

Modeling soil moisture with closure phase bias

Yujie Zheng^{1,2} and Heresh Fattahi³



(1) California Institute of Technology, USA

(2) University of Texas at Dallas, USA

(3) Jet Propulsion Laboratory, California Institute of Technology, USA

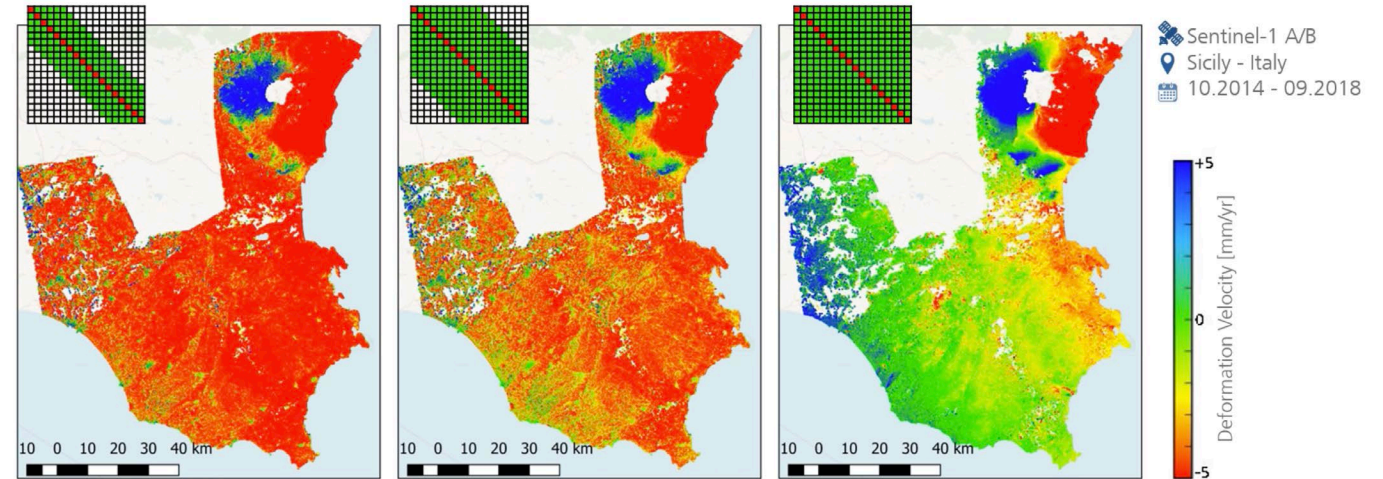
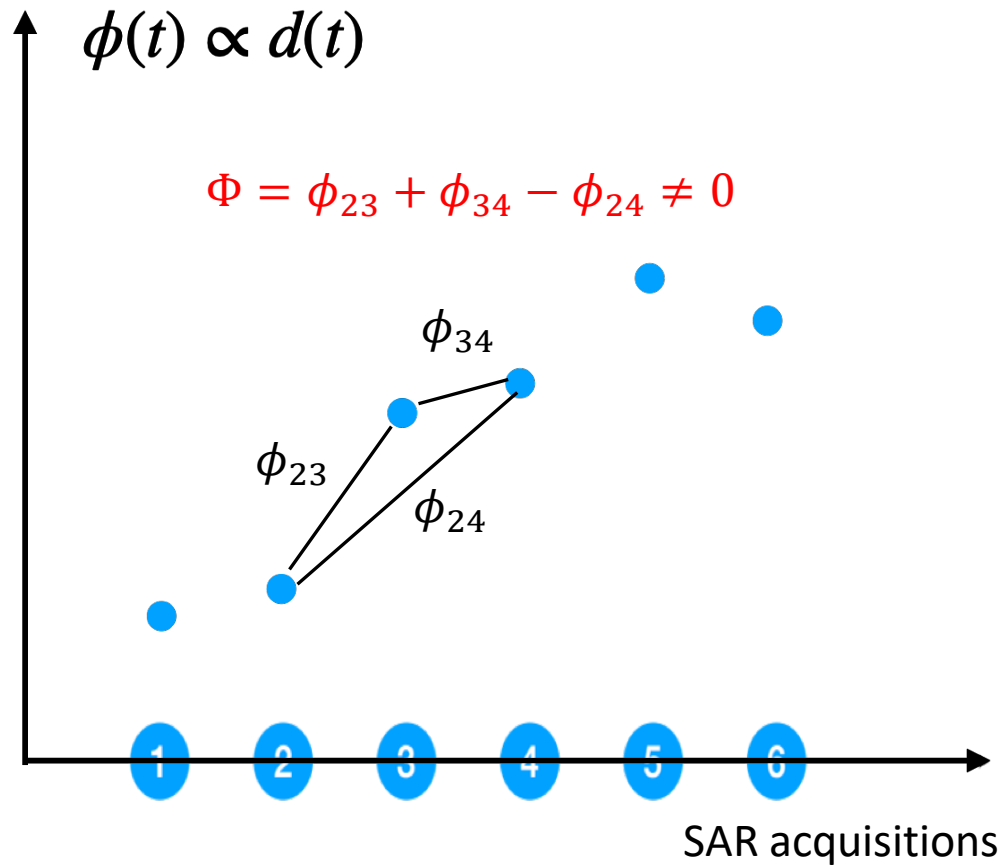
FRINGE 2023

University of Leeds, UK | 11 - 15 September 2023.

