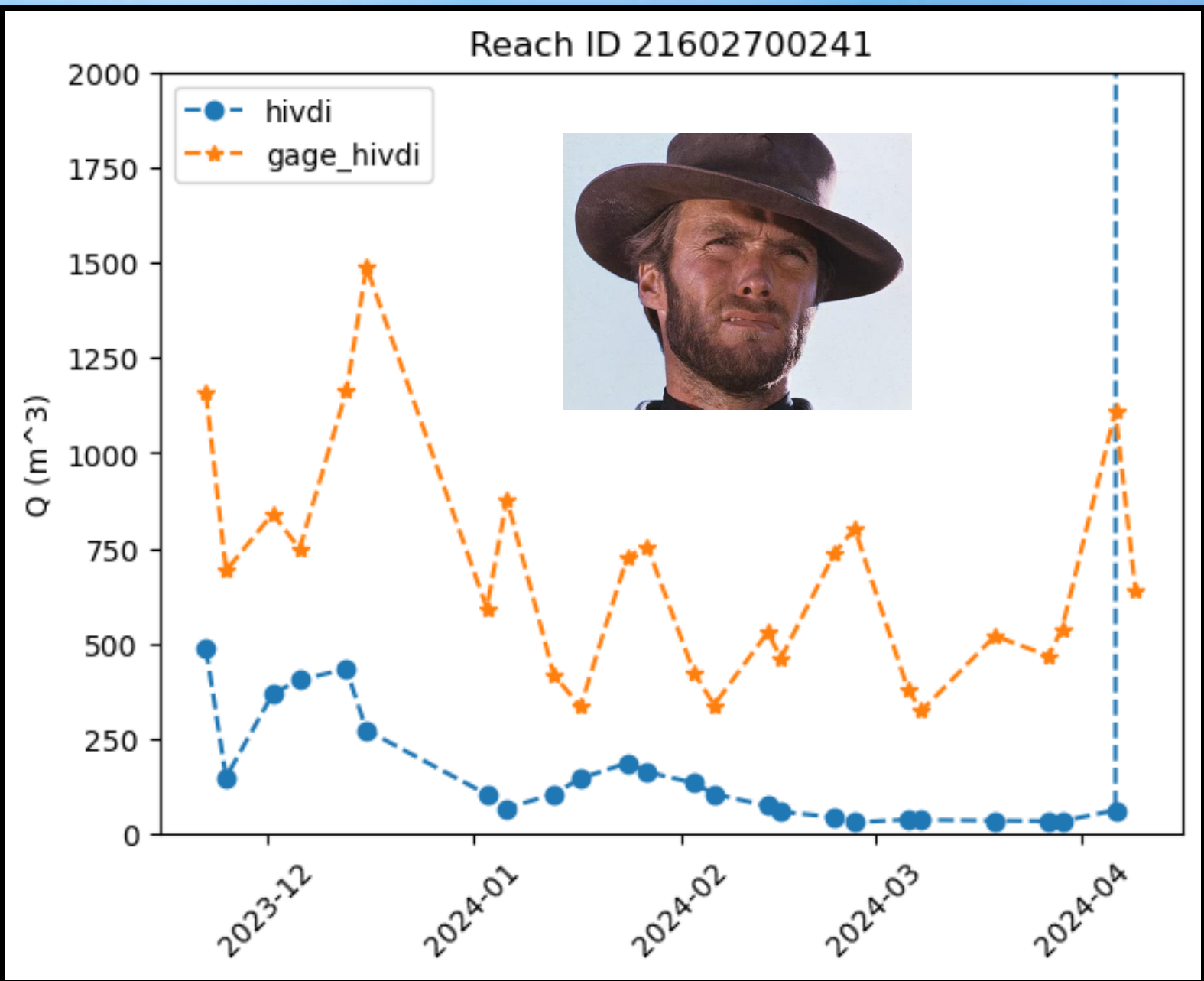
$R^2=0.88$
NSE=-5.53
KGE=-0.77
RMSE=68.37
nRMSE=1.78
nBIAS =1.75

