

Global Data Assimilation of Virtual SWOT data in CaMa-Flood

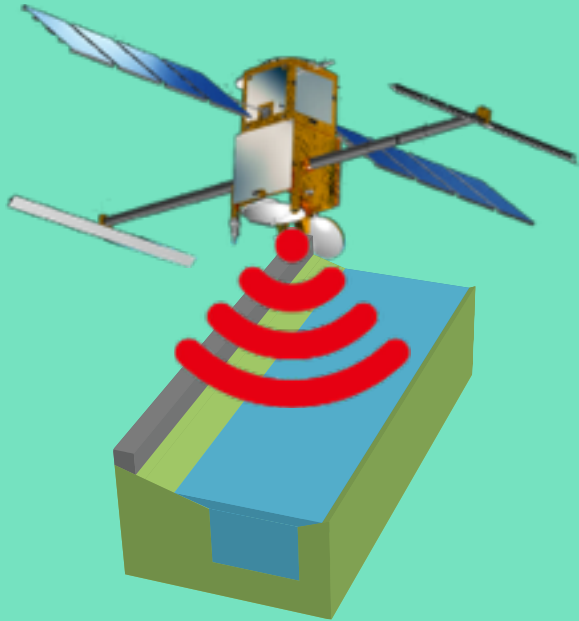
Tokyo Institute of Technology
Daiki Ikeshima

Apr. 17th, 2017 SWOT Teleconference

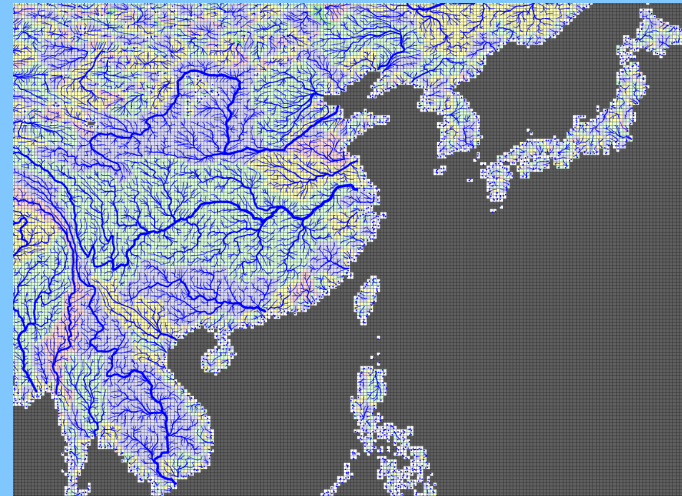
1. SWOT DA Overview
2. Method for DA
3. Results and Discussions
4. Future Steps

1. SWOT DA Overview

SWOT



Water Surface Elevation



**River Model
CaMa-Flood**

**Model Forecasted Discharge
etc...**

Data Assimilation



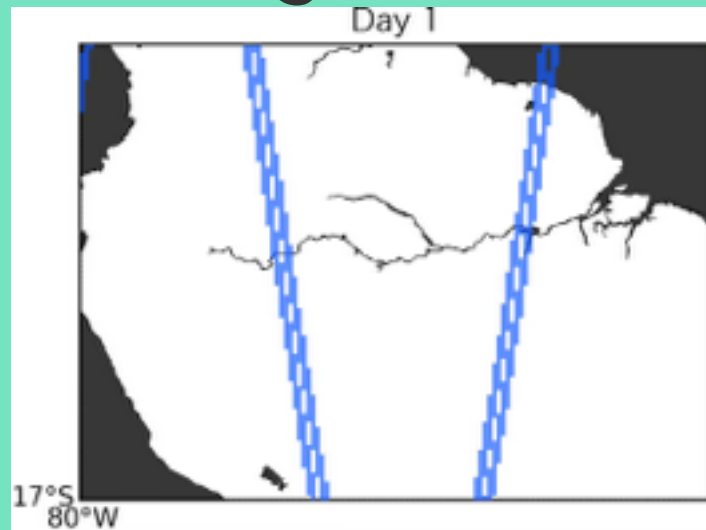
**River
Discharge**

Get Better Estimation by merging "data" into "simulation"

1. SWOT DA Overview

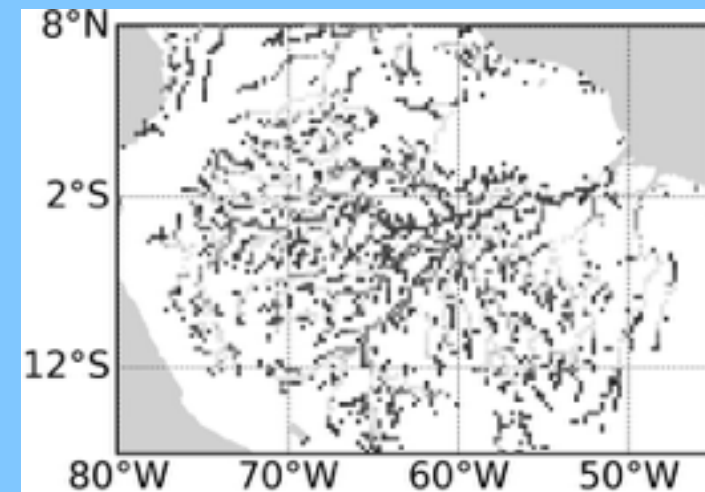
SWOT

Daily SWOT Observation
Coverage is Limited

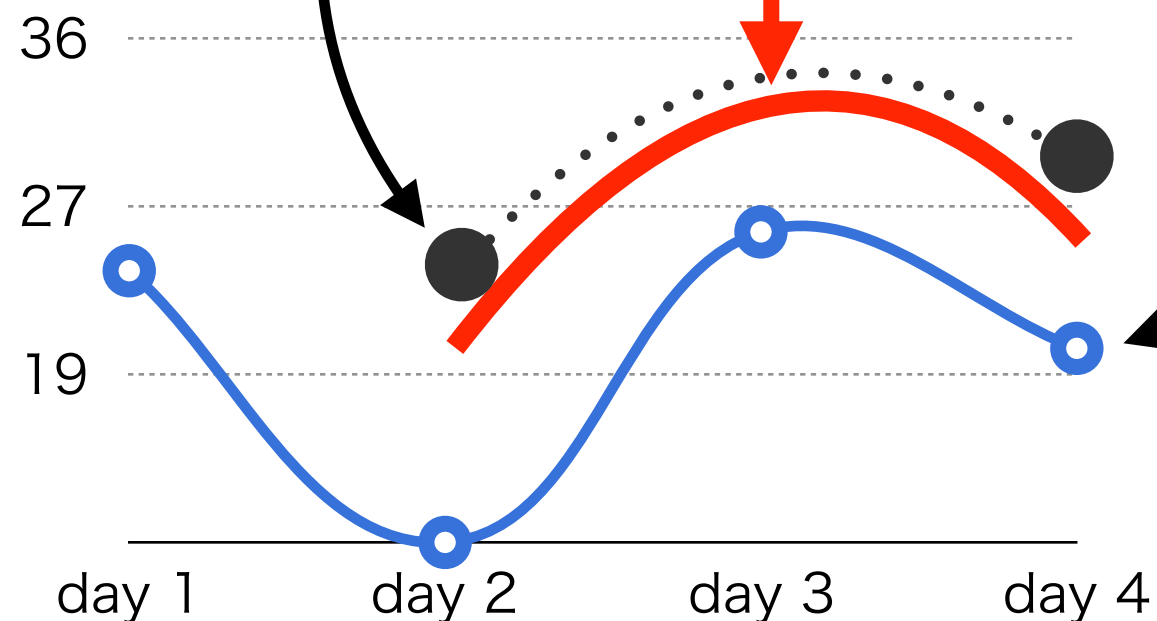


River Model CaMa-Flood

Simulation has
Full Area Forecast



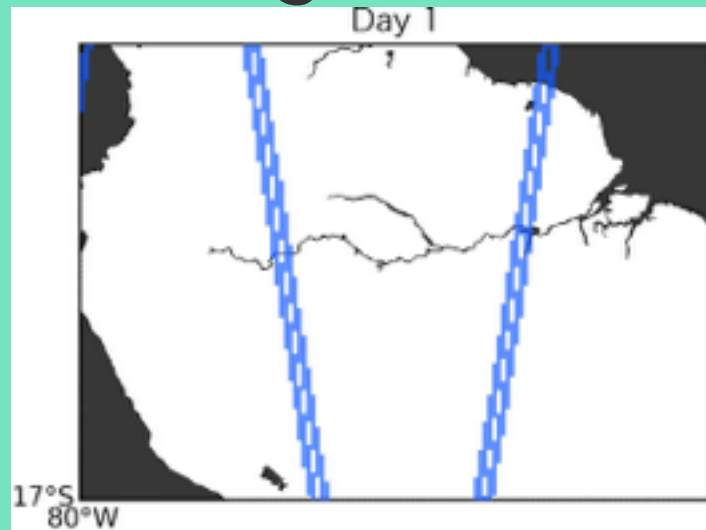
Data Assimilation



1. SWOT DA Overview

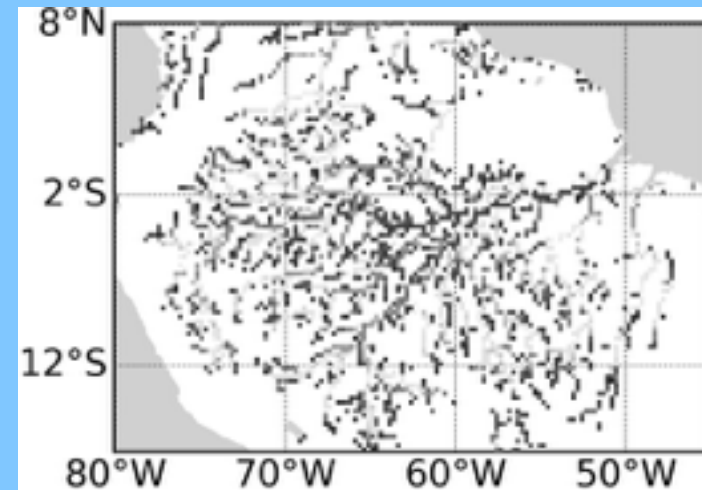
SWOT

Daily SWOT Observation
Coverage is Limited



River Model CaMa-Flood

Simulation has
Full Area Forecast

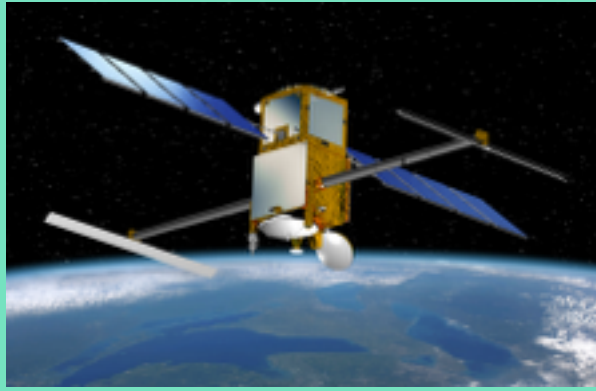


Objective

- ✓ To Develop a Global Framework of SWOT Data Assimilation
- ✓ Evaluated the Effectiveness Before Launch

2. Method of Data Assimilation

SWOT



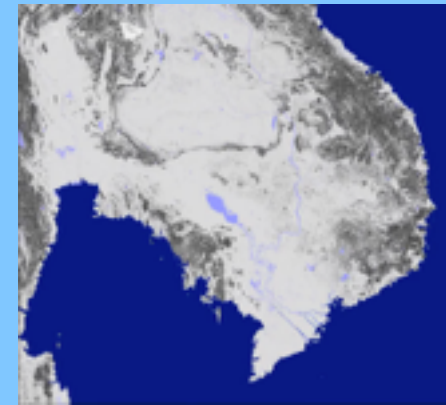
Not launched until 2021



Used Virtual SWOT

(virtual observation data made
from river model)

River Model CaMa-Flood

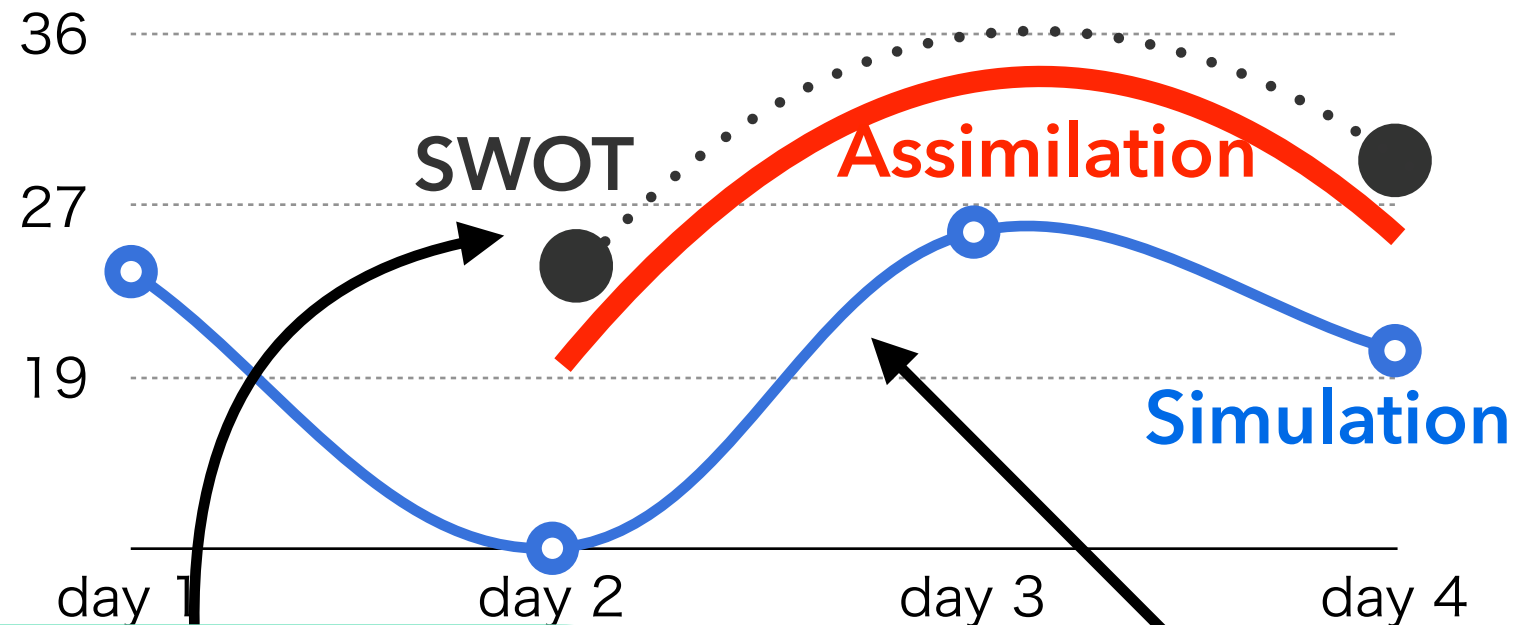


and did Virtual Experiment

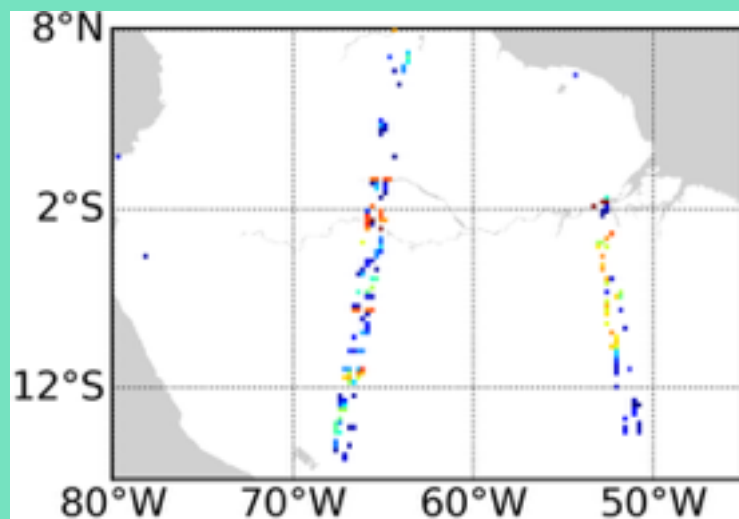
2. Method of Data Assimilation

Virtual Experiment

Virtual SWOT and Simulation must be apart

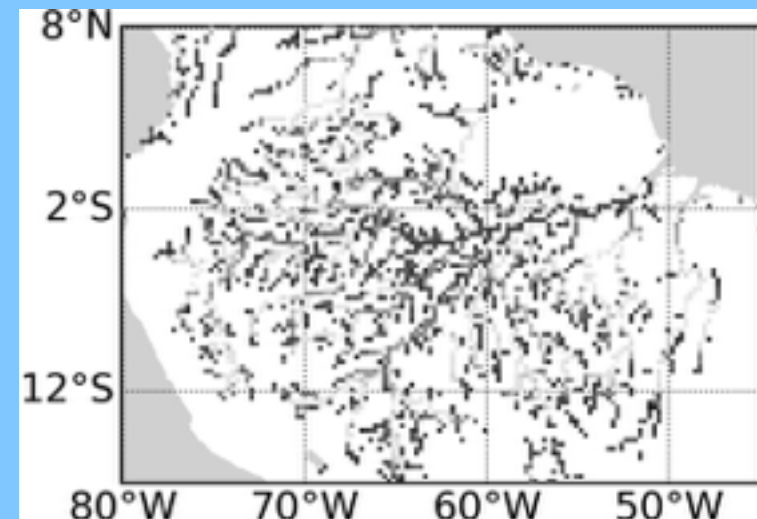


Virtual SWOT



Virtual Observation Data
made from River Model

Simulation



Artificially and Purposely Error
Corrupted Simulation

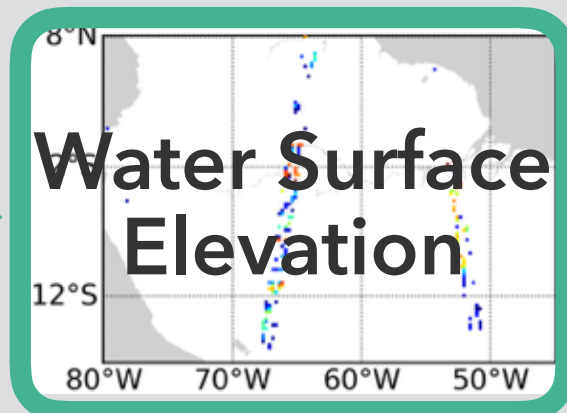
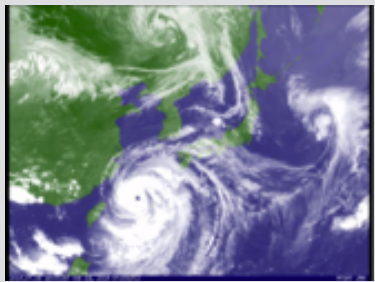
2. Method of Data Assimilation