Non-zero closure phase



Short Temporal Bias in time-series



Ansari et.al. (2020)

Both phenomena are indicators of processes that induce complex changes of the interferometric measurements. (Zheng et al., 2022)

Bias correction methods :

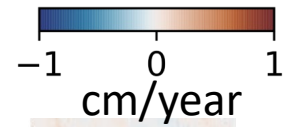
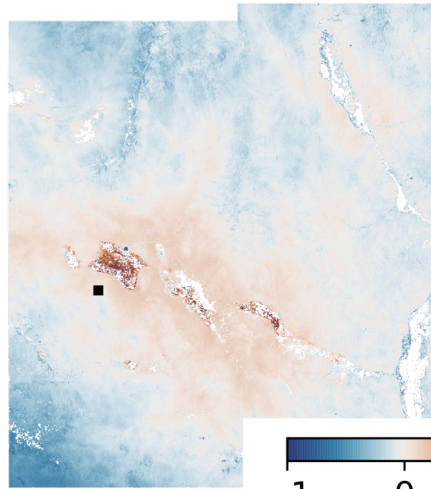
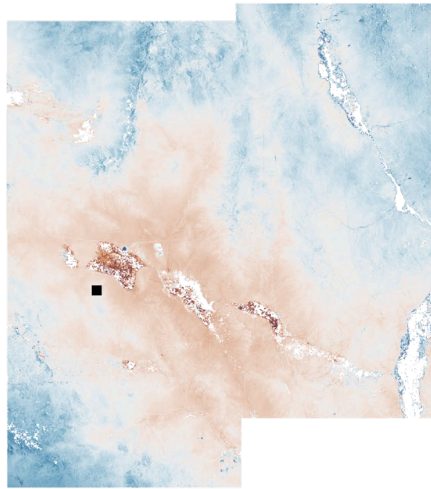
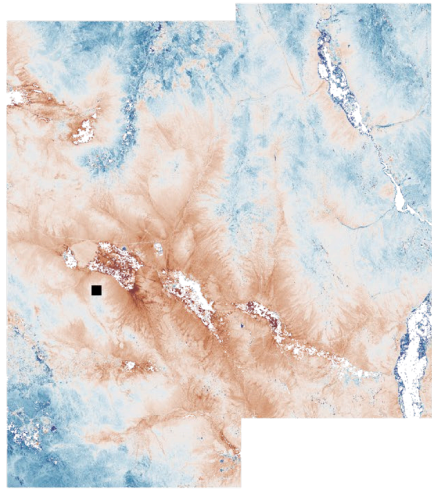
Falabella and Pepe, 2022; Maghsoudi et al., 2022; Zheng et al., 2022

Bias Correction with Sequential Closure Phase

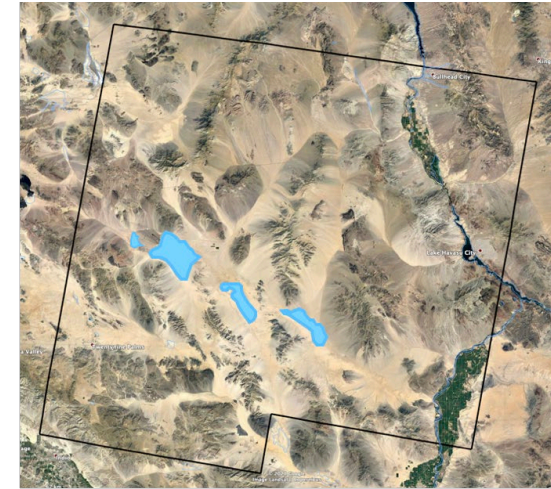
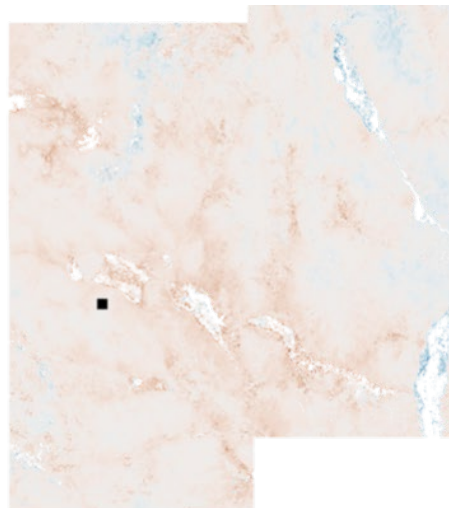
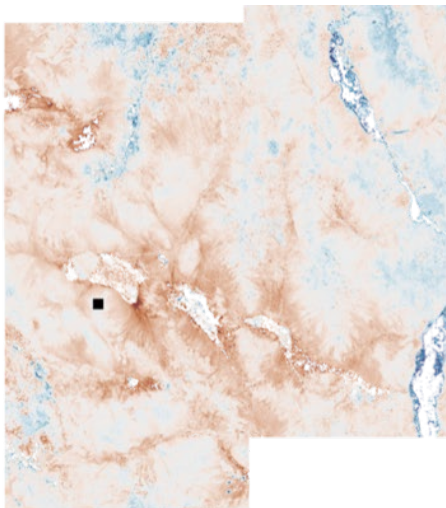
$\overline{\Delta t} = 9 \text{ days}$