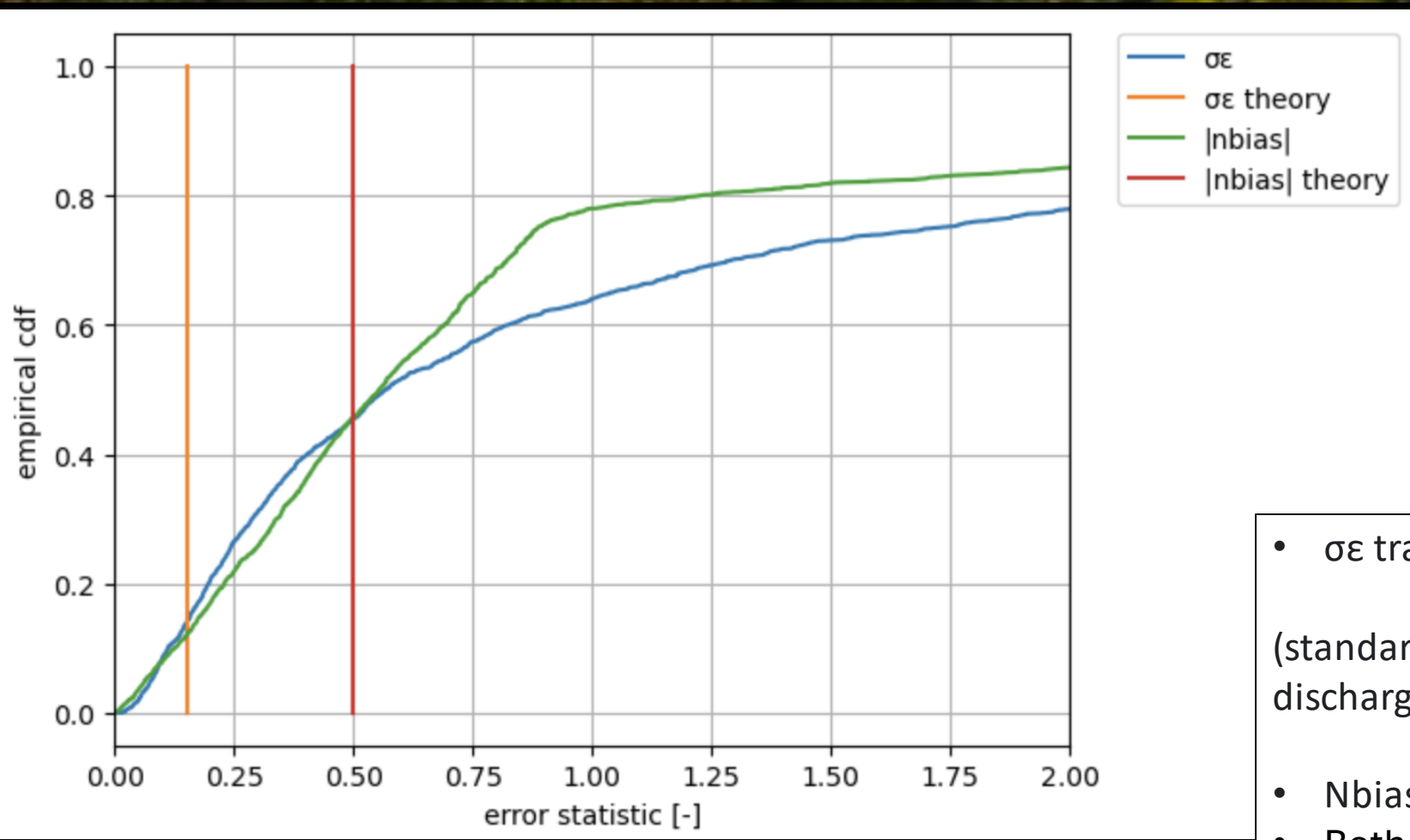
n=12



The bad

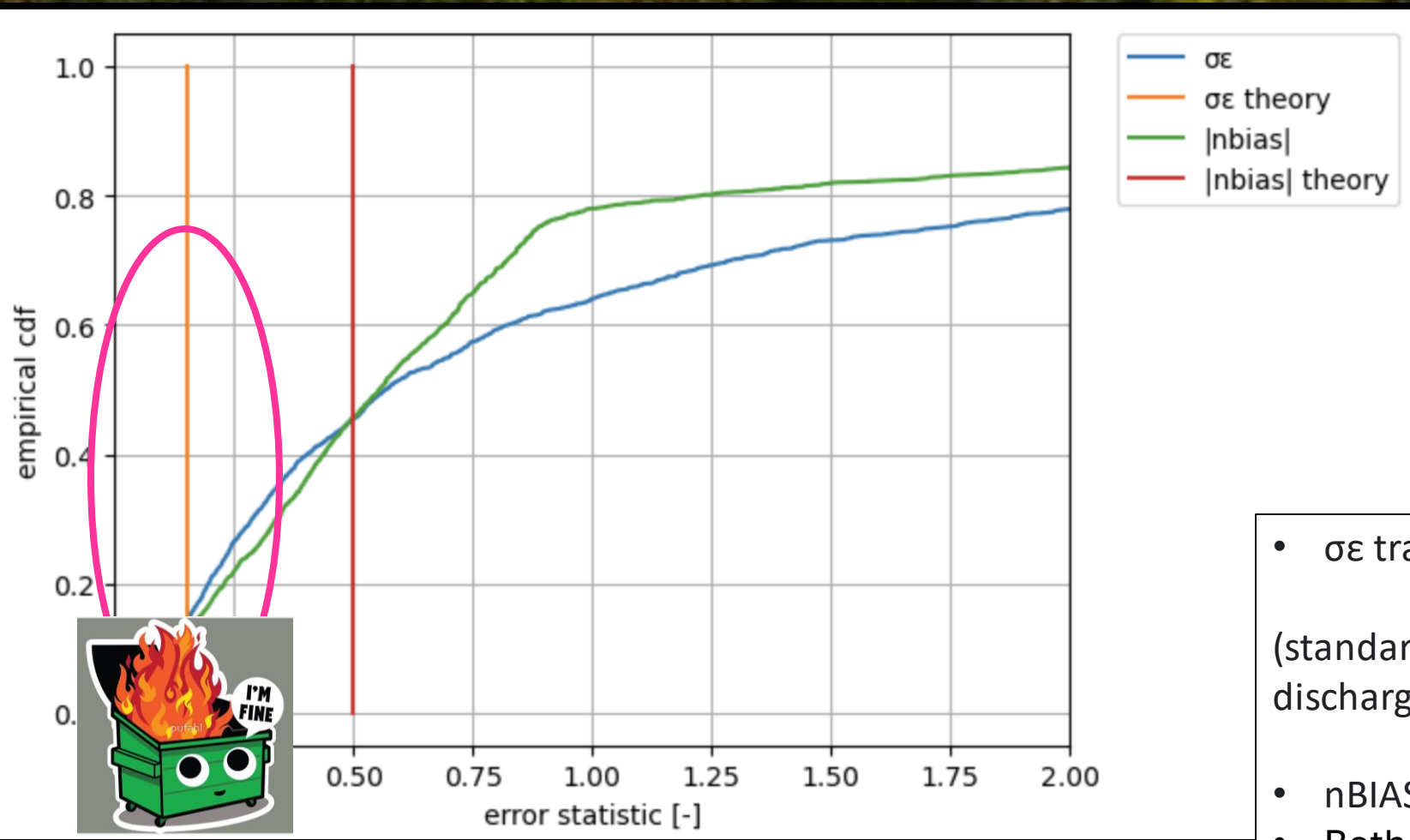
$R^2=0.0006$
 $NSE=-3824.33$
 $KGE=-59.08$
 $RMSE=18142.90$
 $nRMSE=26.56$
 $nBIAS = 4.56$
 $n=25$





- $\sigma\epsilon$ tracks error in dynamics
(standard deviation of the unnormalized discharge error, divided by the mean flow)
- Nbias (gage mean normalized)
- Both error in dynamics and bias are exceeding expectations (parenthetical)
- 68%tile of $\sigma\epsilon$ at 100% (15%)
- 68%tile nBIAS at 75%(50%)

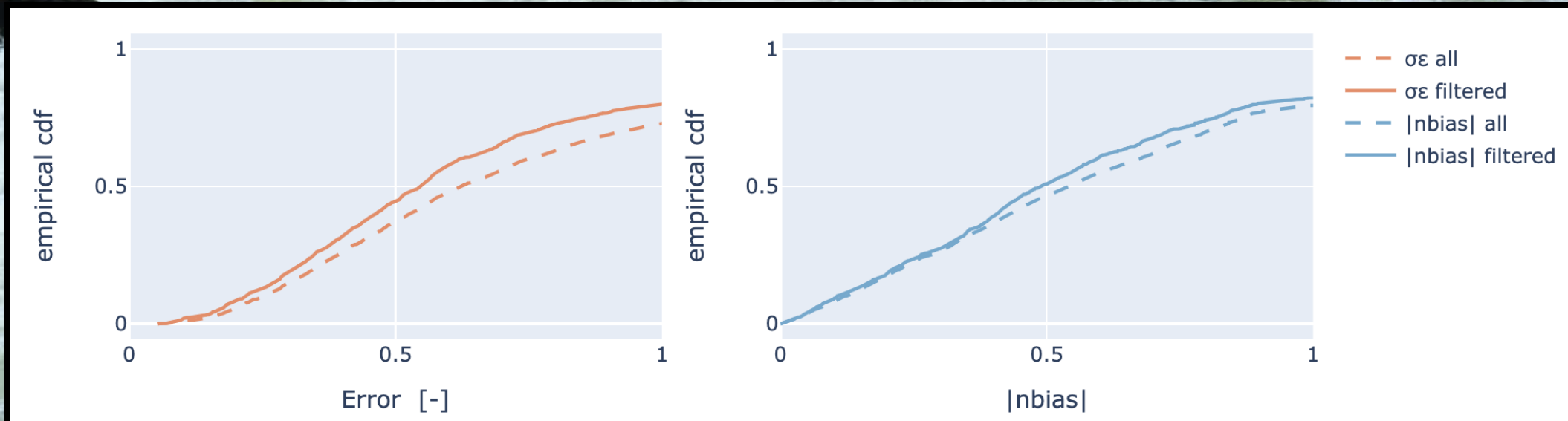
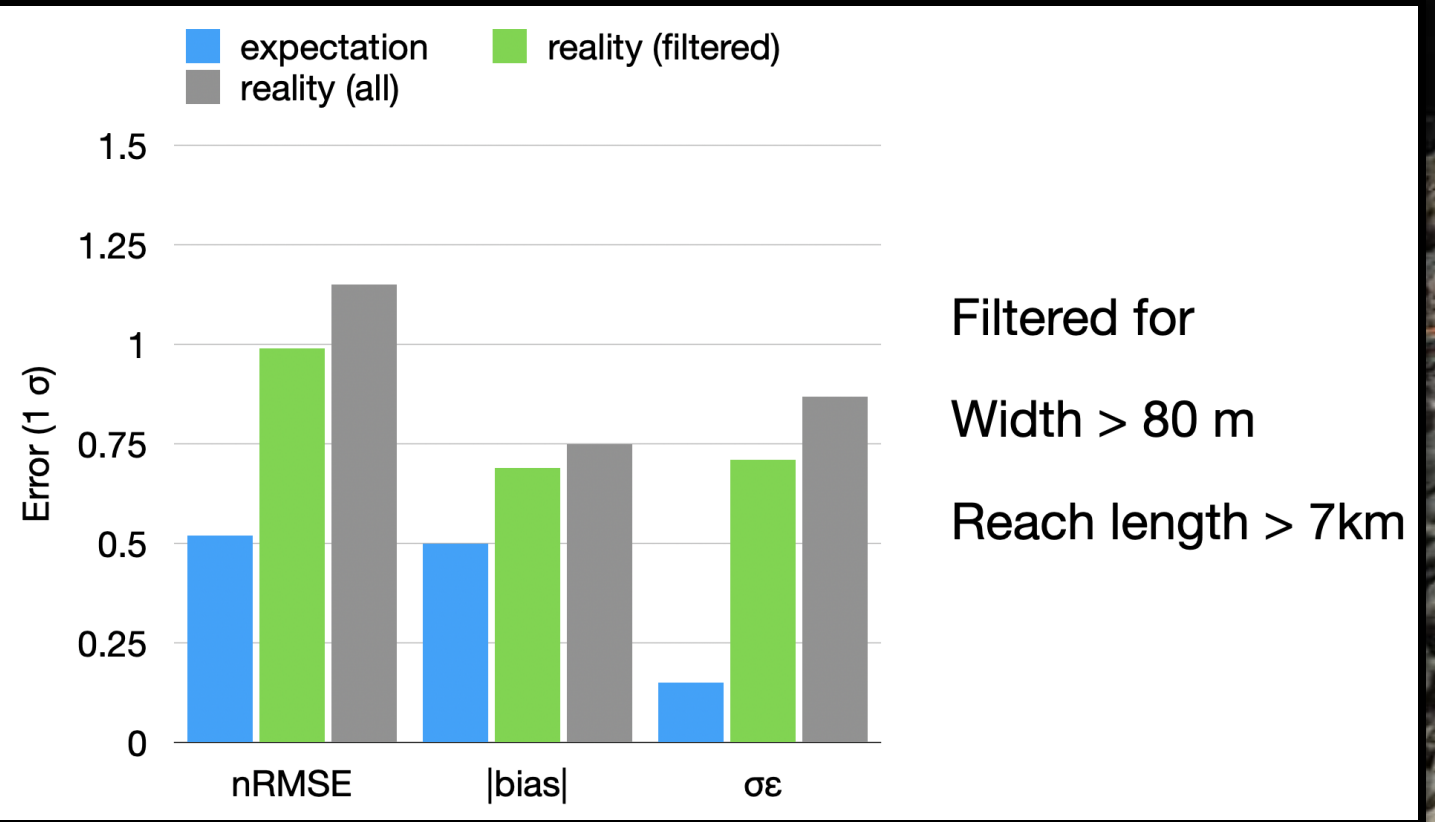
Durand e al