Virtual Experiment

we did 2 Different Patterns (Experiment) for making them apart

(A) -25% Experiment

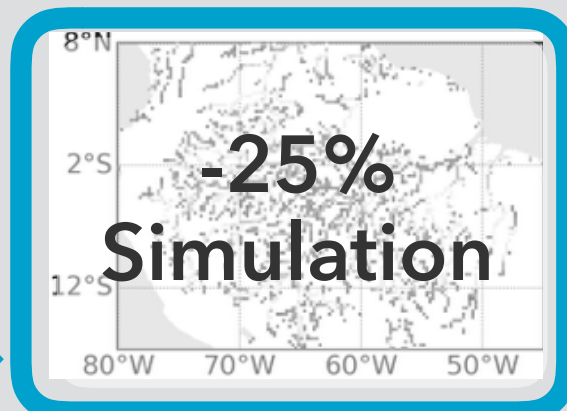
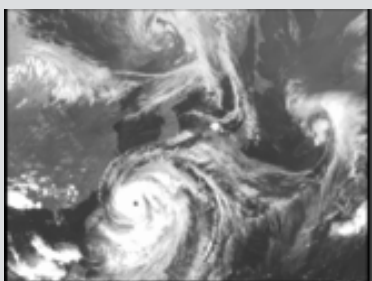
Runoff (Input)



Virtual SWOT

Reduce 25%

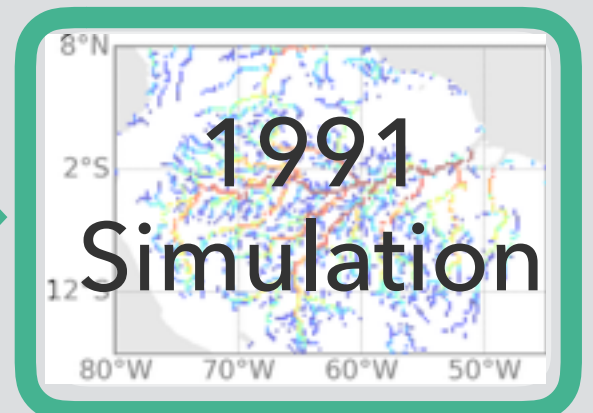
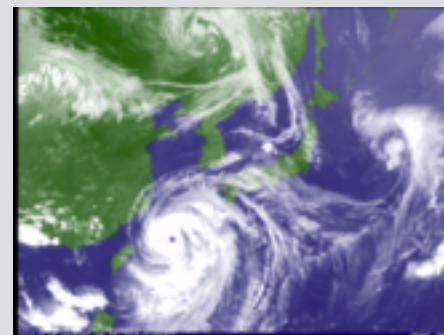
25% Reduced
Runoff (Input)



Corrupted
Simulation

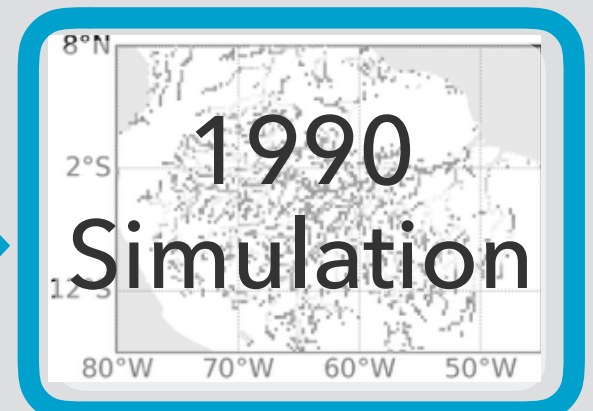
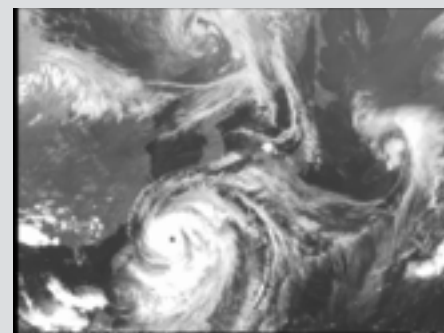
(B) Blind Runoff Experiment

1991 Runoff



Virtual SWOT

1990 Runoff



Corrupted
Simulation

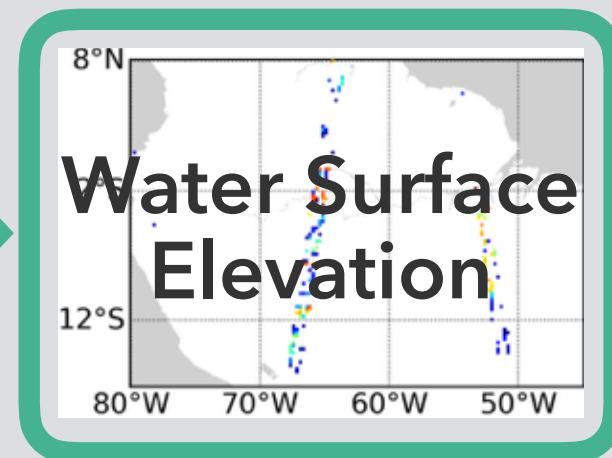
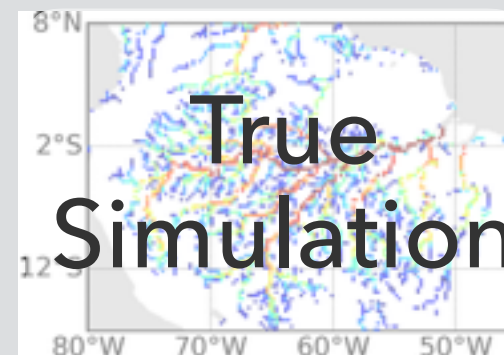
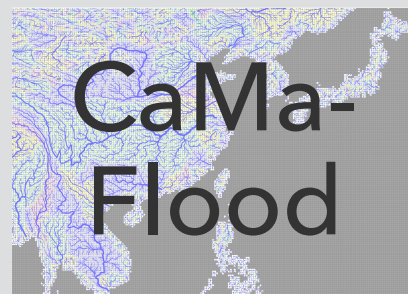
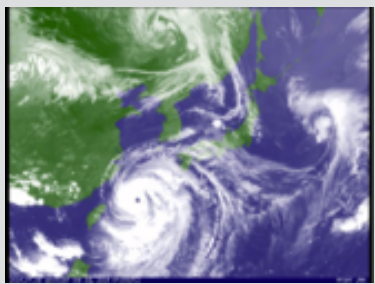
2. Method of Data Assimilation

Virtual Experiment

we did 2 Different Patterns (Experiment) for making them apart

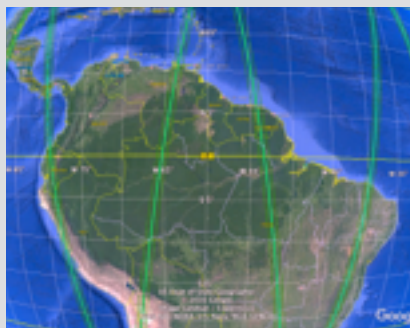
(A) -25% Experiment (Andreadis et al., 2007)

Runoff (Input)

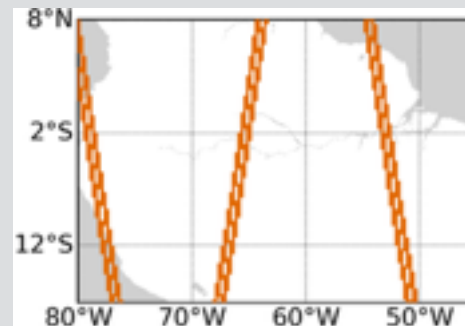


Virtual SWOT

Planned
SWOT Orbit Data



SWOT
Coverage Mask

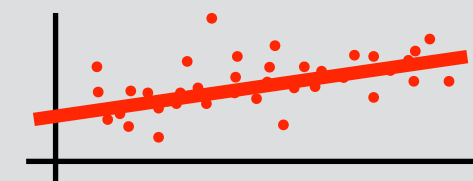


mask out



add

Measurement Error



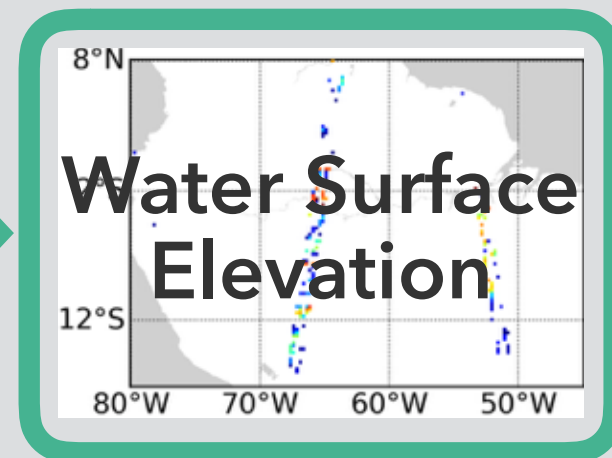
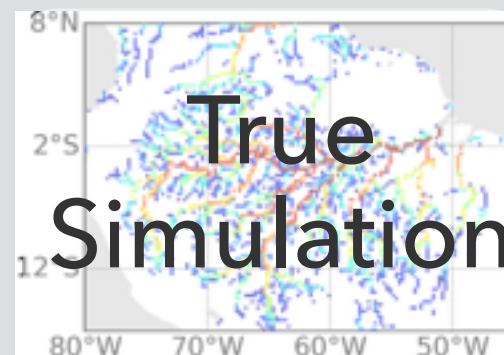
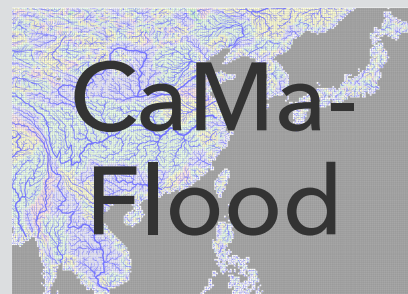
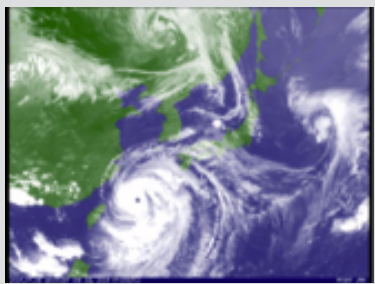
2. Method of Data Assimilation

Virtual Experiment

we did 2 Different Patterns (Experiment) for making them apart

(A) -25% Experiment (Andreadis et al., 2007)

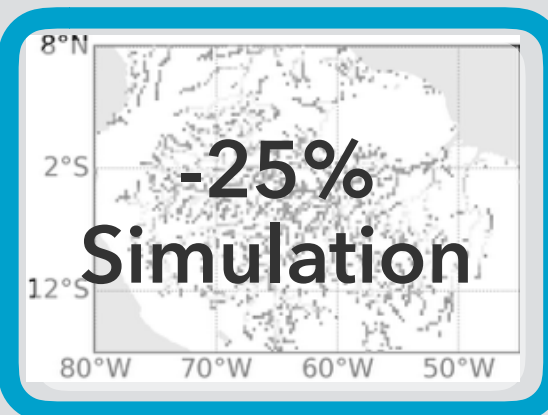
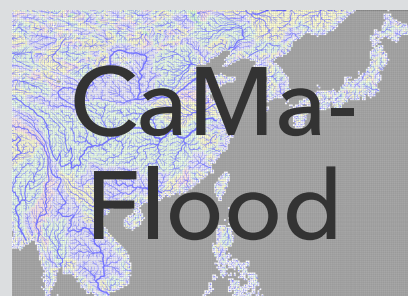
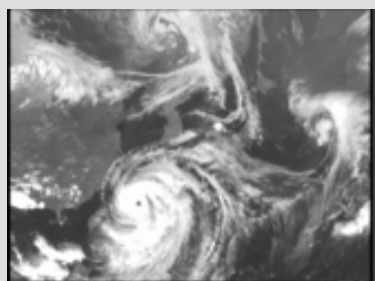
Runoff (Input)



Virtual SWOT

Reduce 25%

25% Reduced
Runoff (Input)



Corrupted
Simulation

← Roughly
25% Smaller

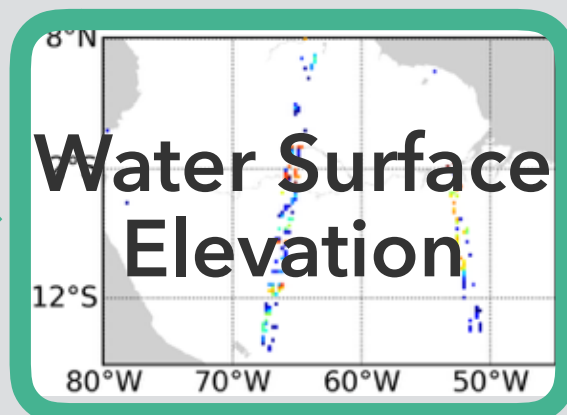
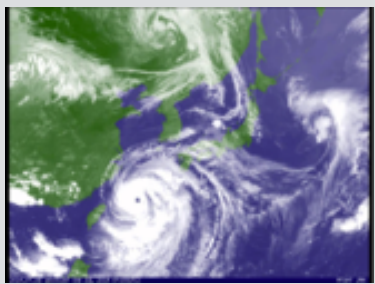
2. Method of Data Assimilation

Virtual Experiment

we did 2 Different Patterns (Experiment) for making them apart

(A) -25% Experiment

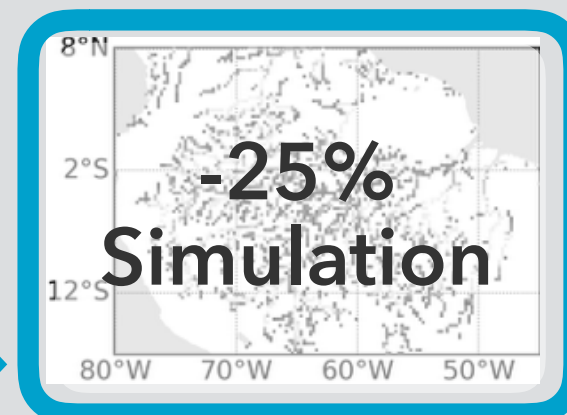
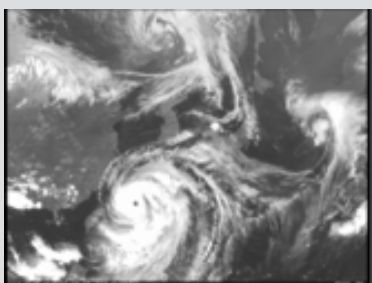
Runoff (Input)



Virtual SWOT

Reduce 25%

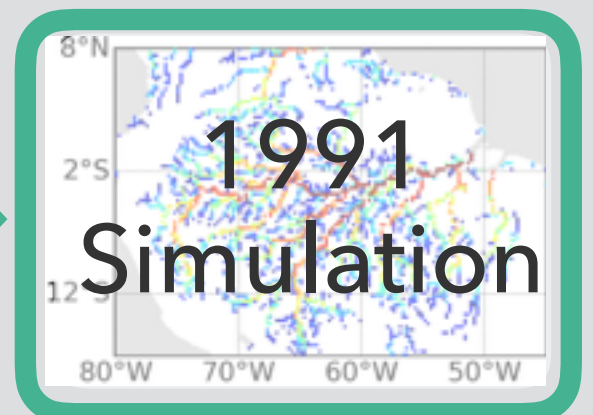
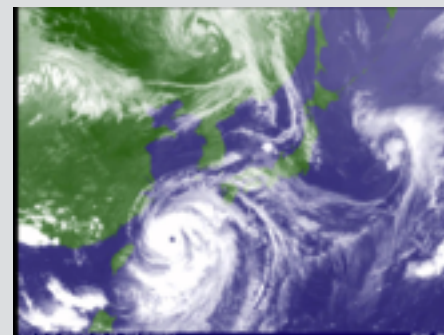
25% Reduced
Runoff (Input)