$\overline{\Delta t} = 27 \text{ days}$

$\overline{\Delta t} = 49 \text{ days}$



Estimated Bias



Sentinel-1A/B



Barstow-Bristol Trough, CA, United States

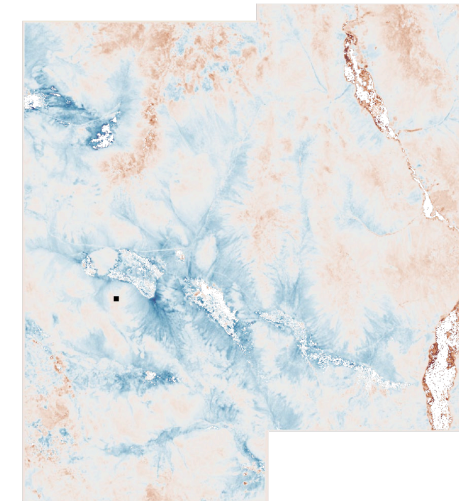


Feb 2017 - Jan 2021



SBAS, MintPy

Cumulative Sequential Closure Phase



Zheng et al., 2022

Correction code available in MintPy

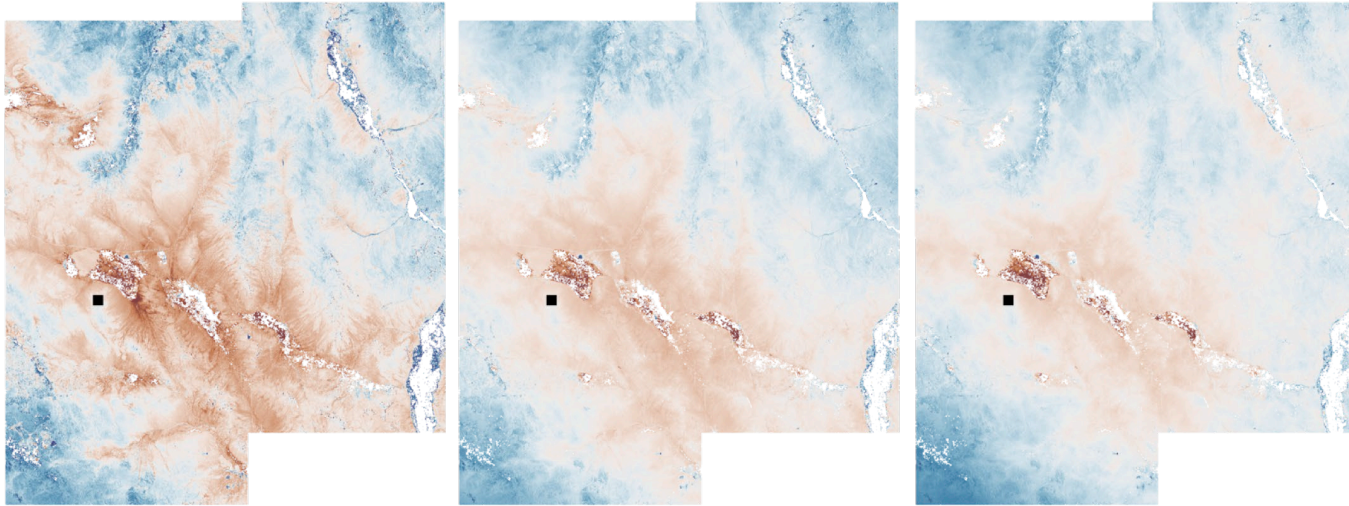
Bias Correction with Sequential Closure Phase