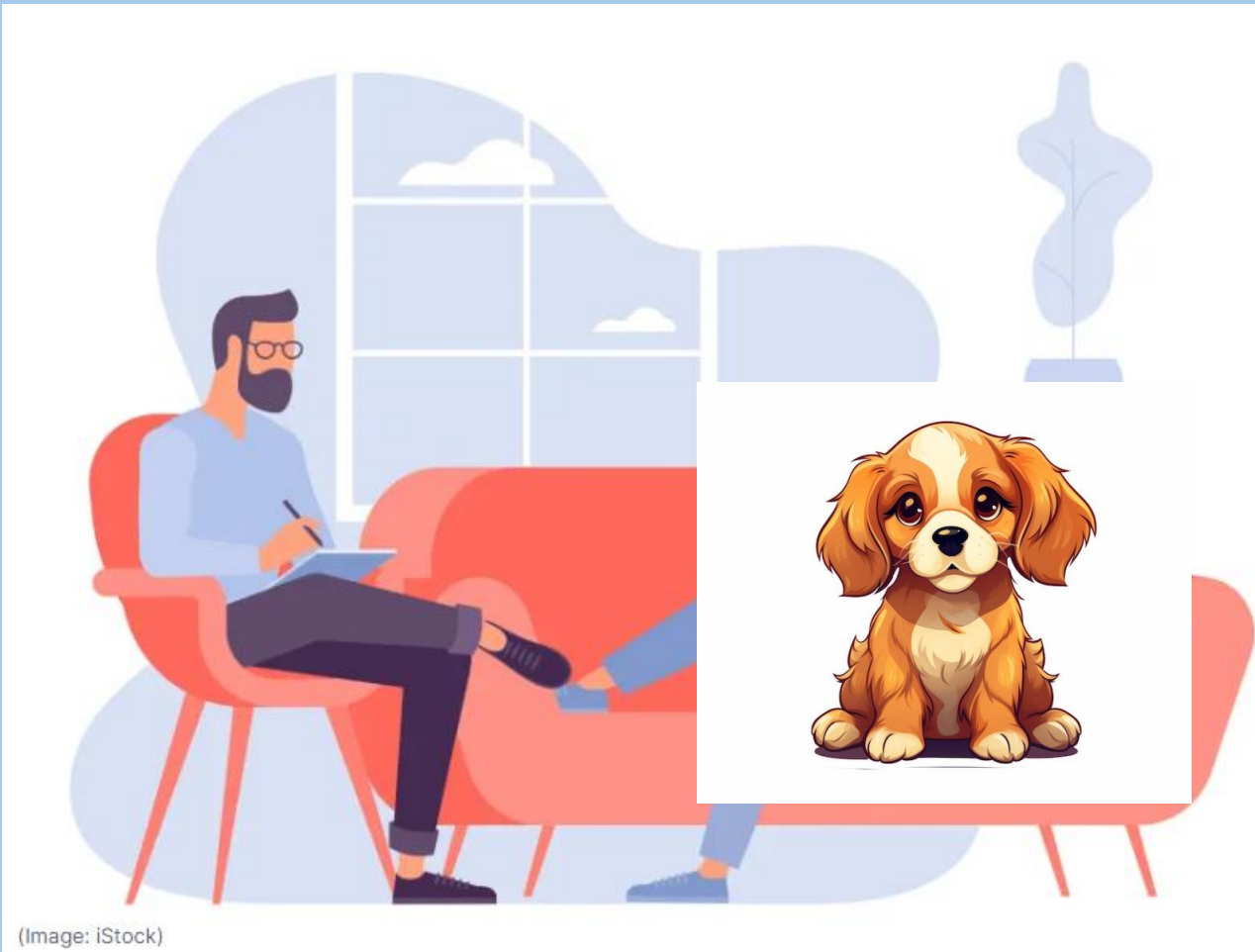
- σ_ϵ tracks error in dynamics
(standard deviation of the unnormalized discharge error, divided by the mean flow)
- nBIAS (gage mean normalized)
- Both error in dynamics and bias are exceeding expectations (parenthetical)
- 68%tile of σ_ϵ at 100% (15%)
- 68%tile nBIAS at 75%(50%)

Durand e al

Error summary



Something worth deep consideration



(Image: iStock)

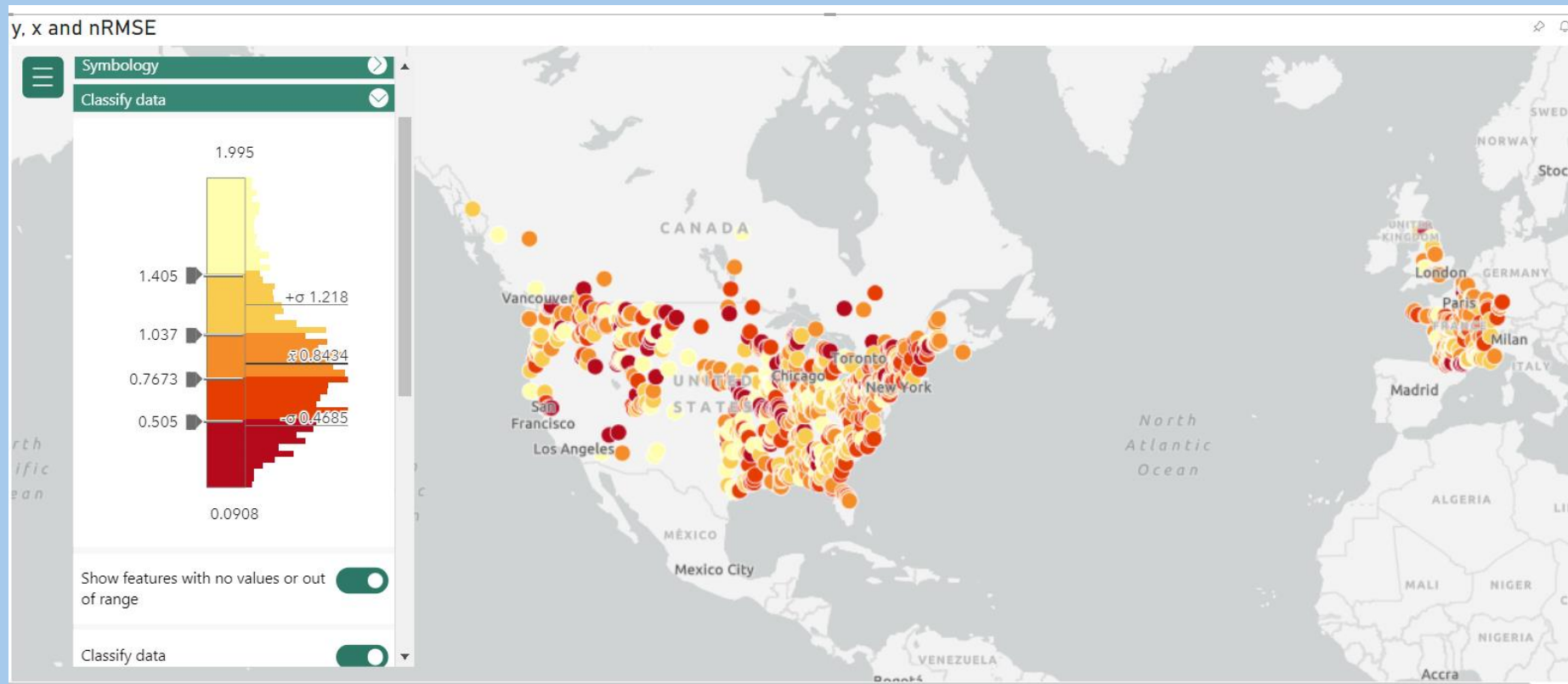
- As this is an unconstrained (no integrator) run stats that include bias are not expected to be all that great
- We are concerned about stats that should preclude bias being as impacted as they were
- We need to better understand underlying causes for why we are not tracking discharge variation