Corrupted
Simulation

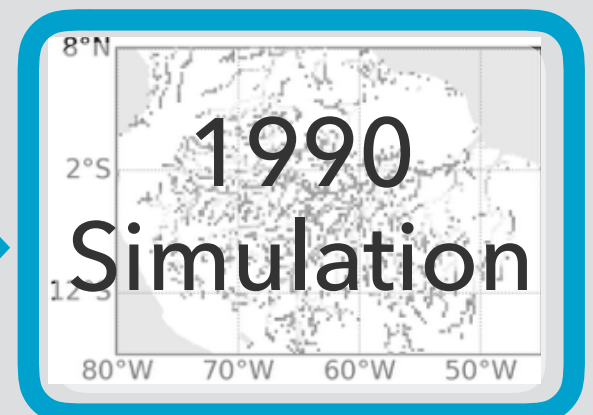
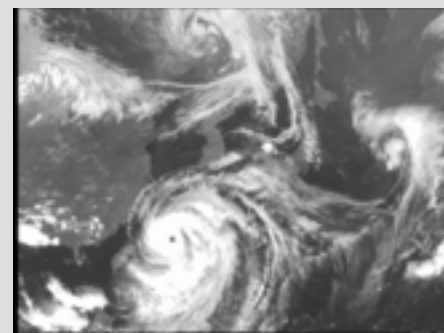
(B) Blind Runoff Experiment

1991 Runoff



Virtual SWOT

1990 Runoff

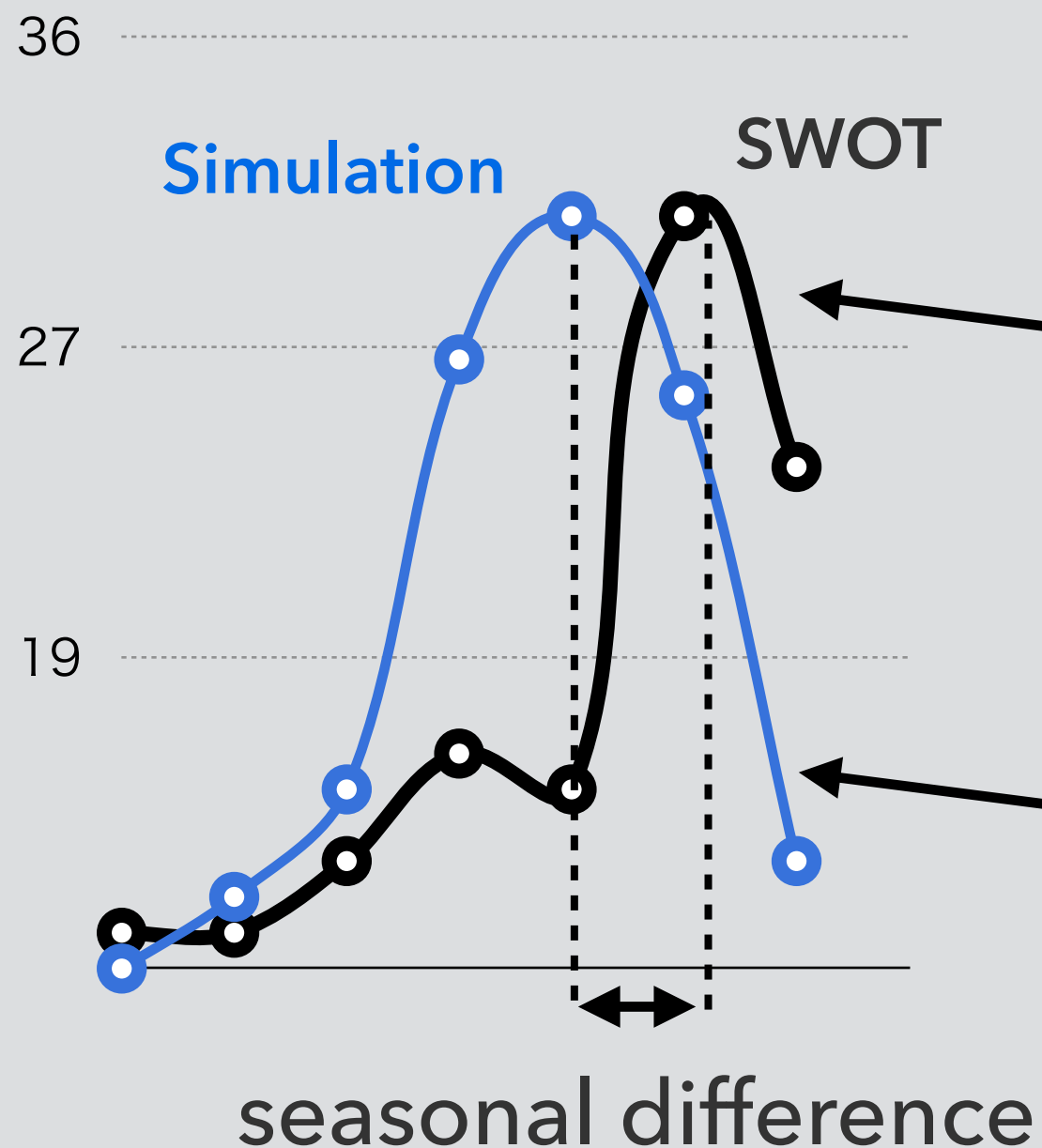


Corrupted
Simulation

2. Method of Data Assimilation

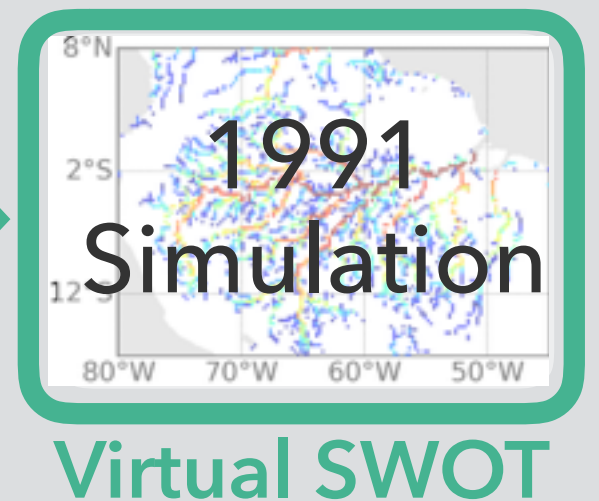
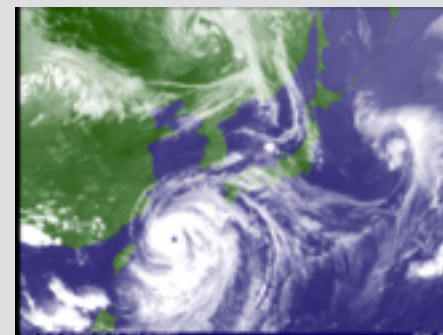
Virtual Experiment

we did 2 Different Patterns (Experiment) for making them apart

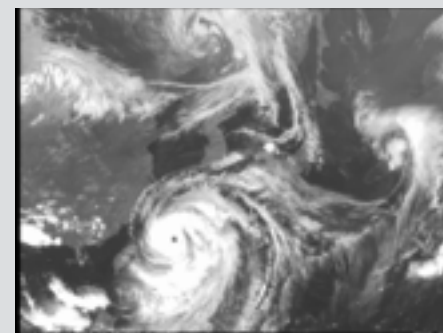


(B) Blind Runoff Experiment

1991 Runoff



1990 Runoff



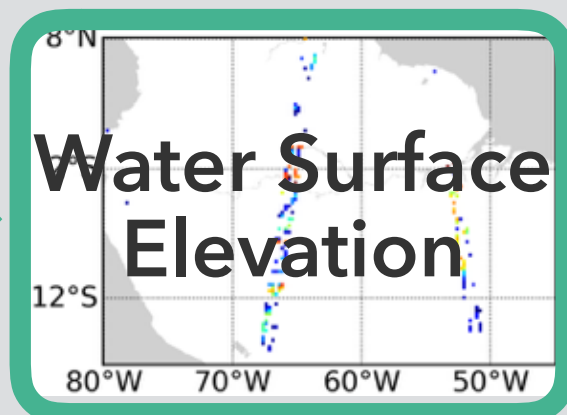
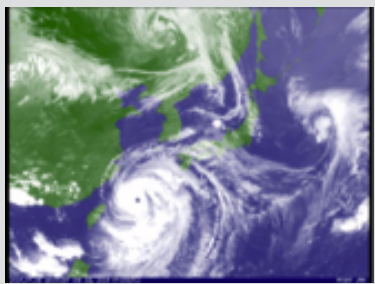
2. Method of Data Assimilation

Virtual Experiment

we did 2 Different Patterns (Experiment) for making them apart

(A) -25% Experiment

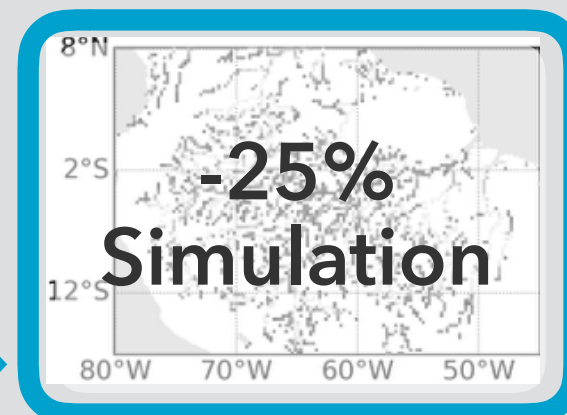
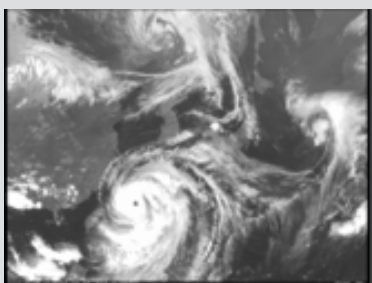
Runoff (Input)



Virtual SWOT

Reduce 25%

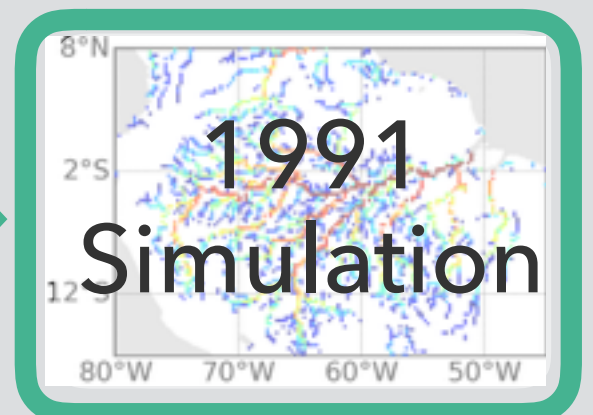
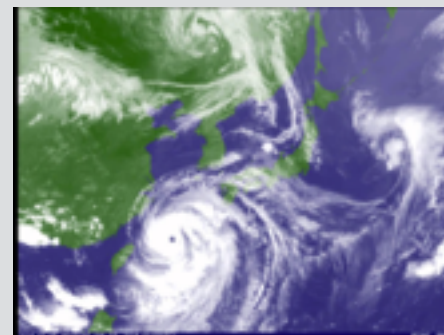
25% Reduced
Runoff (Input)



Corrupted
Simulation

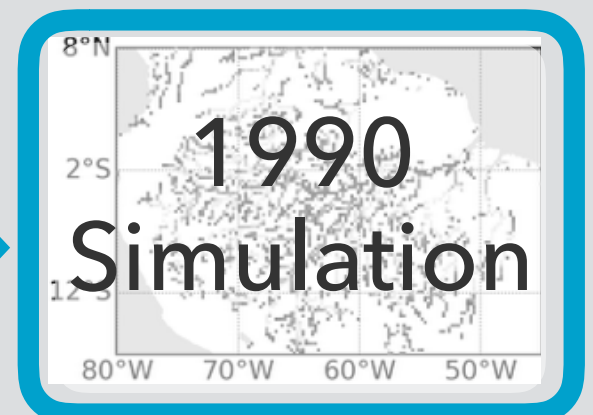
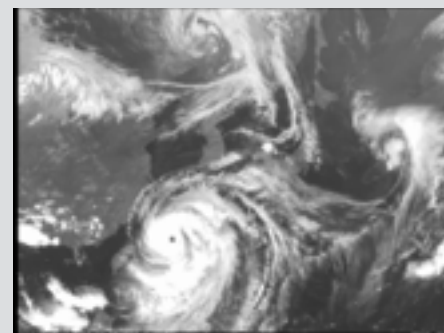
(B) Blind Runoff Experiment

1991 Runoff



Virtual SWOT

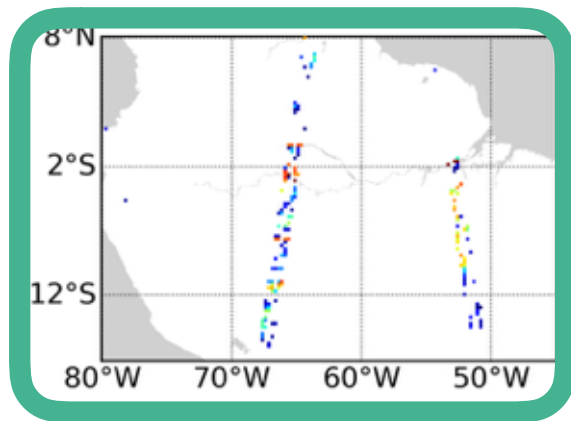
1990 Runoff



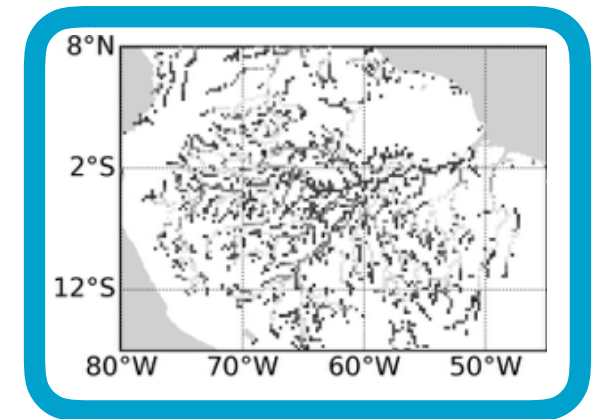
Corrupted
Simulation

2. Method of Data Assimilation

Virtual Experiment



Virtual SWOT

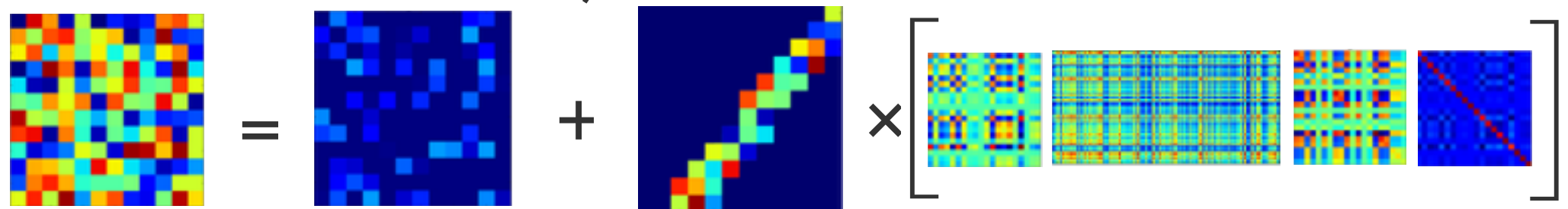


Corrupted Simulation

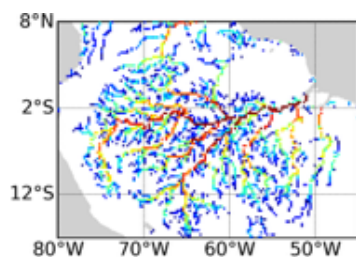
Data Assimilation

Ensemble Kalman Filter (EnKF)

Implemented with LETKF (Local Ensemble Transformation Filter)

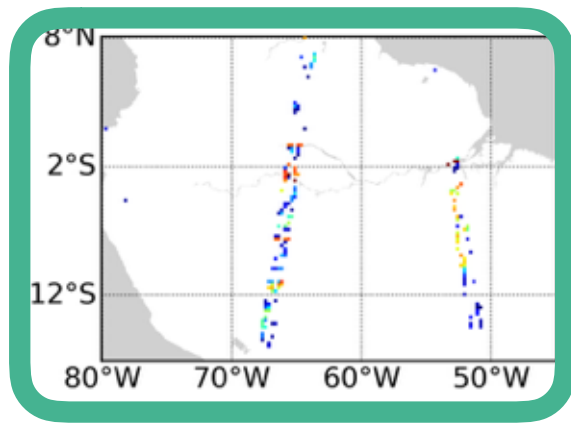
$$\begin{bmatrix} \text{Assimilation} \end{bmatrix} = \begin{bmatrix} \text{Simulation} \end{bmatrix} + \begin{bmatrix} \text{Observation} \end{bmatrix} \times \left[\begin{bmatrix} \text{Ensemble Statistics} \end{bmatrix} \right]$$
The diagram illustrates the EnKF equation using heatmaps. On the left is a heatmap for 'Assimilation'. This is followed by an equals sign, then a heatmap for 'Simulation', a plus sign, a heatmap for 'Observation' (showing a diagonal pattern), a multiplication sign, and a bracket containing four heatmaps representing 'Ensemble Statistics'.