$\overline{\Delta t} = 9 \text{ days}$

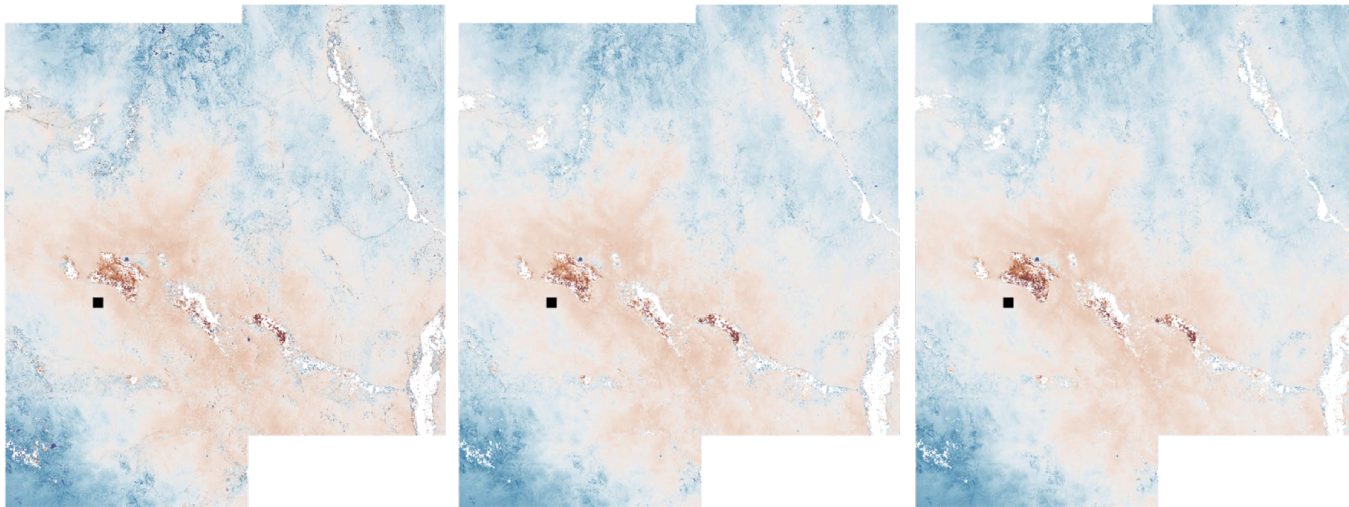
$\overline{\Delta t} = 27 \text{ days}$

$\overline{\Delta t} = 49 \text{ days}$

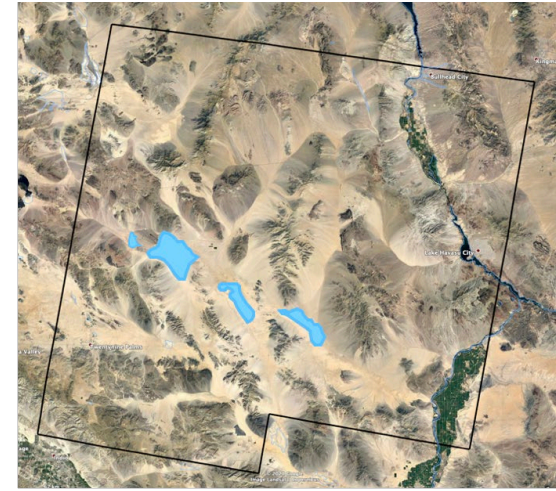
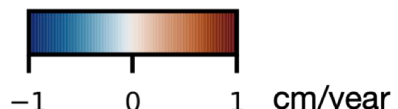
Before Correction