Thanks everyone who came to the workshop Sunday!



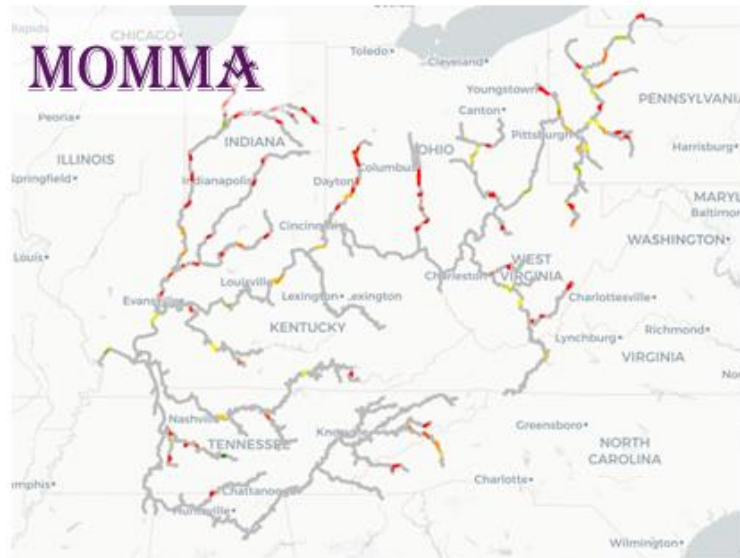
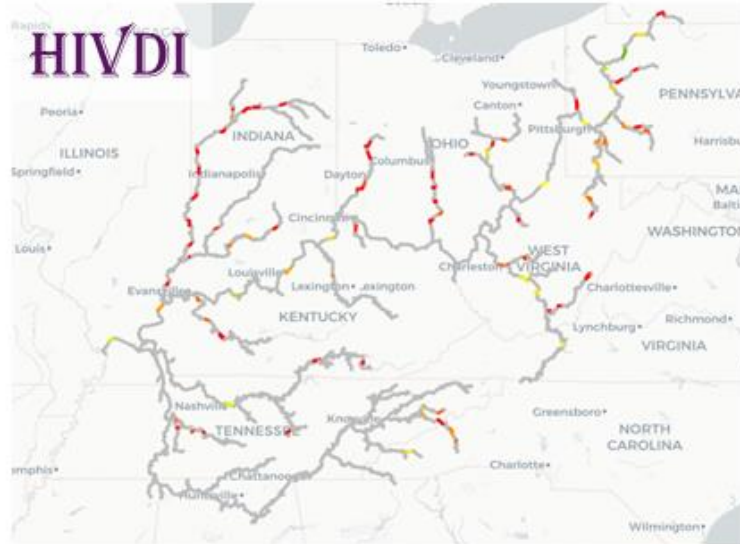
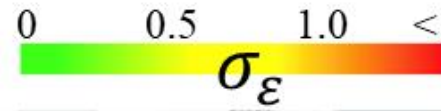
We searched for correlation in error with.....

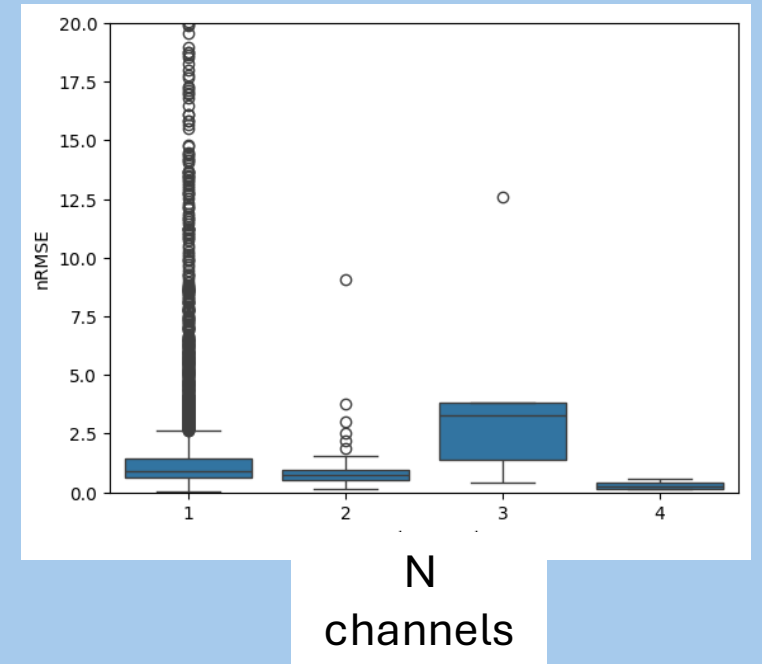
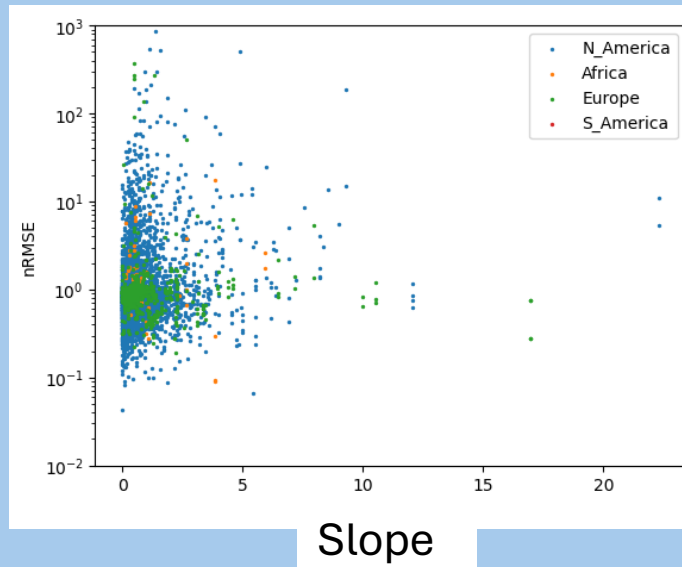
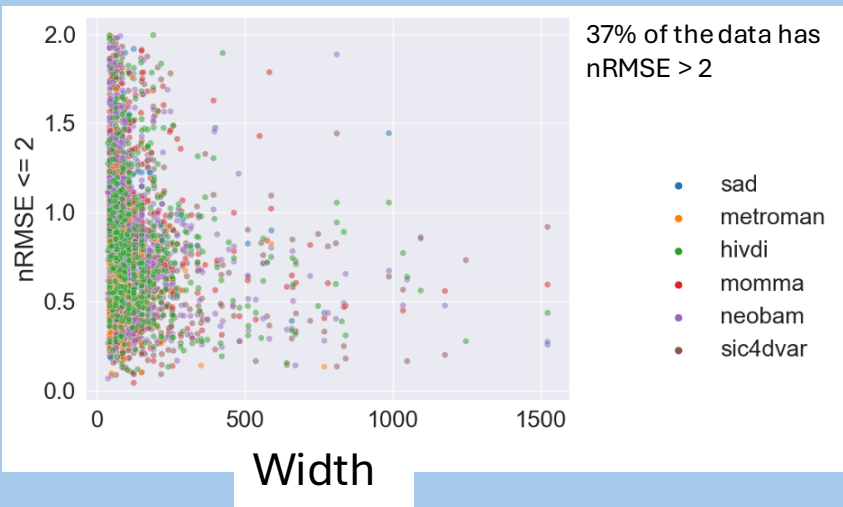
We were able to eliminate a few possibilities