Assimilation Simulation Observation Ensemble Statistics



Assimilated Estimation

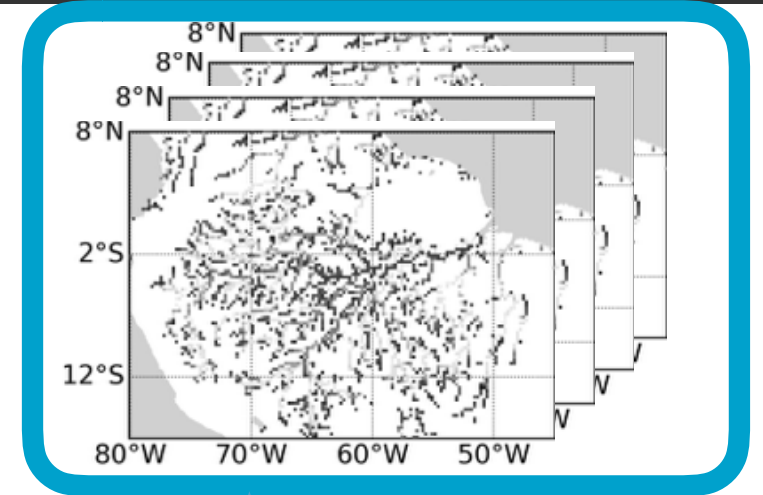
2. Method of Data Assimilation



Virtual SWOT

Virtual Experiment

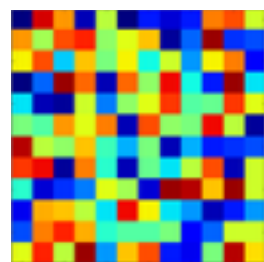
Ensemble
are needed →



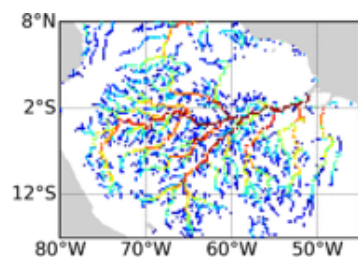
Corrupted
Simulation

Data Assimilation

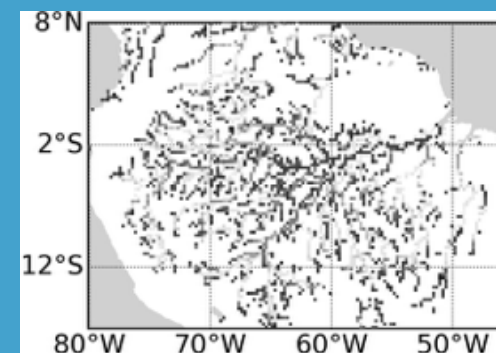
Ensemble
Implement
Transform



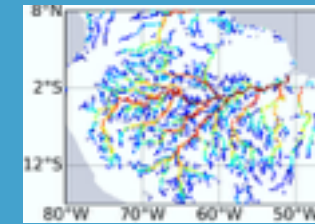
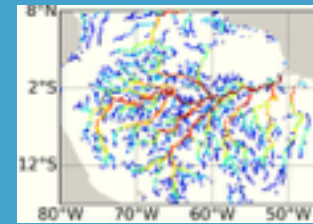
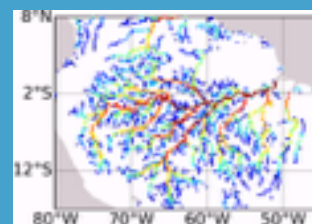
Assimilation



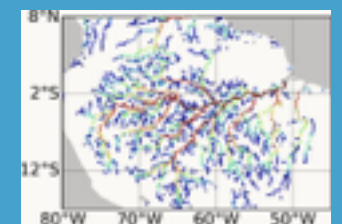
Assimilated



Corrupted
Simulation

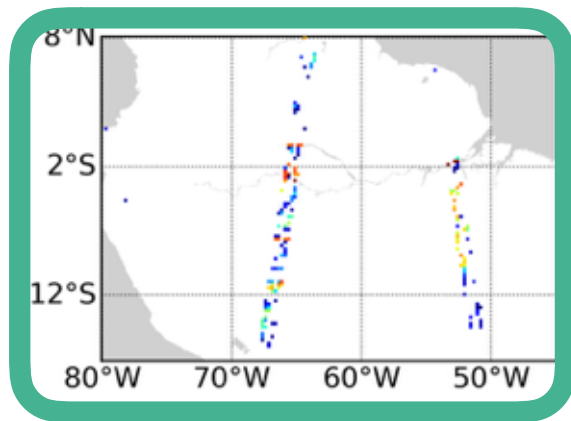


...



Made 20 Ensemble Members (using Gaussian Noise)

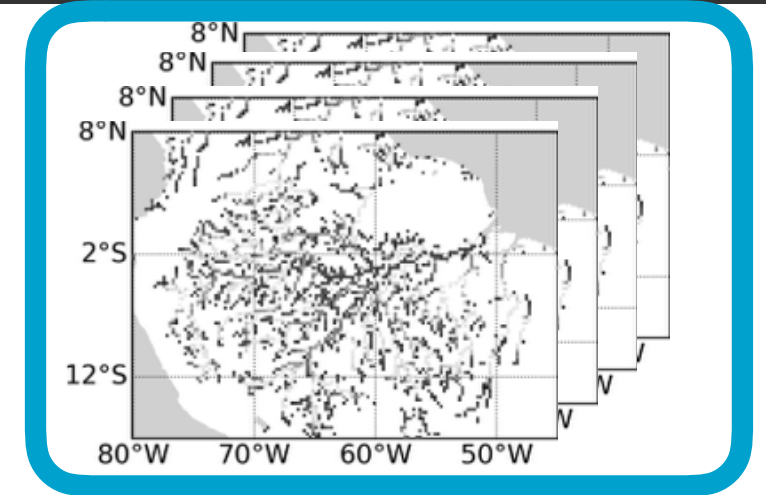
2. Method of Data Assimilation



Virtual SWOT

Virtual Experiment

Ensemble
are needed →



Corrupted
Simulation

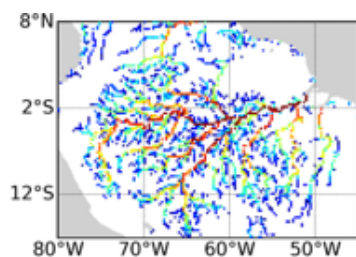
Data Assimilation

Ensemble Kalman Filter (EnKF)

Implemented with LETKF (Local Ensemble
Transformation Filter)

$$\begin{bmatrix} \text{Assimilation} \end{bmatrix} = \begin{bmatrix} \text{Simulation} \end{bmatrix} + \begin{bmatrix} \text{Observation} \end{bmatrix} \times \left[\begin{bmatrix} \text{Ensemble Statistics} \end{bmatrix} \right]$$

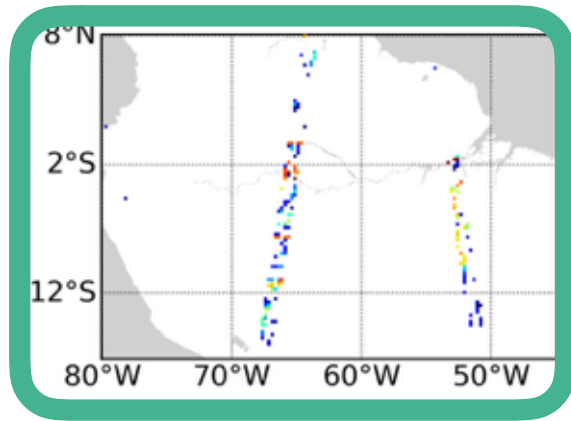
Assimilation Simulation Observation Ensemble Statistics



Assimilated Estimation

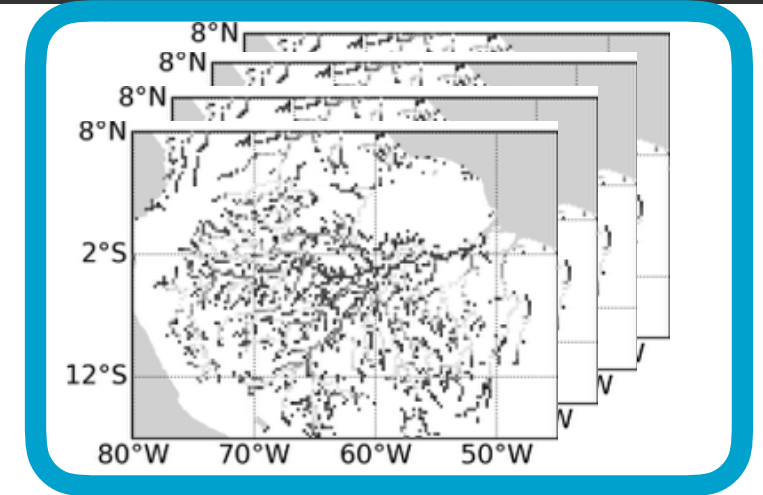
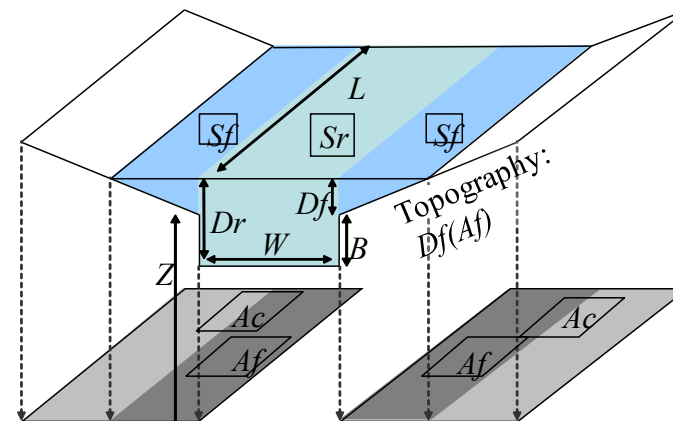
2. Method of Data Assimilation

Virtual Experiment



Virtual SWOT

Water Surface
Elevation (WSE)