After Correction

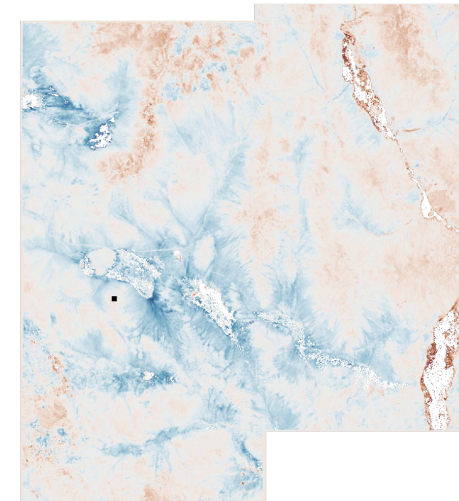


Azimuth
Range



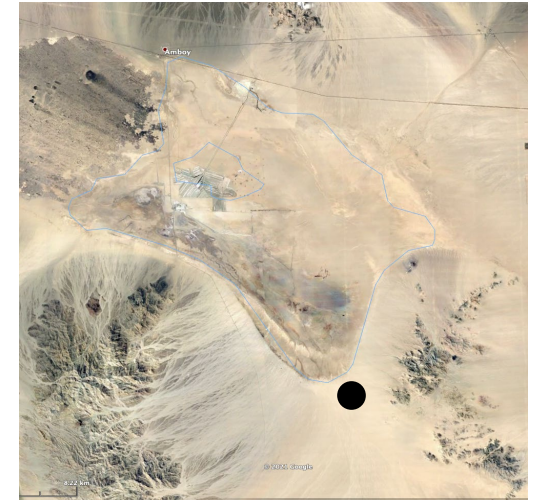
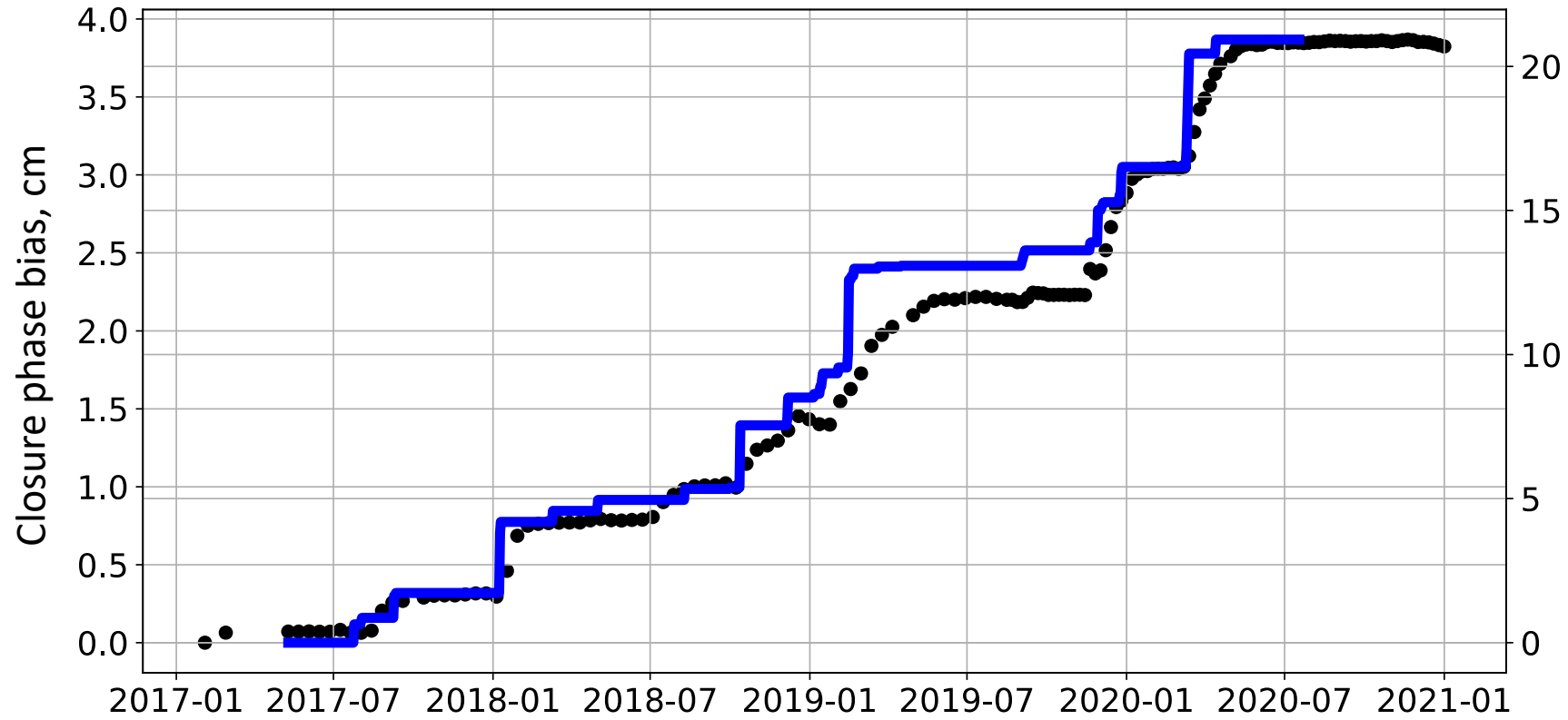
- Sentinel-1A/B
- Barstow-Bristol Trough, CA, United States
- Feb 2017 - Jan 2021
- SBAS, MintPy