No obvious spatial patterns in nRMSE

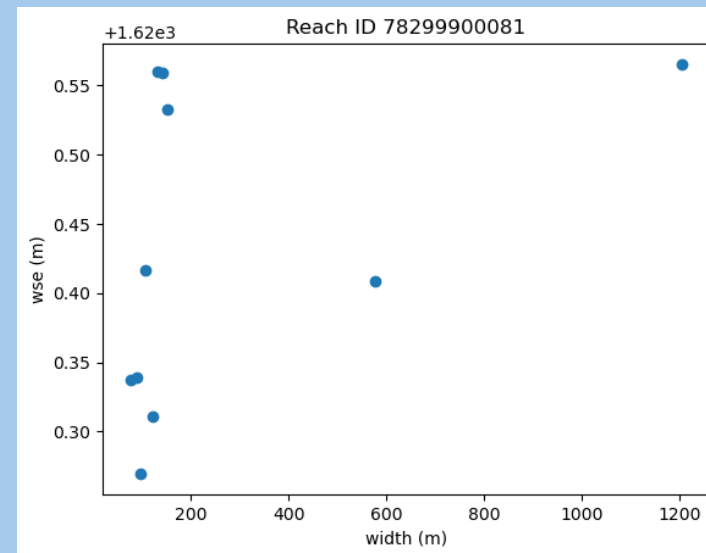
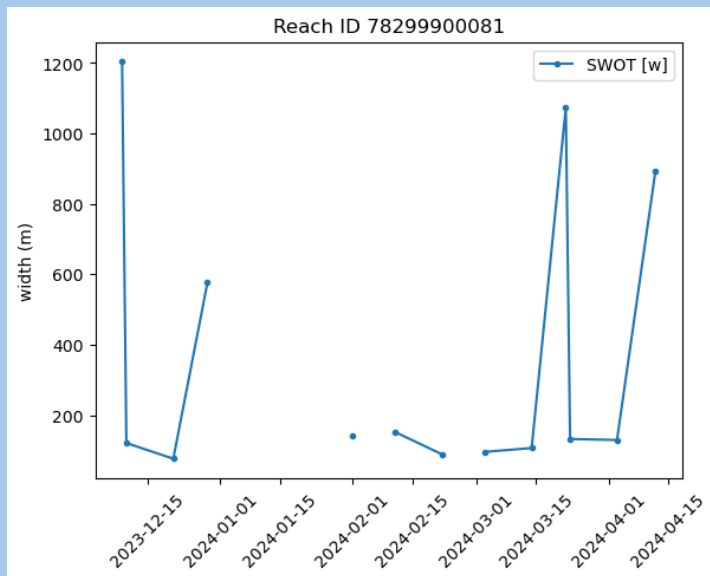
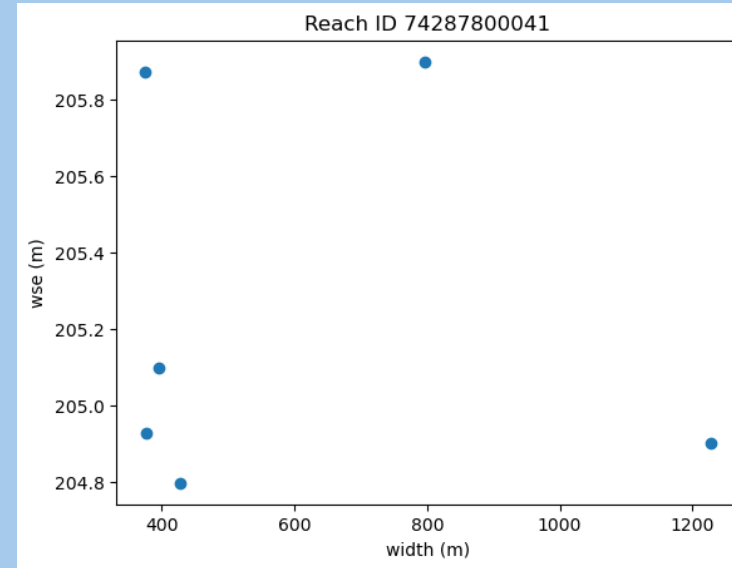
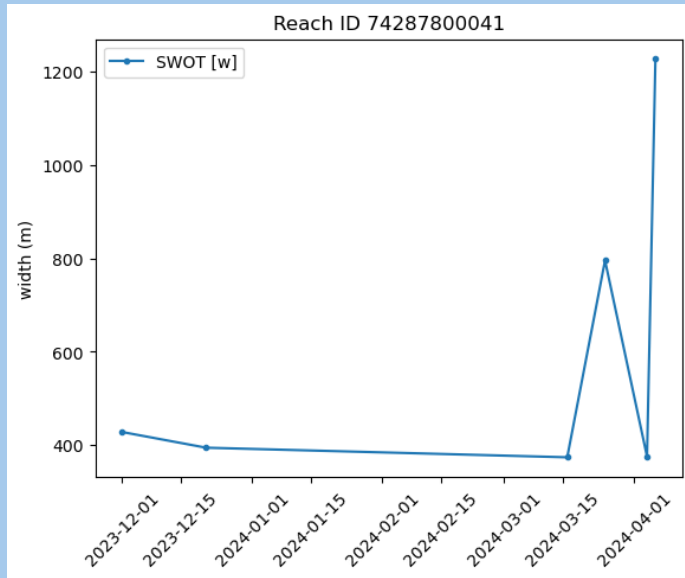
No obvious spatial patterns in σ_ε