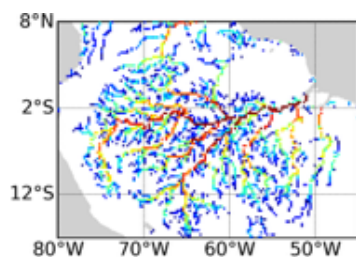


Corrupted
Simulation

Many Variables

This is possible because CaMa-Flood
calculates water dynamics based on WSE

Data Assimilation



Assimilated Estimation

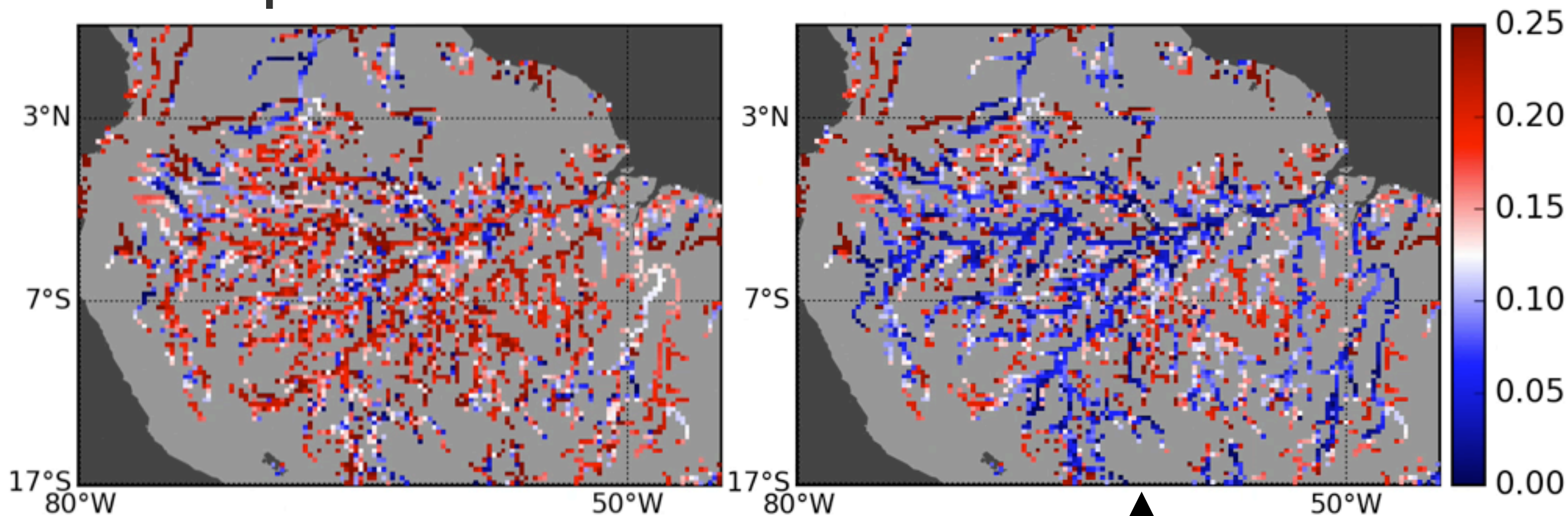
3. Results and Discussions

(A) -25% Experiment

absolute Error Rate of River Discharge

Corrupted Simulation

Assimilation



Error Decreased



the Amazon River

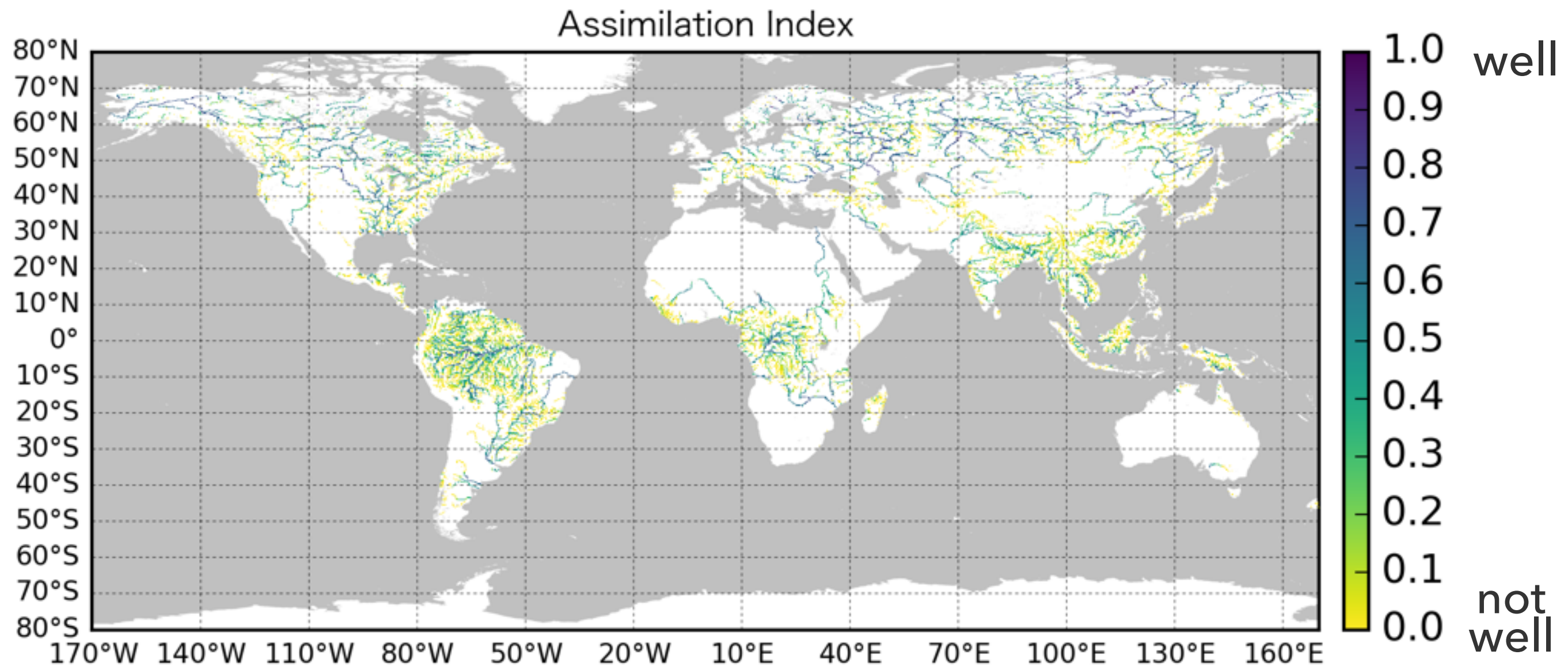
3. Results and Discussions

(B) Blind Runoff

Assimilation Index(AI): relative Assimilation Achievement

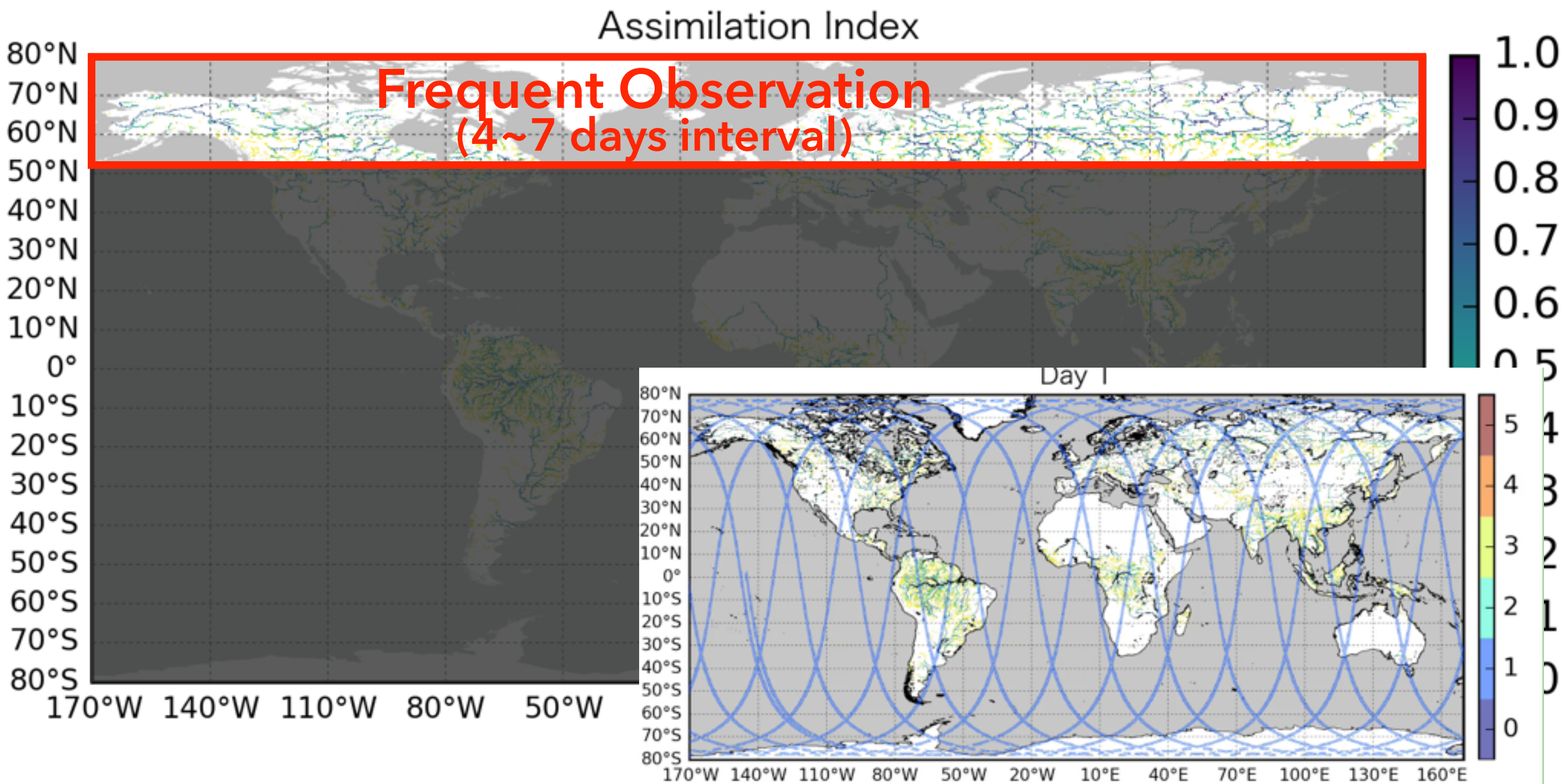
$$AI = 1 - \left| \frac{Assim. - Corrup.}{True - Corrup.} - 1 \right|$$

AI=1: done well
AI=0: not well



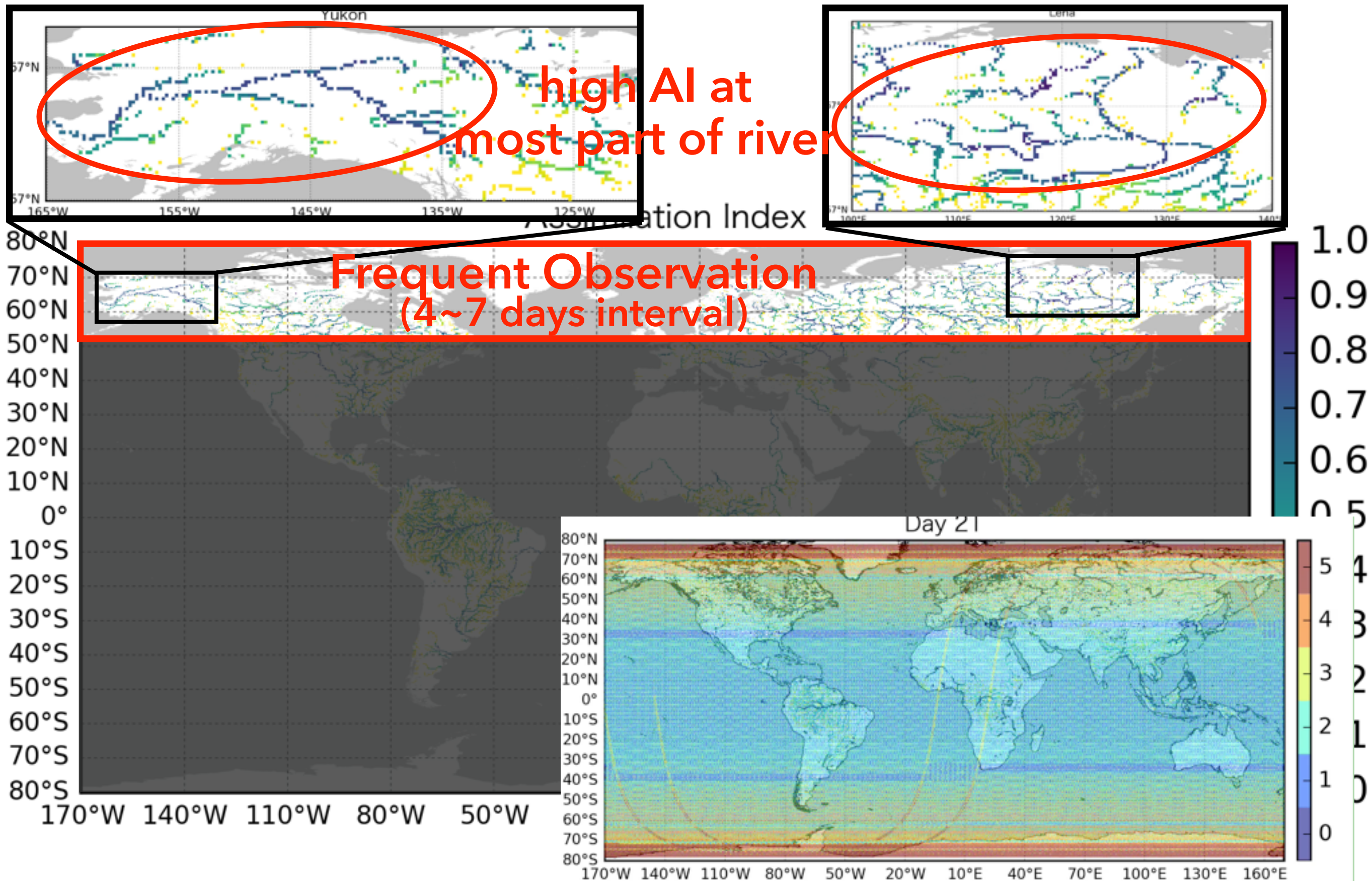
3. Results and Discussions

(B) Blind Runoff

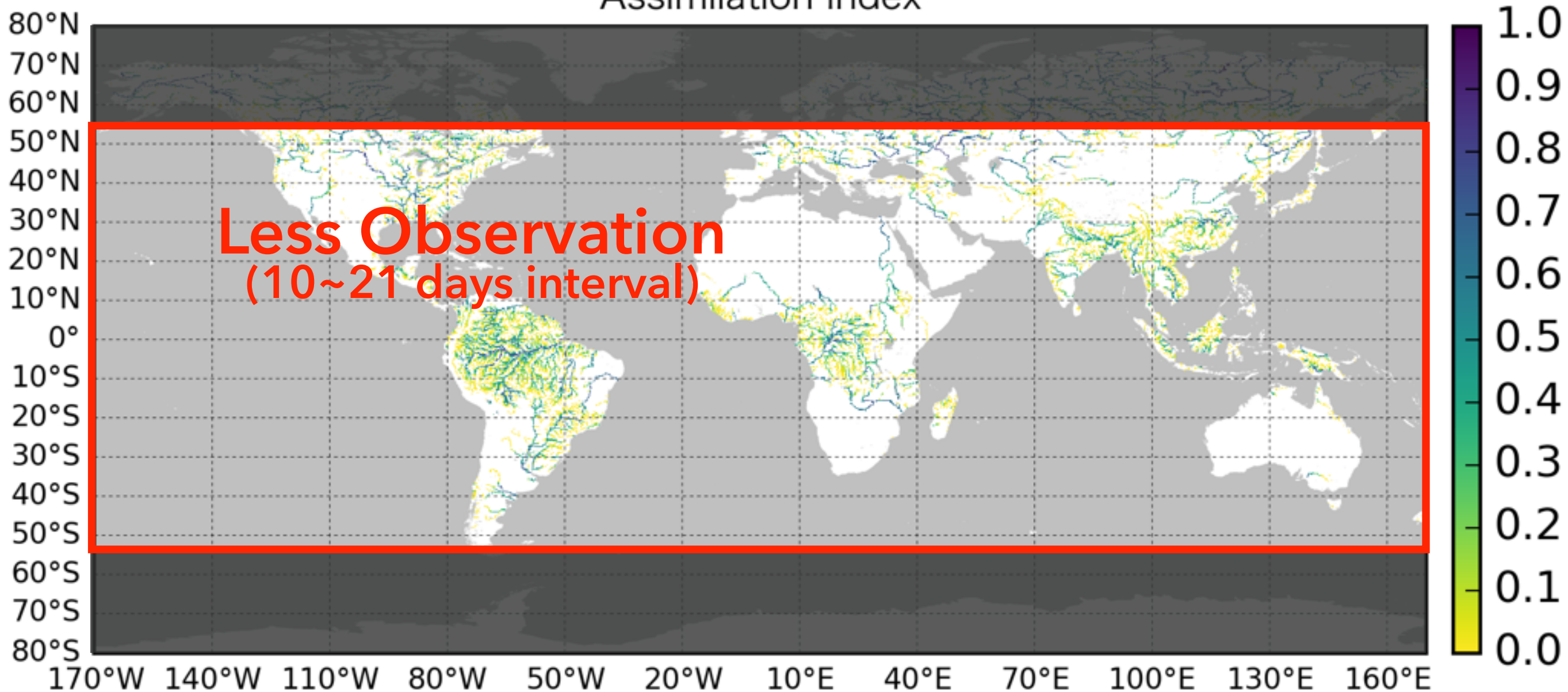
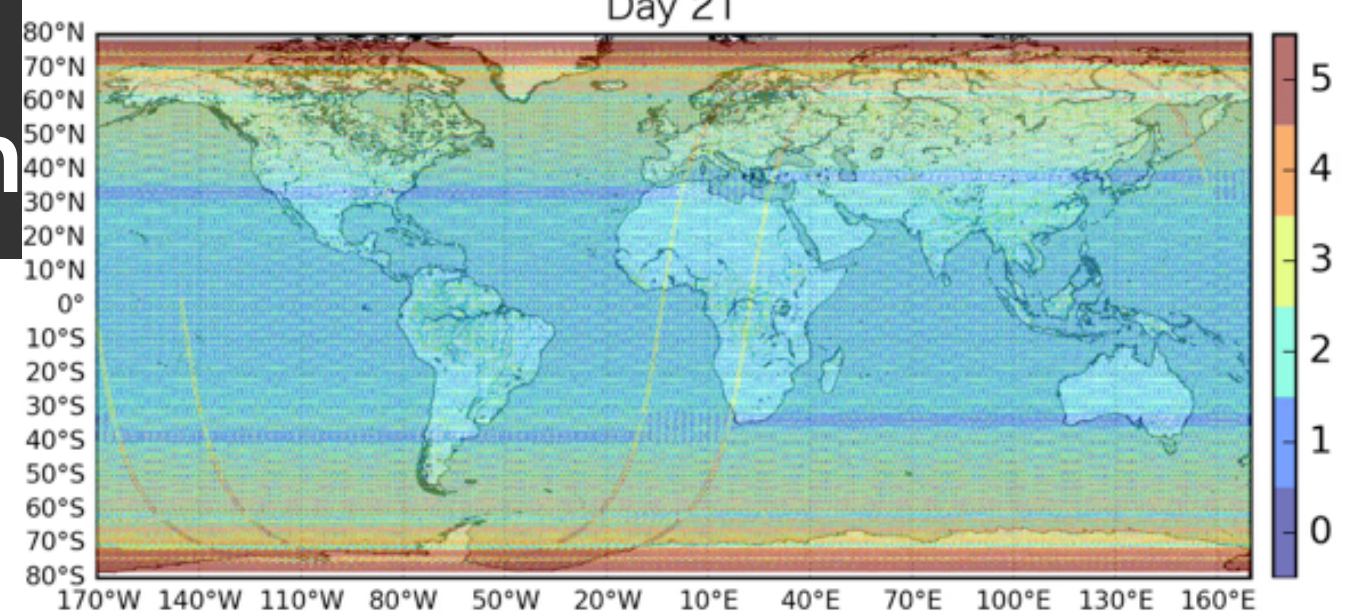


3. Results and Discussions

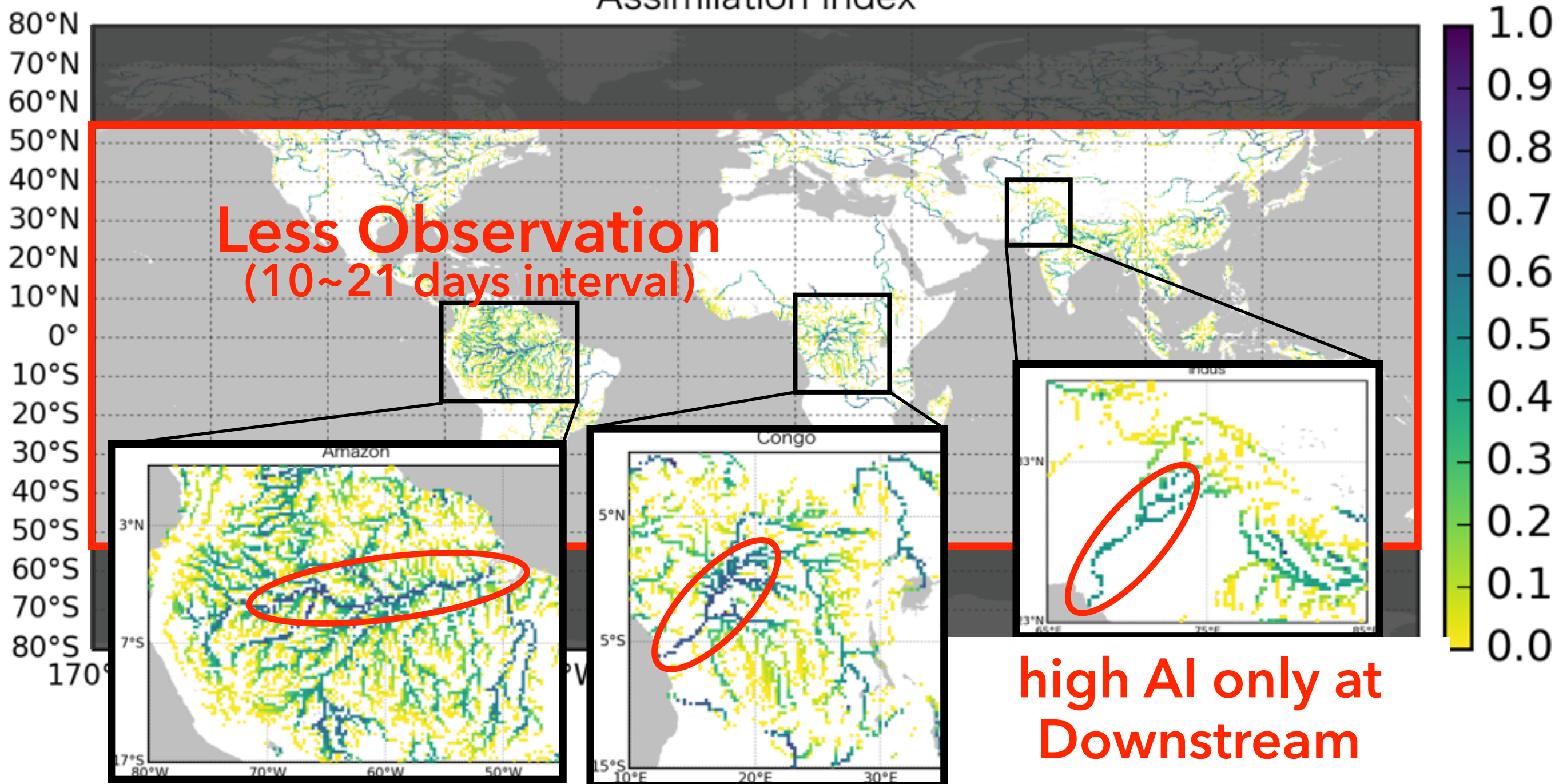
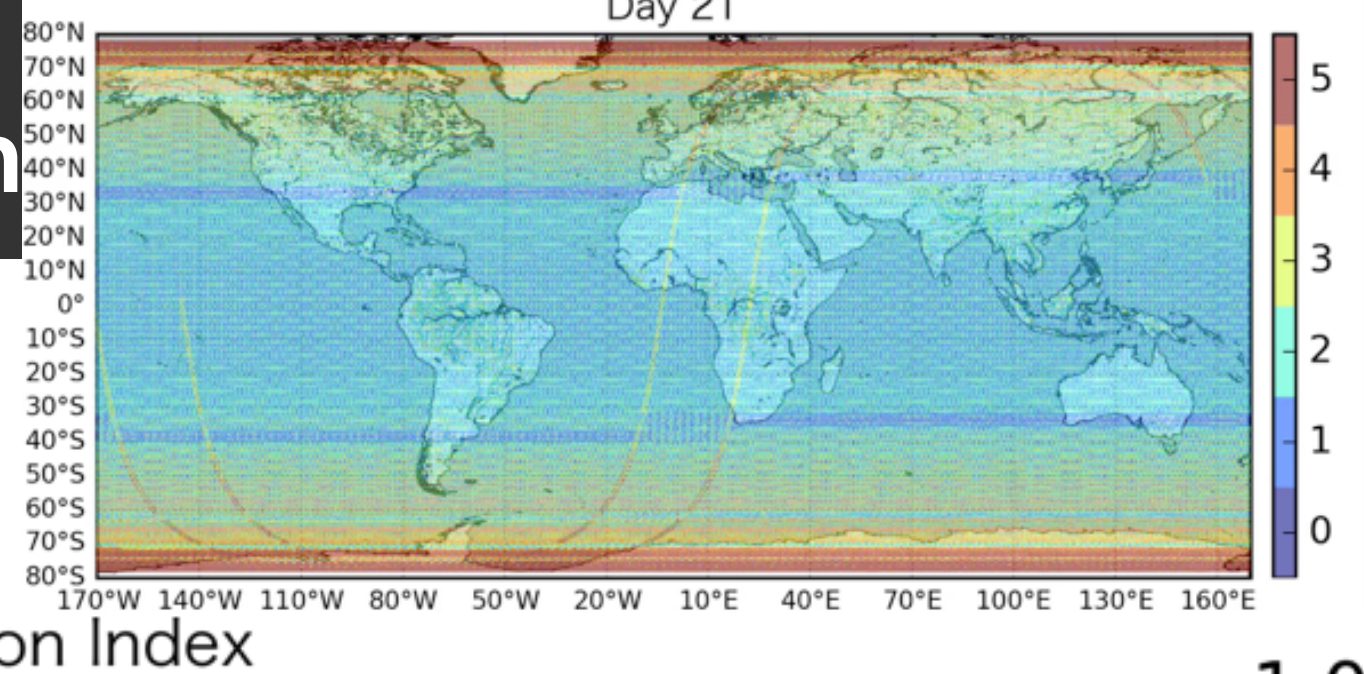
(B) Blind Runoff