Cumulative Sequential Closure Phase



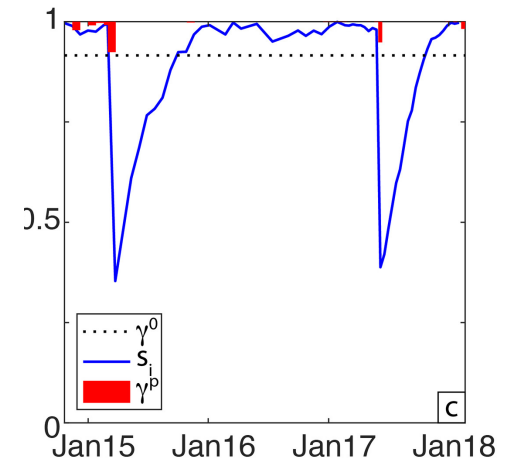
Zheng et al., 2022
Correction code available in MintPy⁴

Closure Phase Bias as SIGNAL



Bristol Dry lake, CA

Cumulative Precipitation, cm



Jordan et al., (2020)

A discrete interferometric model for soil moisture



k – wavenumber

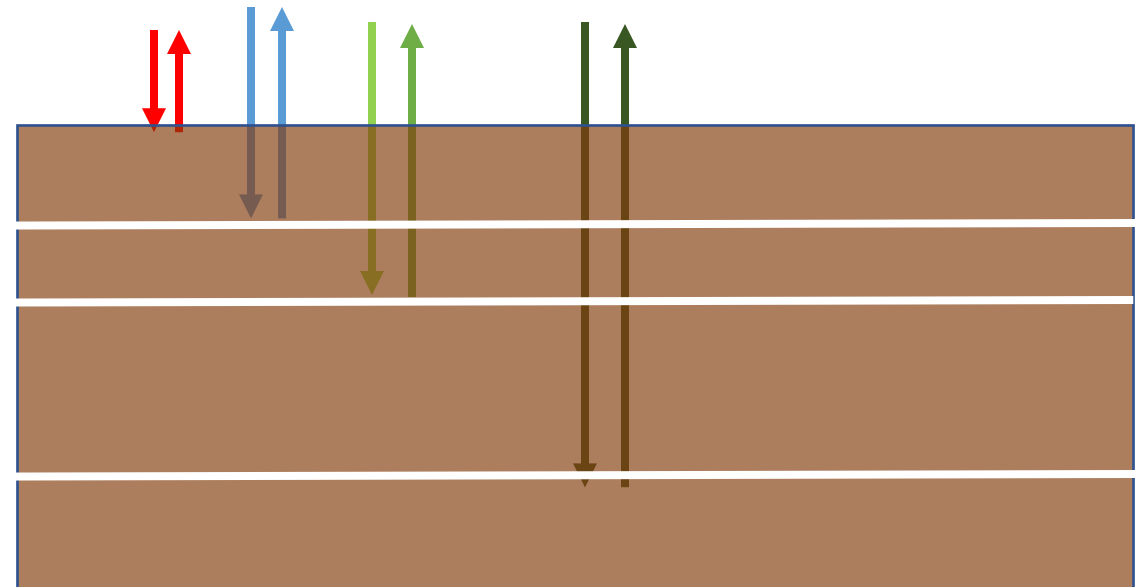
$$\text{Air, } k = \frac{2\pi}{\lambda}$$

Soil, k' is complex



Analytical model for soil moisture
e.g., De Zan et al., 2014; Michaelides, 2020

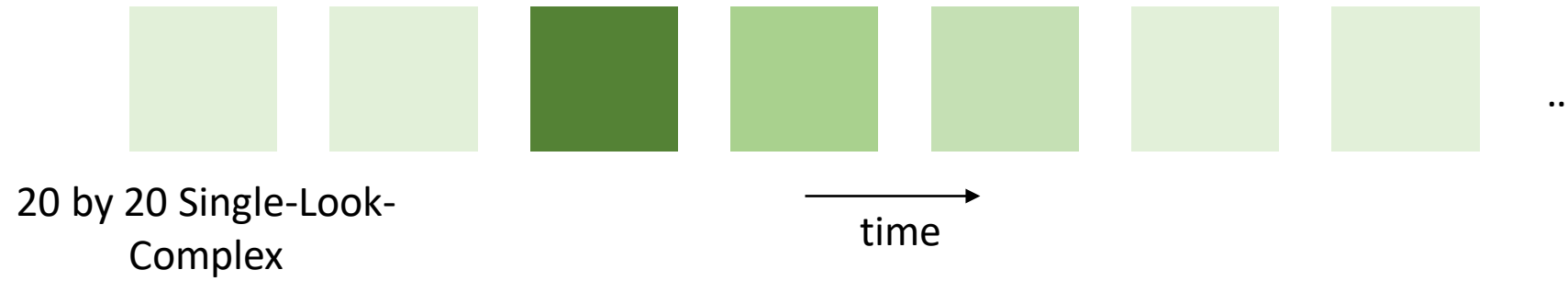
Easy to implement for numerical simulation.
Easier to adapt to more realistic scenarios.
Can start from single-look radar return.