River characteristics (from SWORD) do not explain the error

Width Time Series with unrealistic width variation

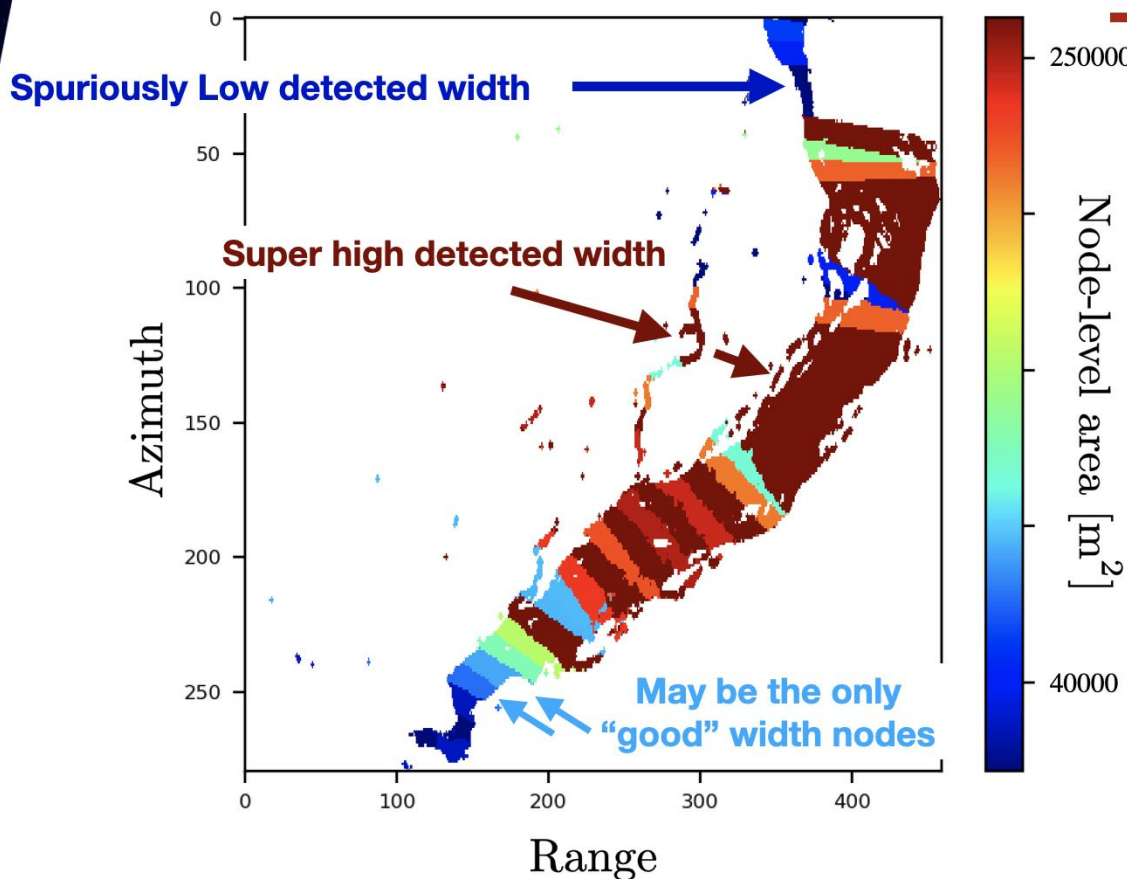




High Width Variability: an “all-in-one” Waimak case study

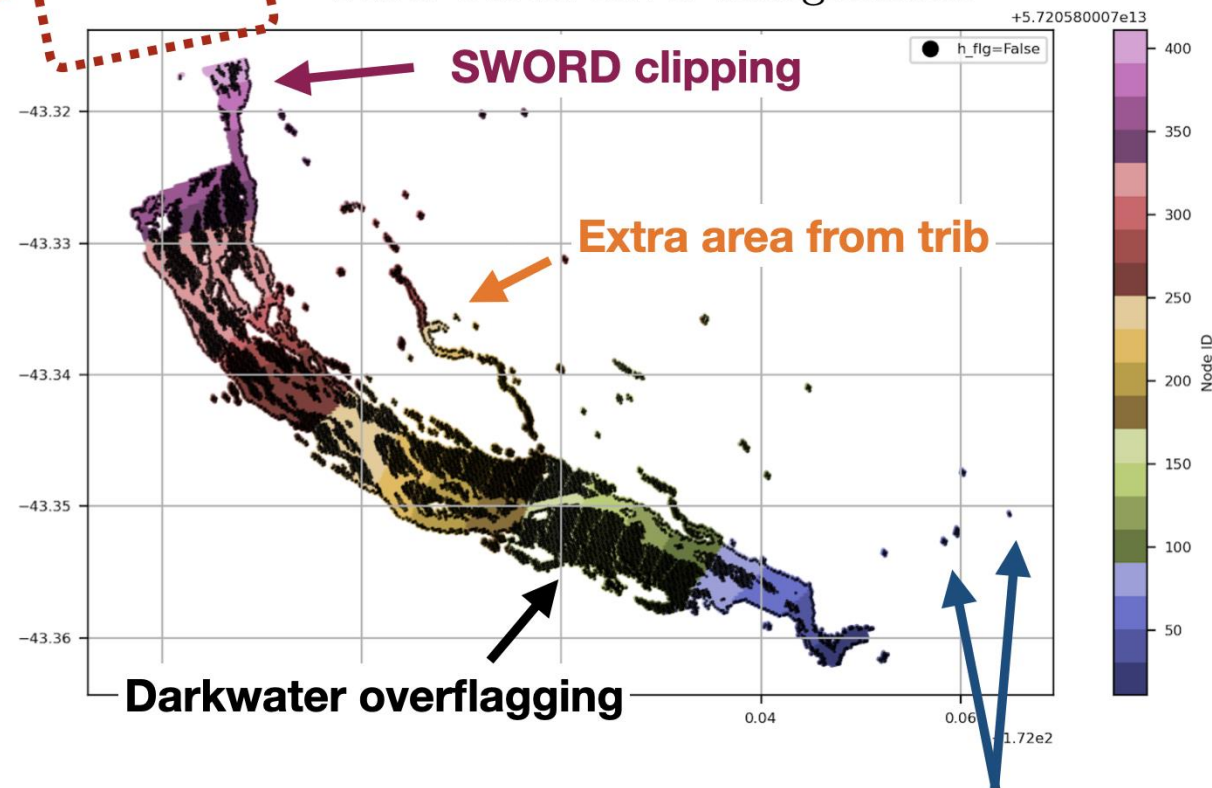
High ($O[10]$) node-to-node variability in area (& width) estimates due to various issues

Area assigned to each node (slant plane)



area=0 edge node
(not visualized here)

Pixel Cloud River Assignments



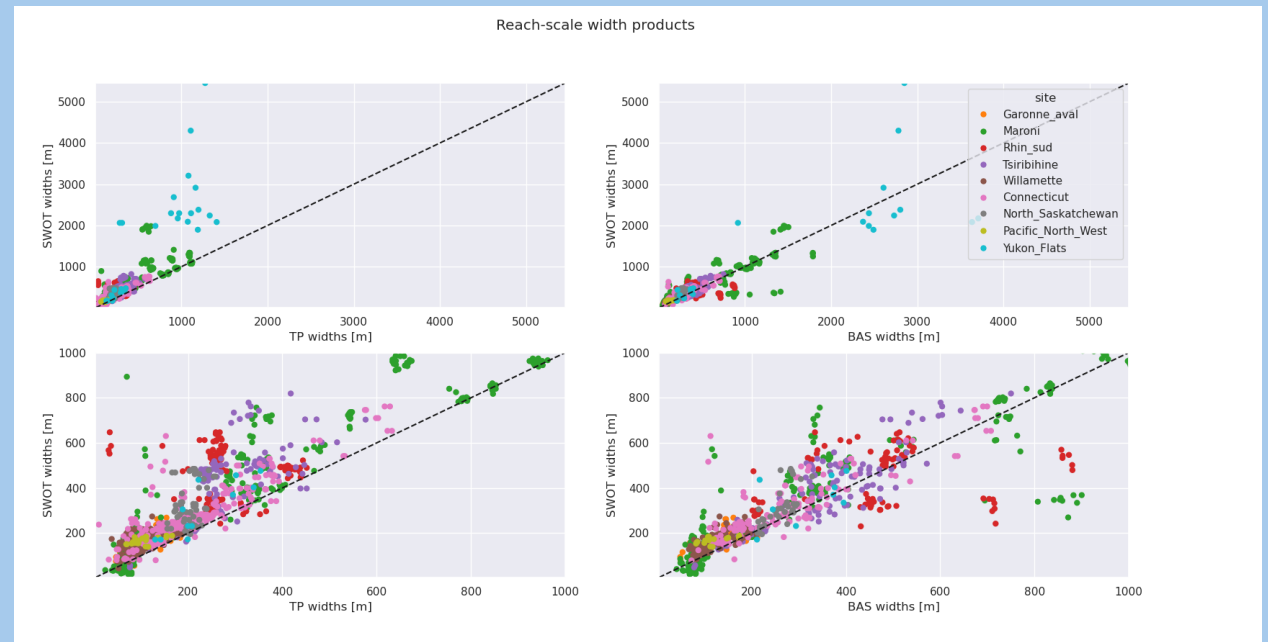
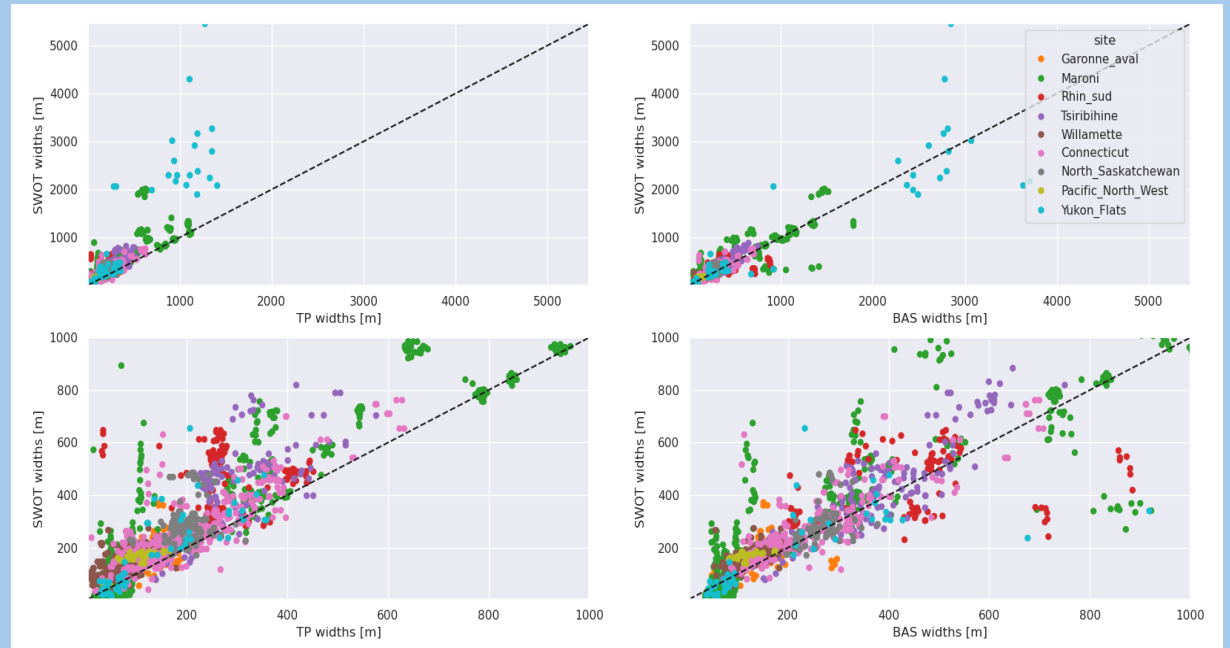
Non-river pixels
mapped to node