3. Results and Discussion

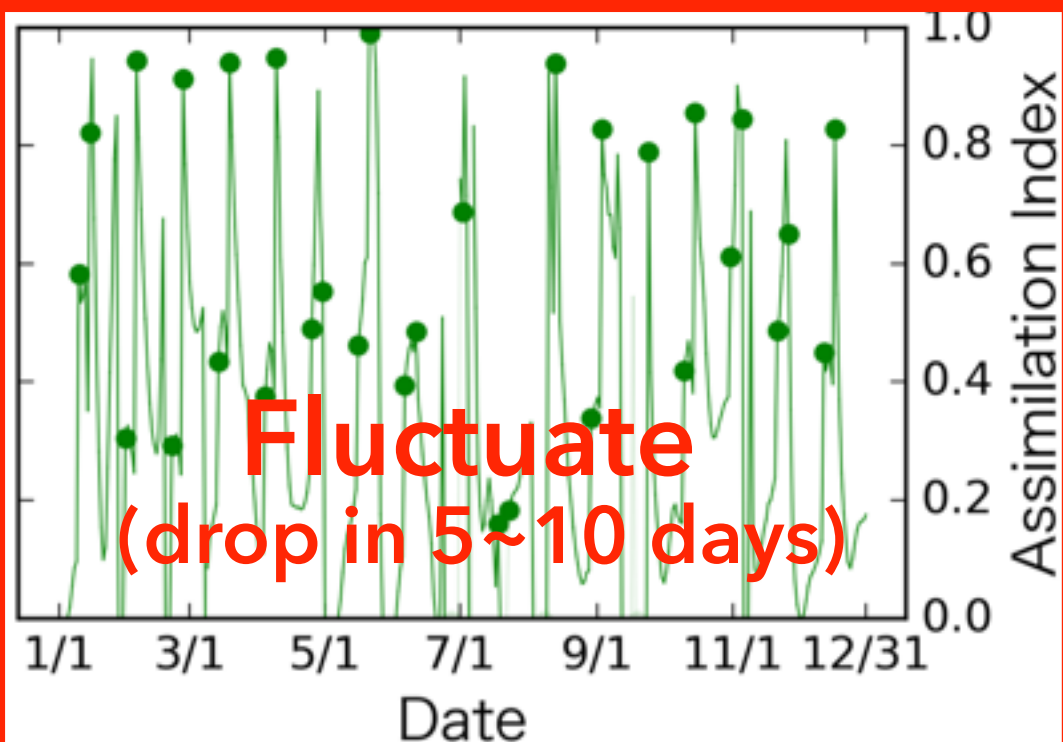


3. Results and Discussion

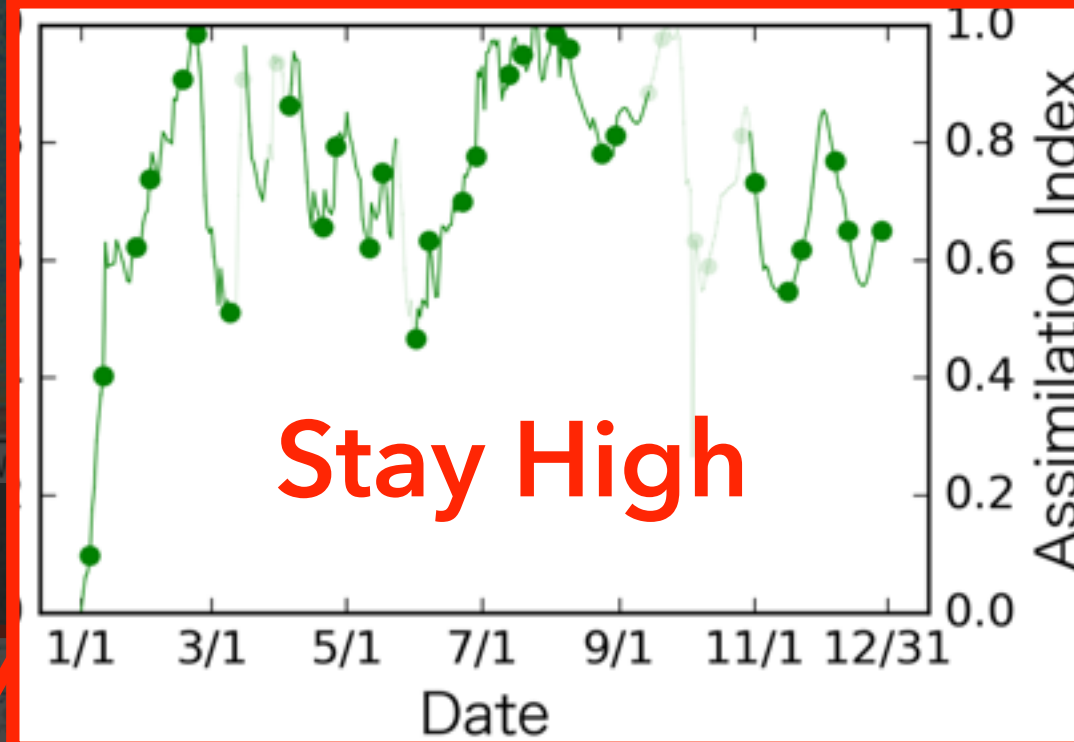


Assimilation Index (A)

Upstream

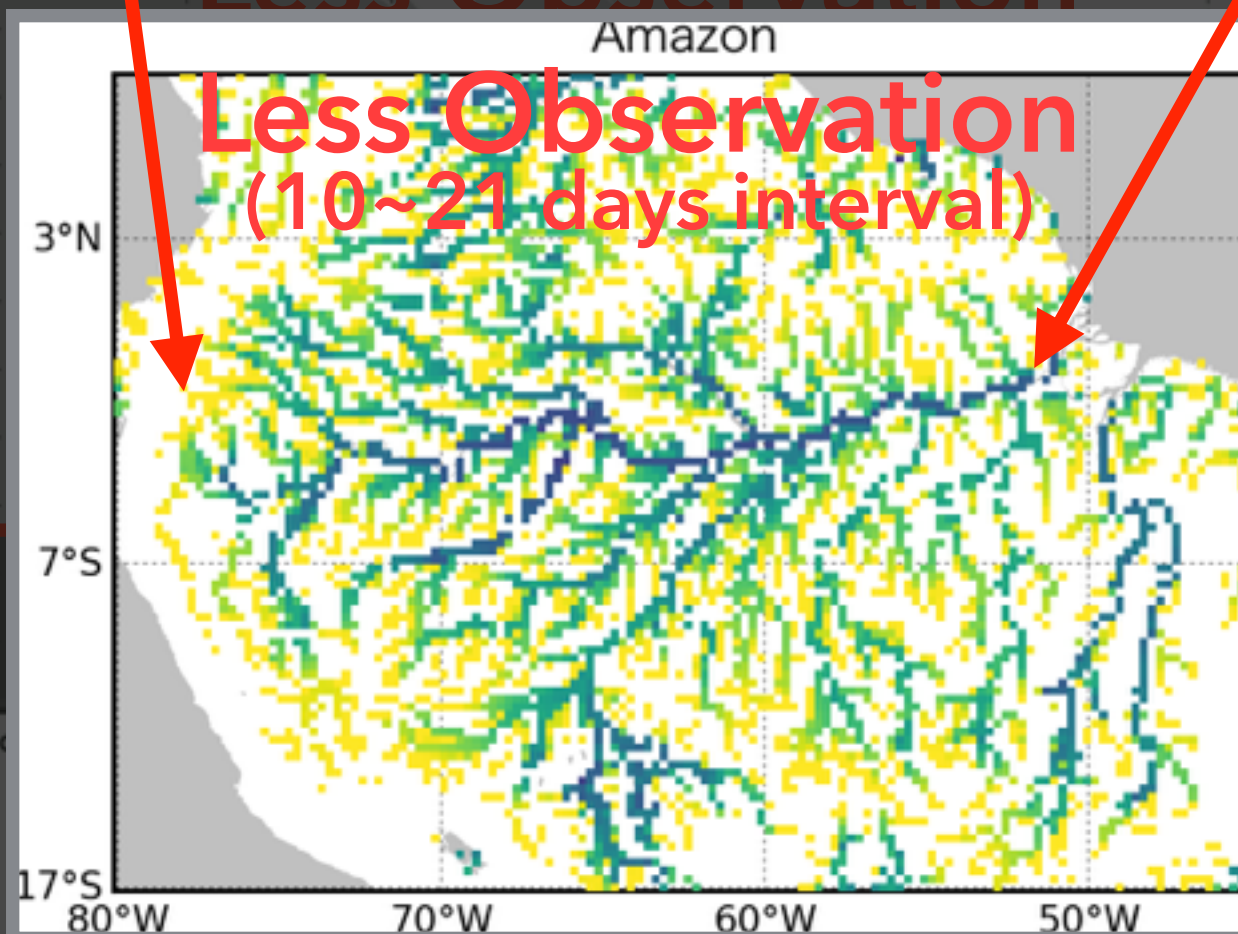


Downstream



Less Observation

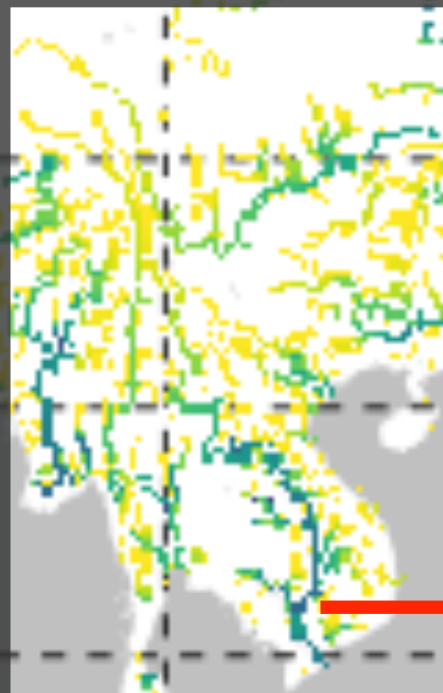
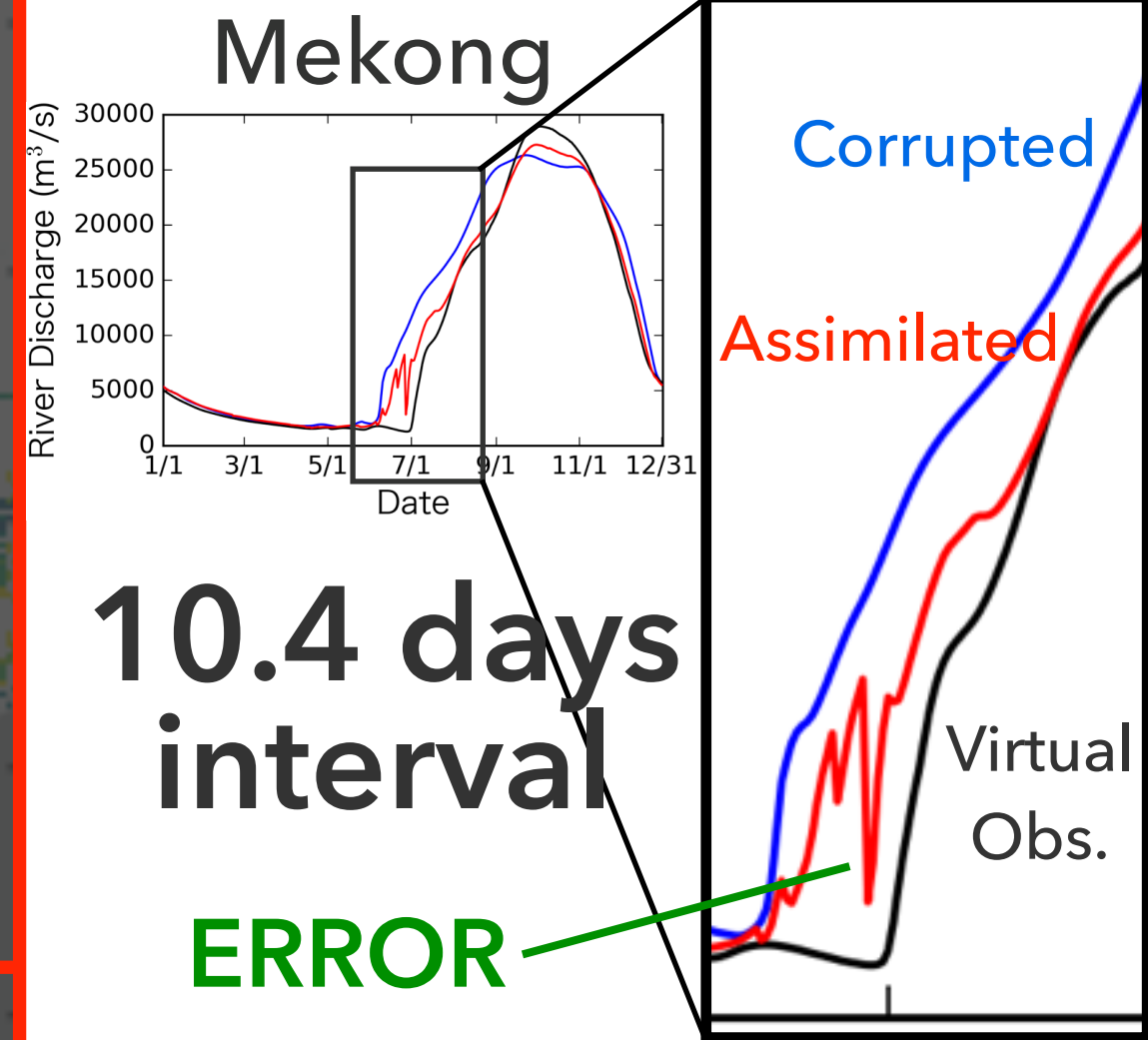
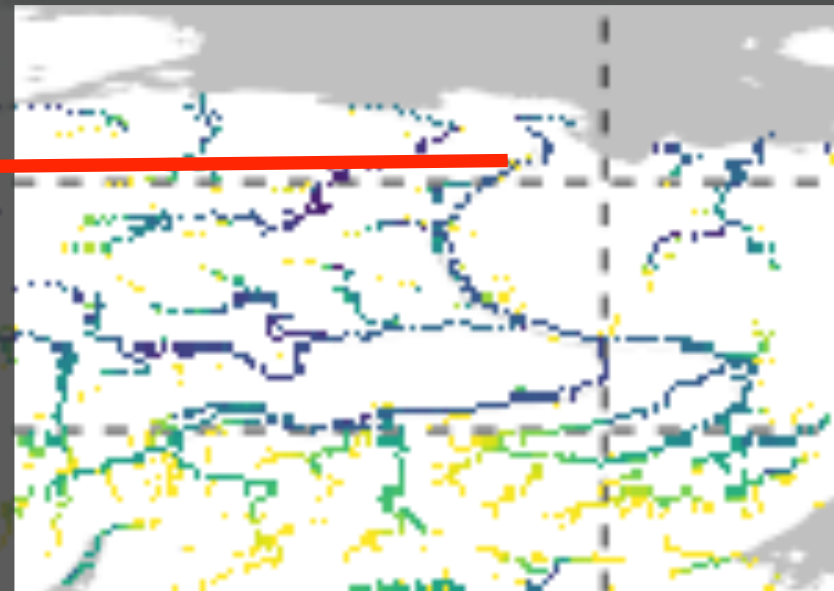
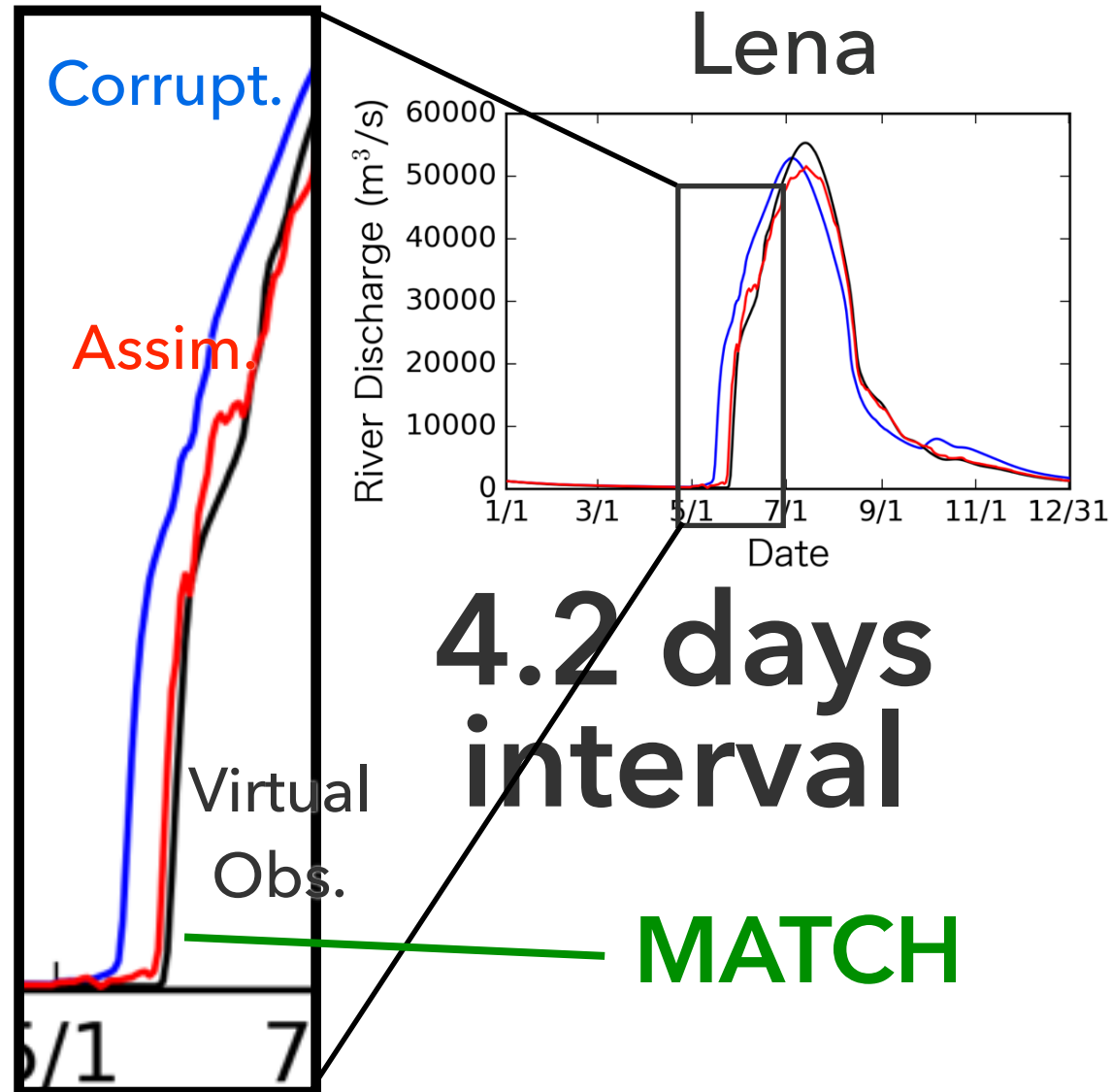
Less Observation
(10~21 days interval)