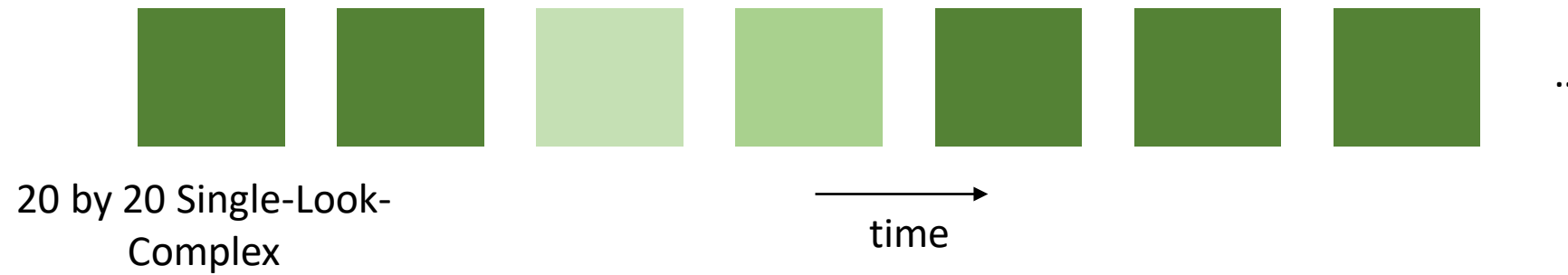
Discretized model for soil moisture

From soil moisture to closure phase bias – Numerical Simulation

Case I – Arid environment



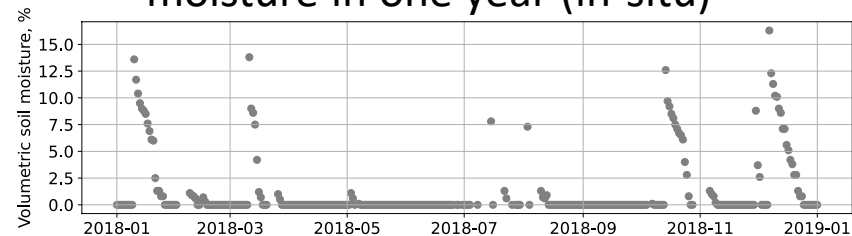
Case II – Wet environment



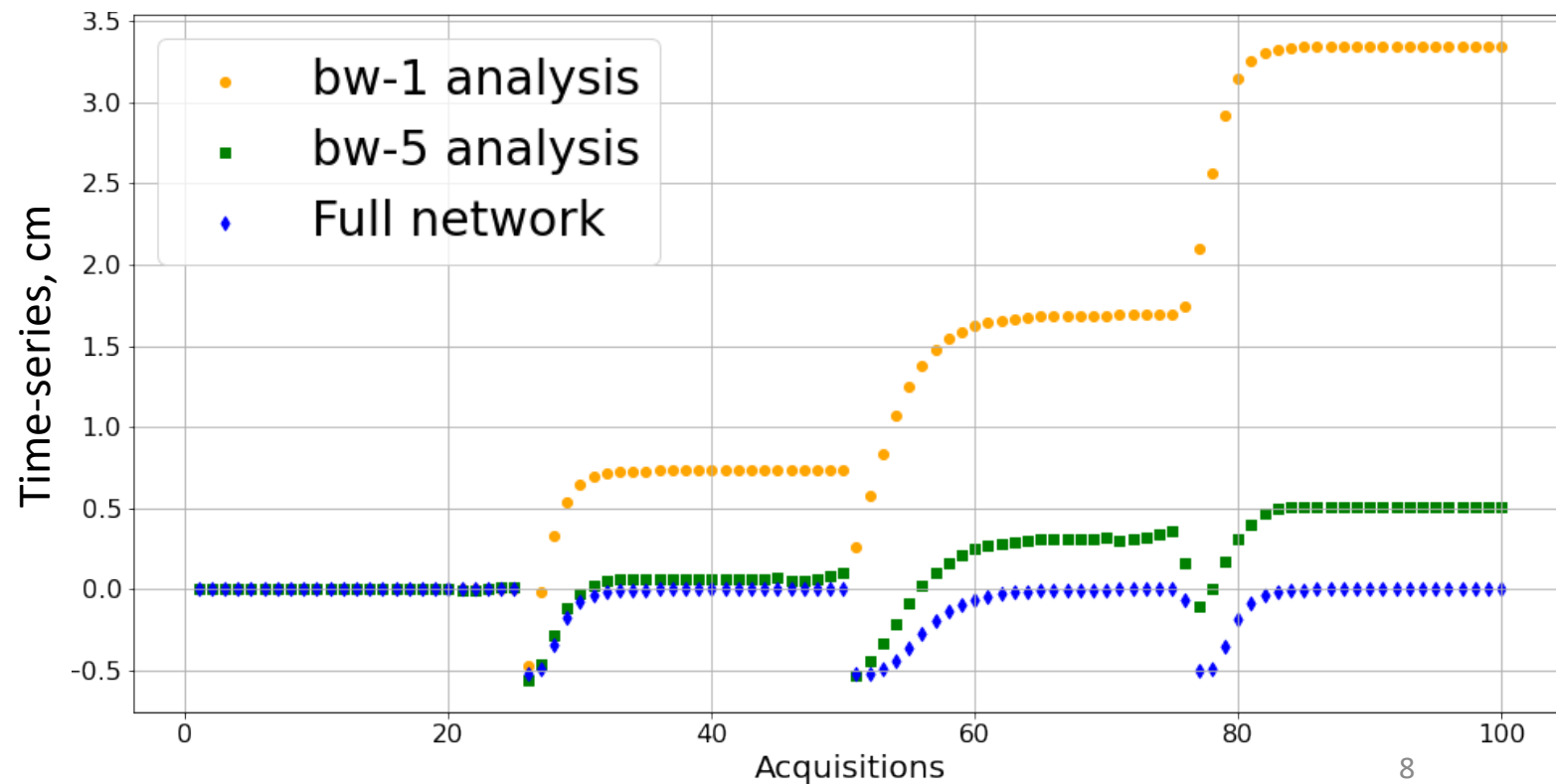
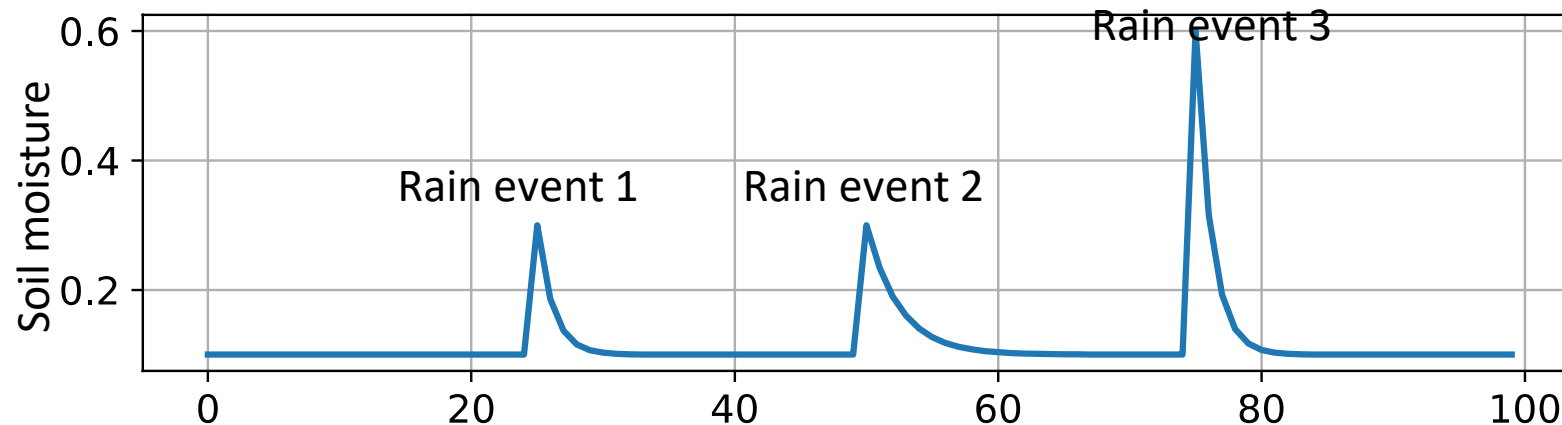
From soil moisture to closure phase bias – Numerical Simulation

Case I – Arid environment

Example of arid environment soil moisture in one year (in-situ)



- Step 1: Simulate 100 SAR acquisitions, each acquisition contains 400 pixels
- Step 2: Form interferograms and multi-look by 400
- Step 3: Time-series analysis of different bandwidth