Courtesy of Charlotte Emery

– CS GROUP - France

From reference water mask (sentinel 1&2 Pleiades and Radarsat images), 2 river width algorithms are used :

- TP : RiverObs adapted to classic water mask
- BAS : Traditional method to compute width from water mask
- => SWOT currently overestimates river widths



New run experiment using set width

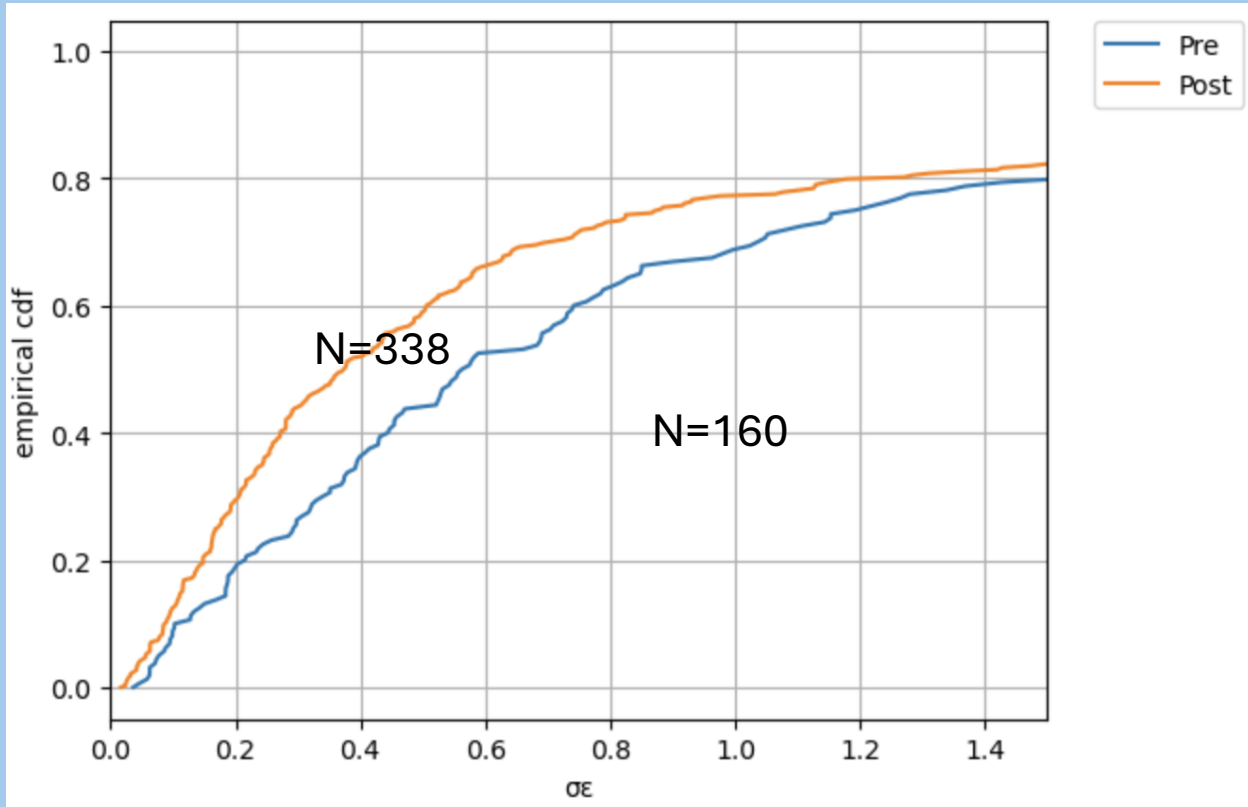
The Hypotheses:

For Momma,
nominal dynamic widths should improve
discharge estimates.

Therefore, if static width improves Q accuracy, the
widths are a major source of error.