**Assimilated Discharge
Inflow**
(6,000,000 km²)

high at Downstream

Seasonal Transition

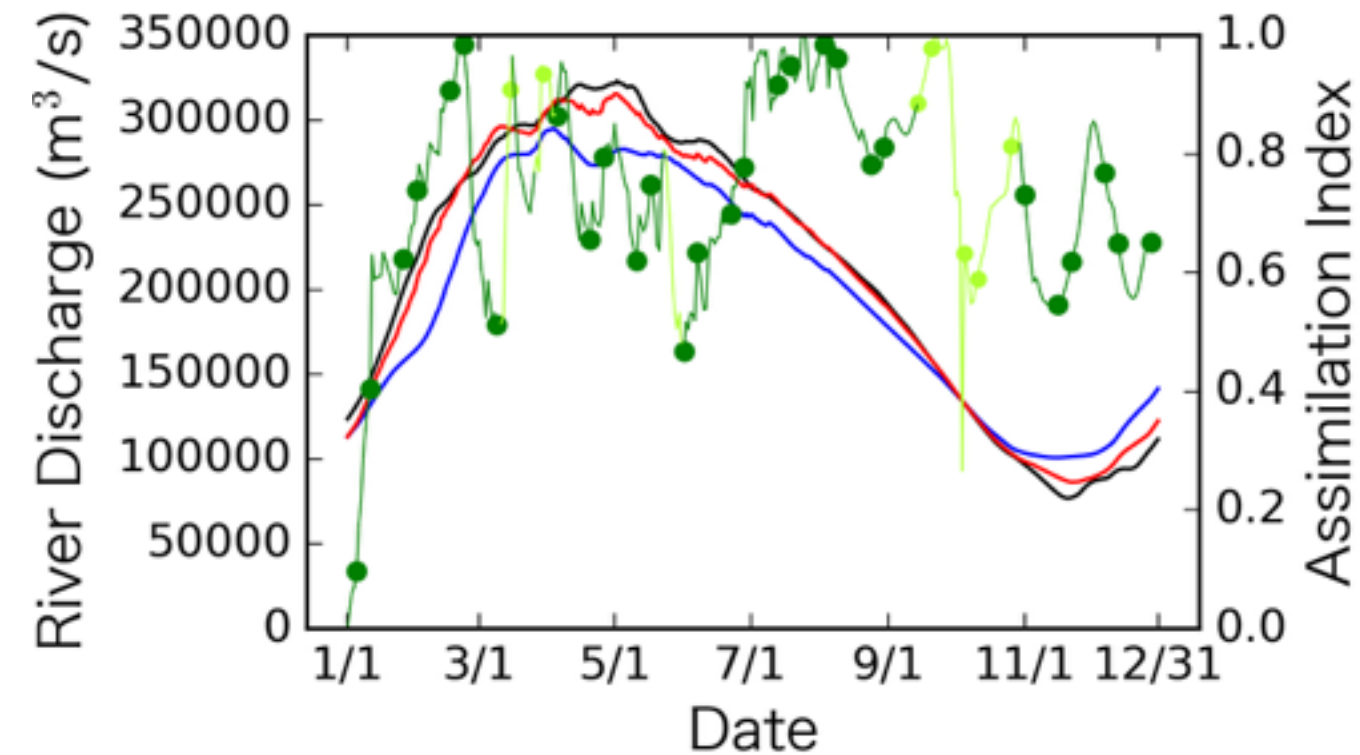


- Inflow from Upstream
- Local and Upstream Observation
- Observation Frequency

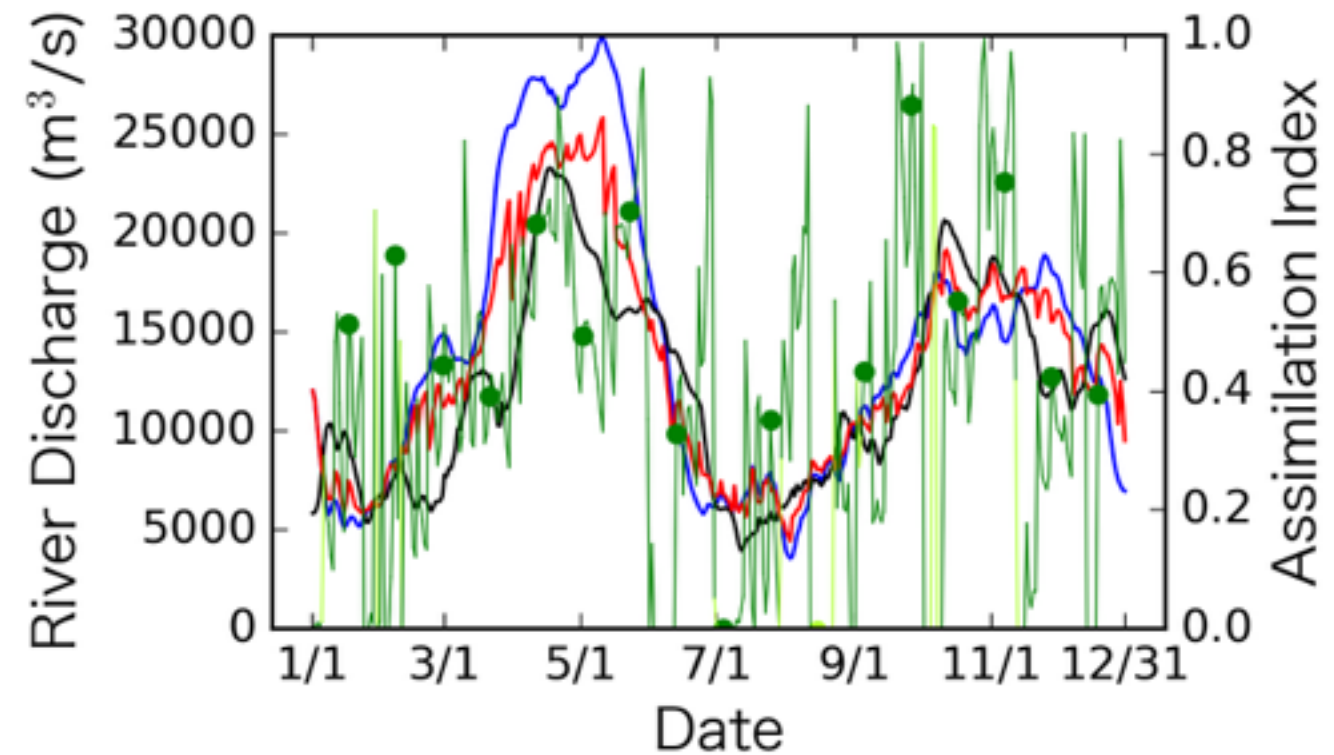
3. Results and Discussions

(B) Blind Runoff

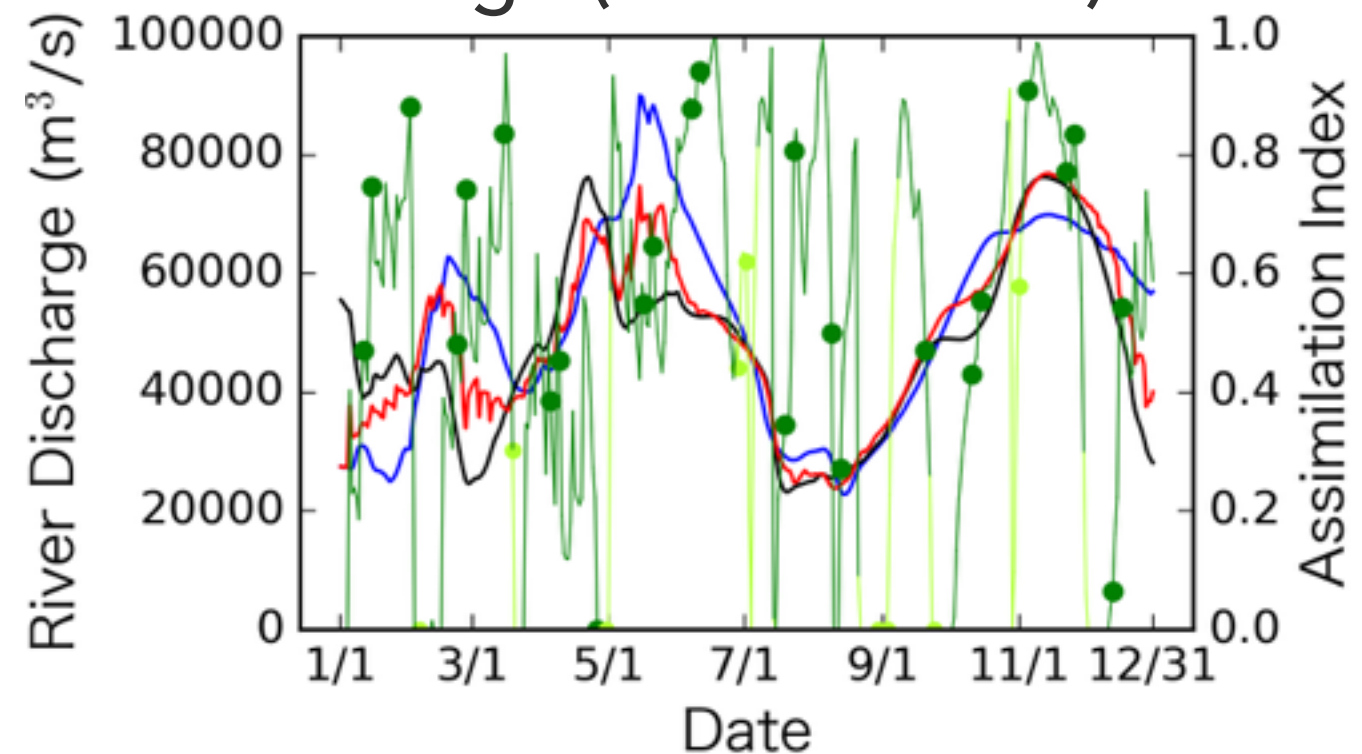
Amazon (Downstream)



Congo (Midstream)



Congo (Downstream)



Discharge

Virtual Obs.