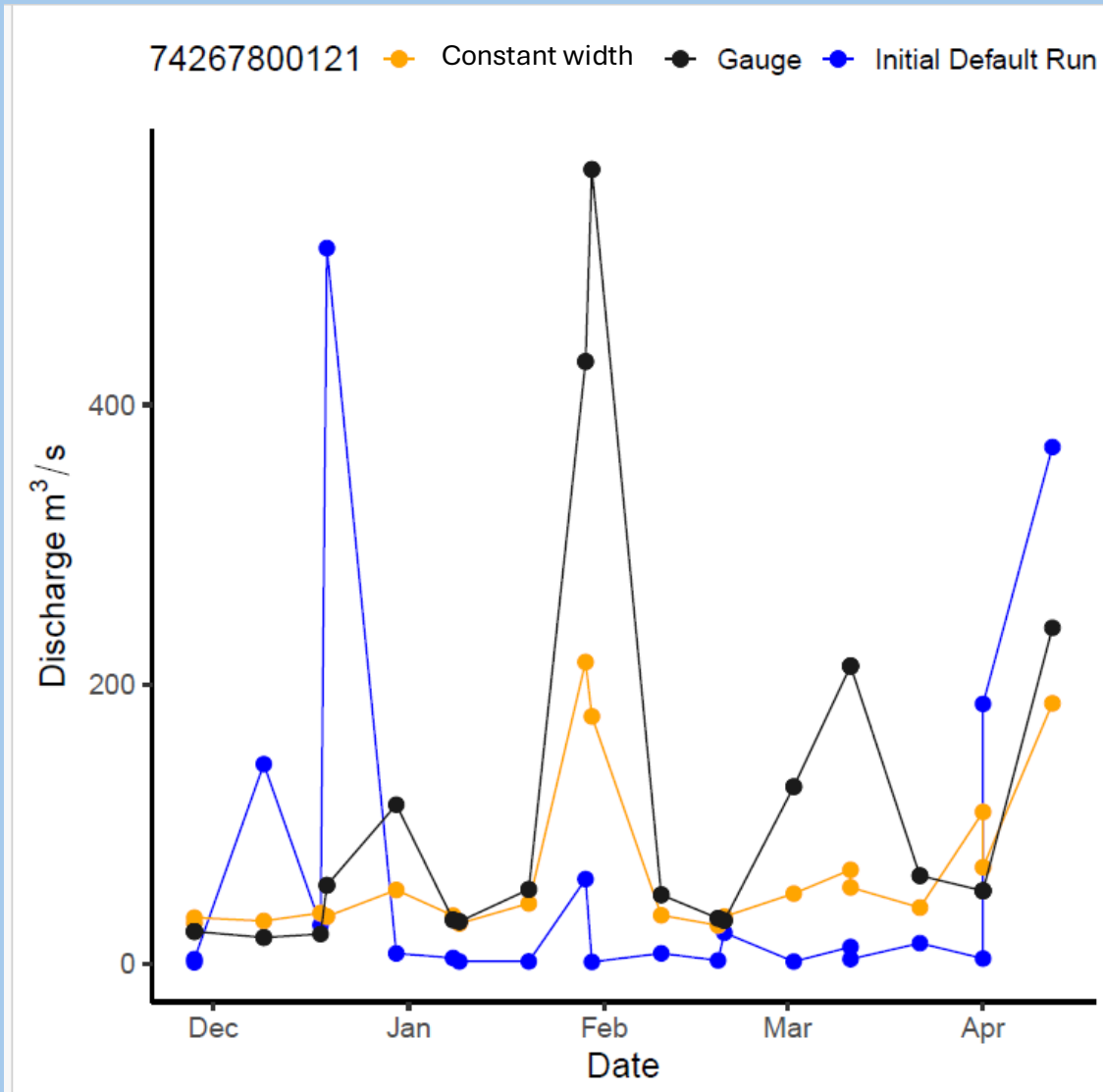


Momma got better!



- **Momma ran on many more reaches (More than doubled)**
- **Errors in dynamics were reduced**
- **Momma devs agree these results can be interpreted this way**

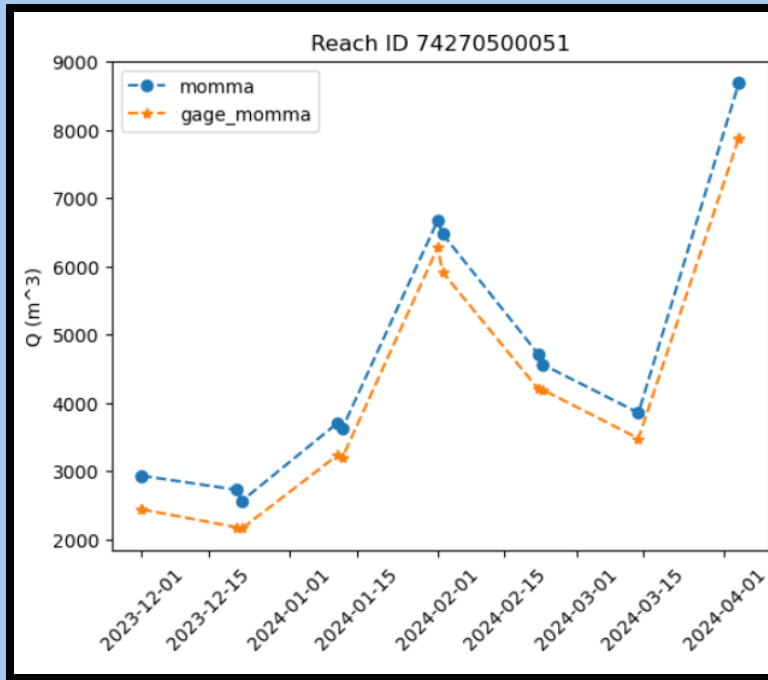
How do we interpret the result



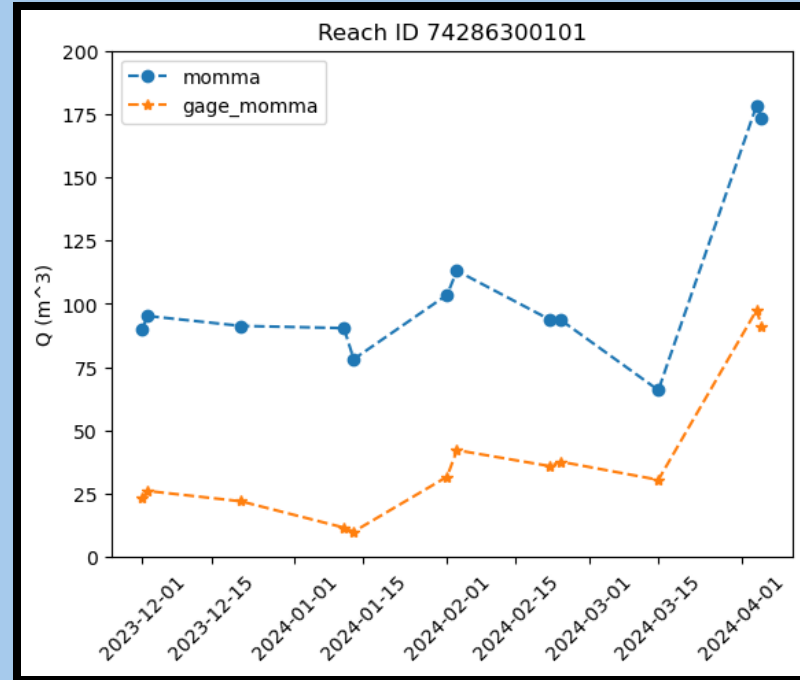
- Here we see the removal unrealistic width changes
- Muted but well-matched dynamics are present as expected
- Width is not a silver bullet, but this experiment show us that we must address width before we can move on to other issues

Courtesy of Merritt Harlin

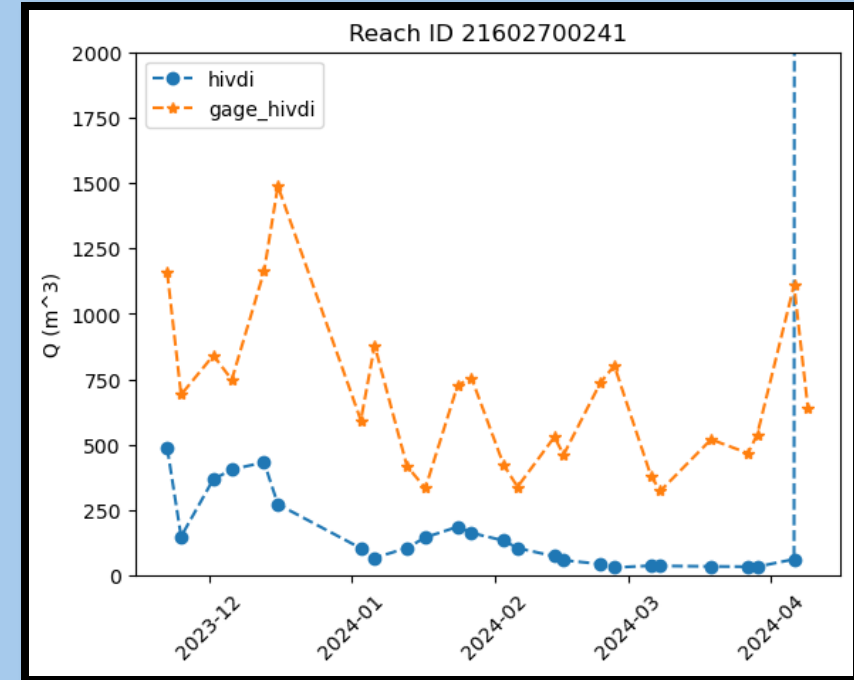
How much data did Cassie's recommended filter throw out?



4.8%



87%



95%

There is a tradeoff between data quality and data quantity!

Summary

- We successfully ran on our largest reach set yet
- Despite overall validation results being below expectations, there are still many examples of successfully estimating discharge
- Confluence is working when inputs are good
- We have a few lines of evidence that point to width data quality driving poor discharge results and this is a solvable problem
- As the dataset grows confluence will have improved capacity to manage noise in the data



Thanks so much to everyone who contributed !