Assimilated

Corrupted

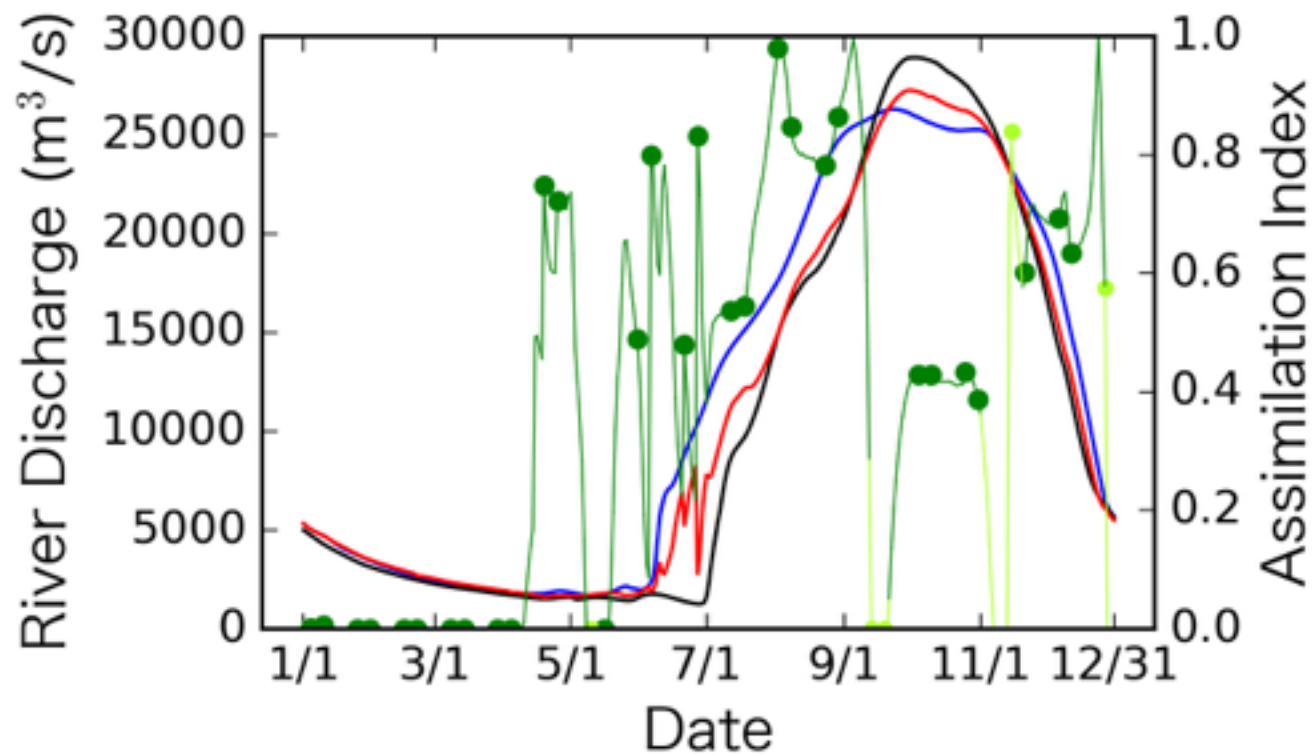
Assimilation Index

●: Observation

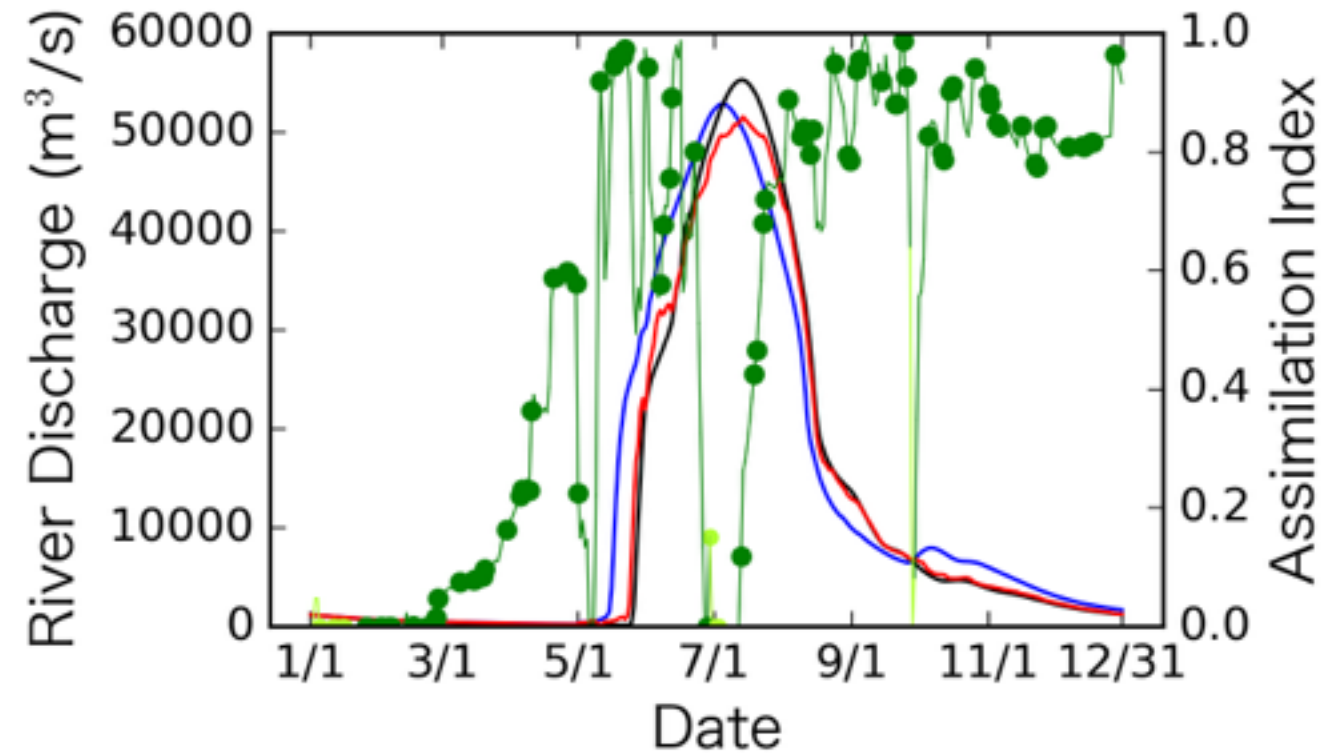
3. Results and Discussions

(B) Blind Runoff

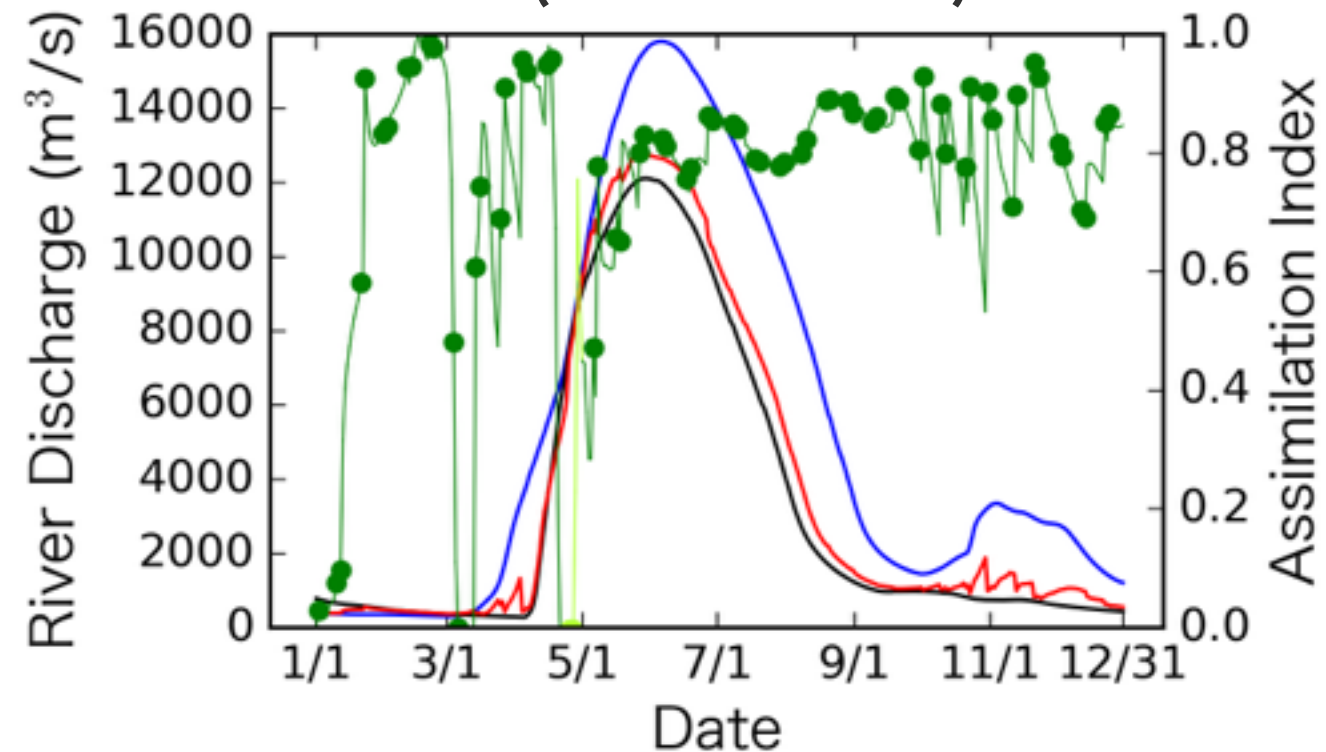
Mekong (Downstream)



Lena (Downstream)



Ob (Midstream)



Discharge

Virtual Obs.

Assimilated

Corrupted

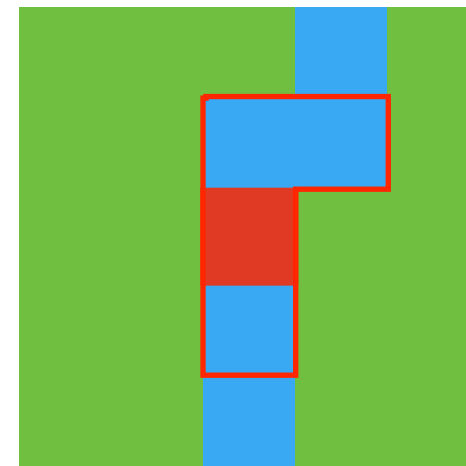
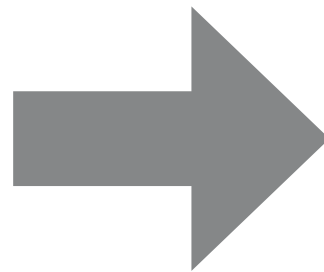
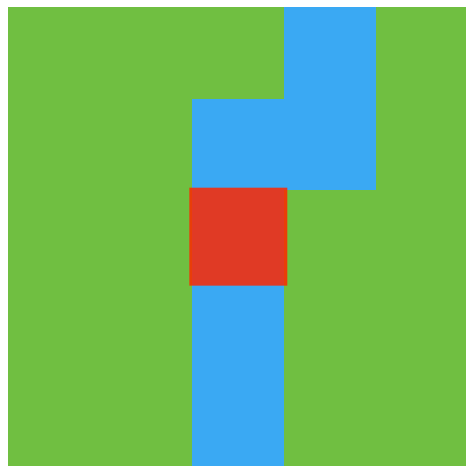
Assimilation Index

●: Observation

4. Future Steps

Enlarging Local Patch

Assimilation is possible only when there is Observation at that location



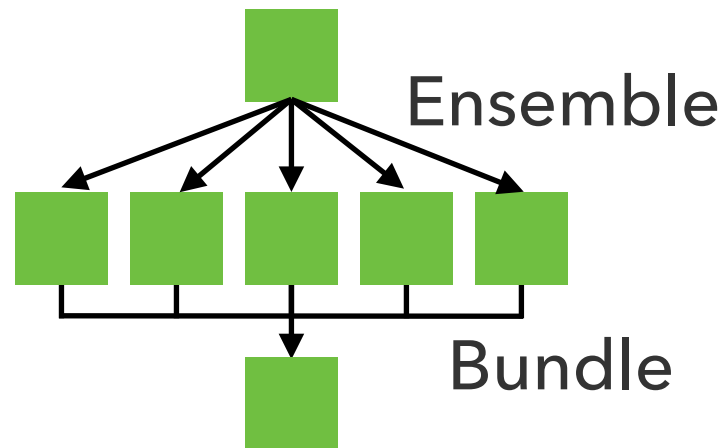
Originally, only target pixel is calculated at Assimilation

Using information of near-by pixel

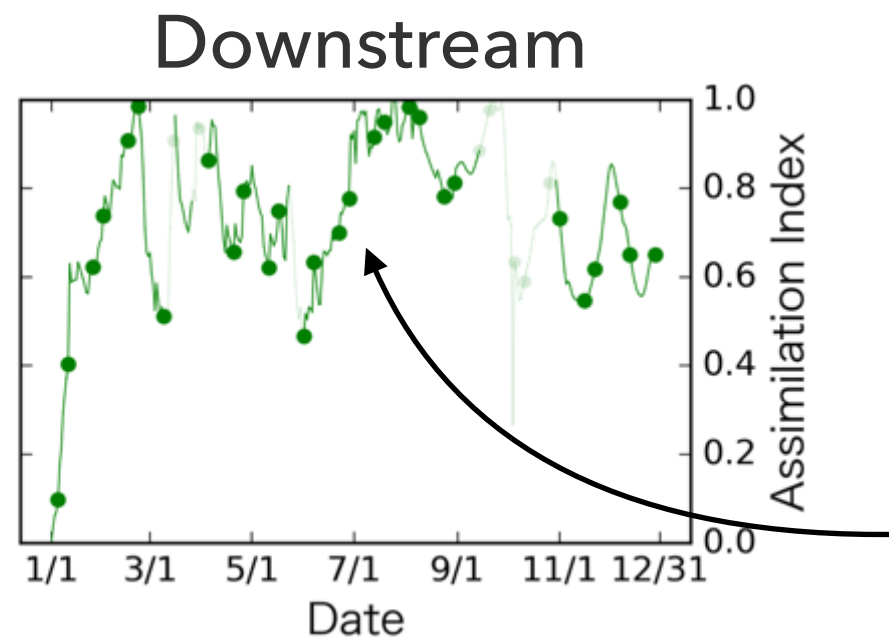
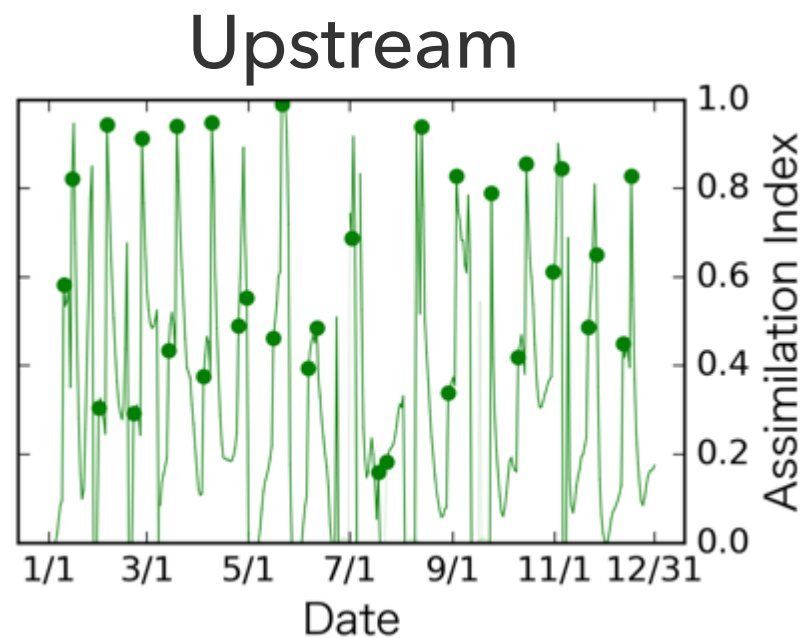
Assimilation Correction will be possible when there is observation at somewhere in the local patch

4. Future Steps

Improving Ensemble Spread



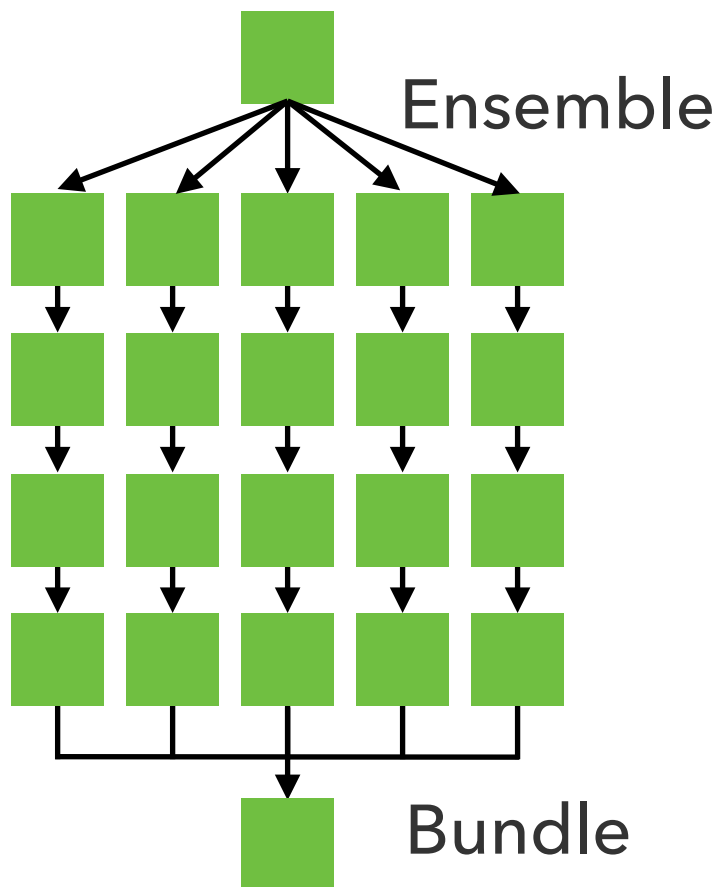
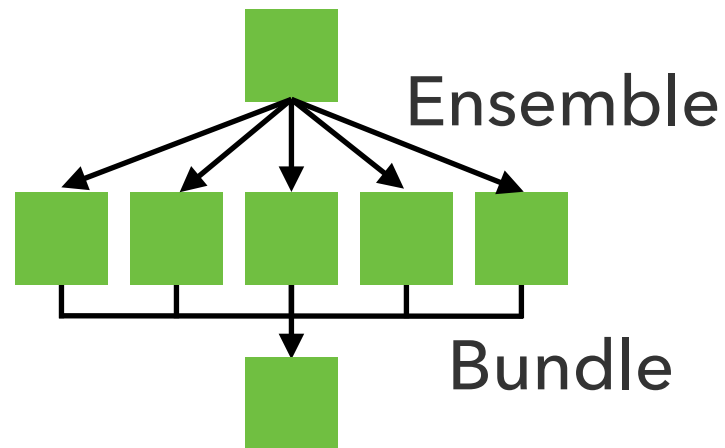
- Usually, ensemble is bundled in daily step
- This is often used in Atmospheric Model, which Ensemble easily spread
- However Ensemble and Bundle at daily step is too short for River Model to spread Ensemble



Observation at
Downstream is not
much effective

4. Future Steps

Improving Ensemble Spread



- Usually, ensemble is bundled in daily step
- This is often used in Atmospheric Model, which Ensemble easily spread
- However Ensemble and Bundle at daily step is too short for River Model to spread Ensemble



- Only bundle when there is Observation & Assimilation (once in few days)
- This will allow the Ensemble to Spread
- This will make assimilation at downstream more effective