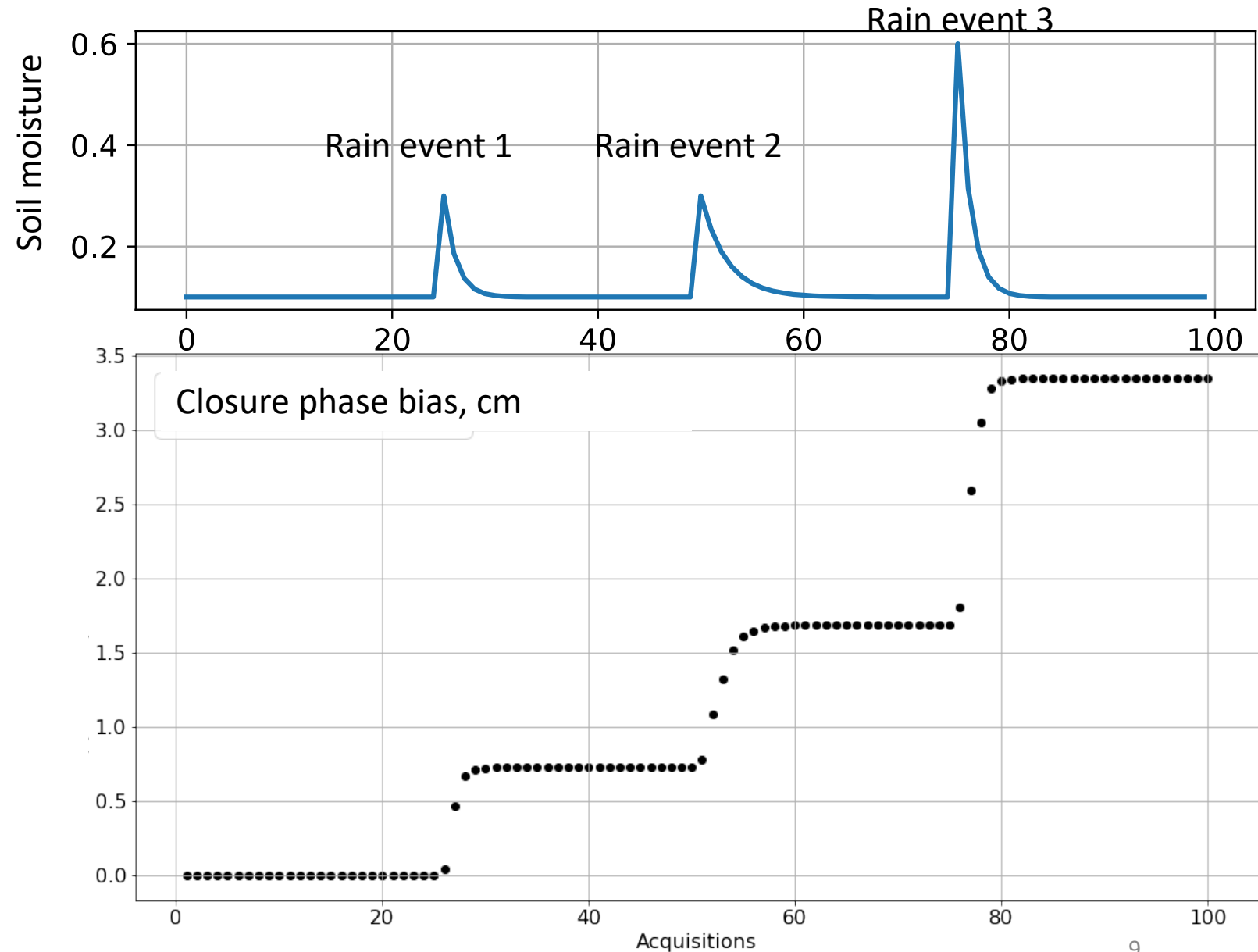


From soil moisture to closure phase bias – Numerical Simulation

Case I – Arid environment

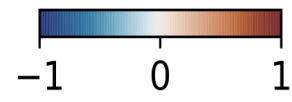
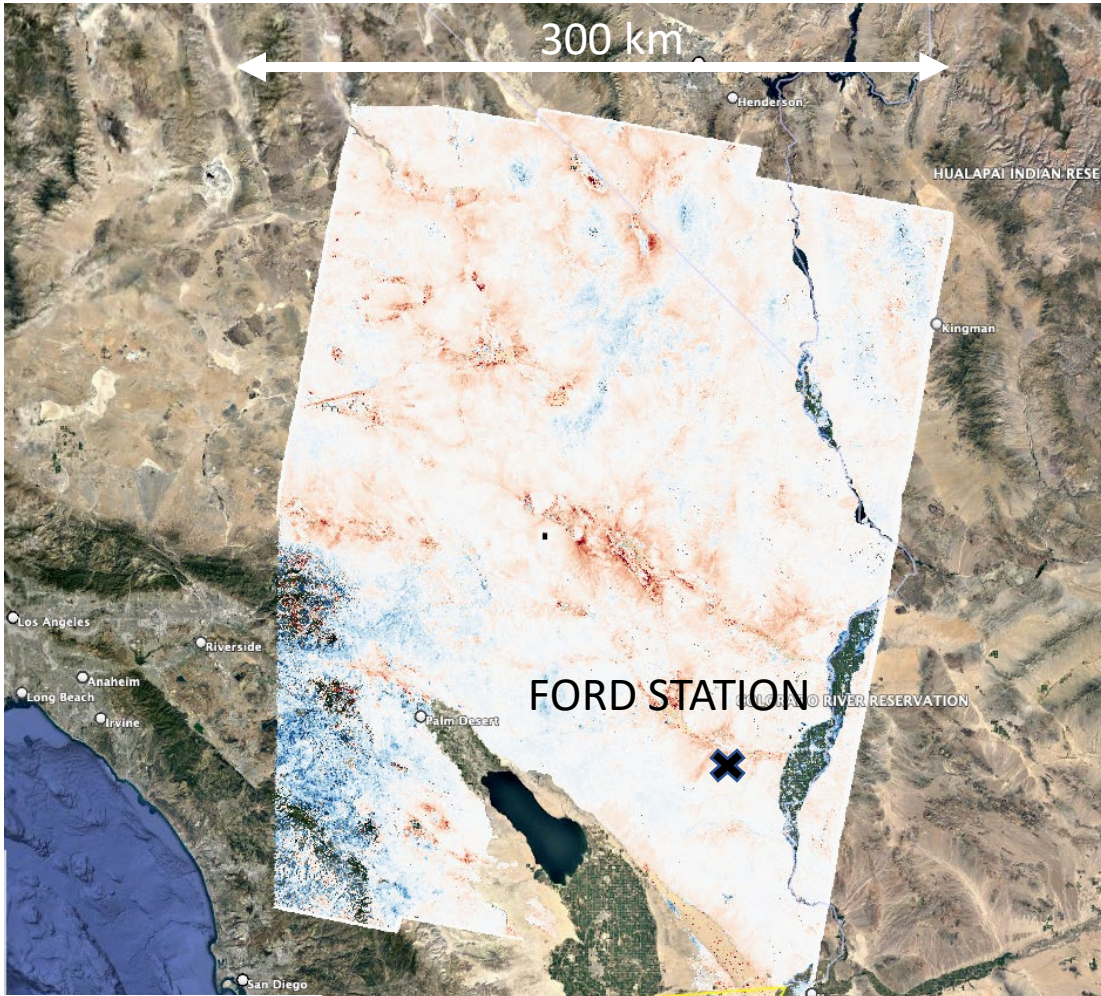
Key findings:

1. “Uplift” in closure phase bias corresponds to soil drying after rainfall.
2. Uplift magnitude depends on soil moisture decay rate and variation
3. Slower decay and greater variation result in larger uplift



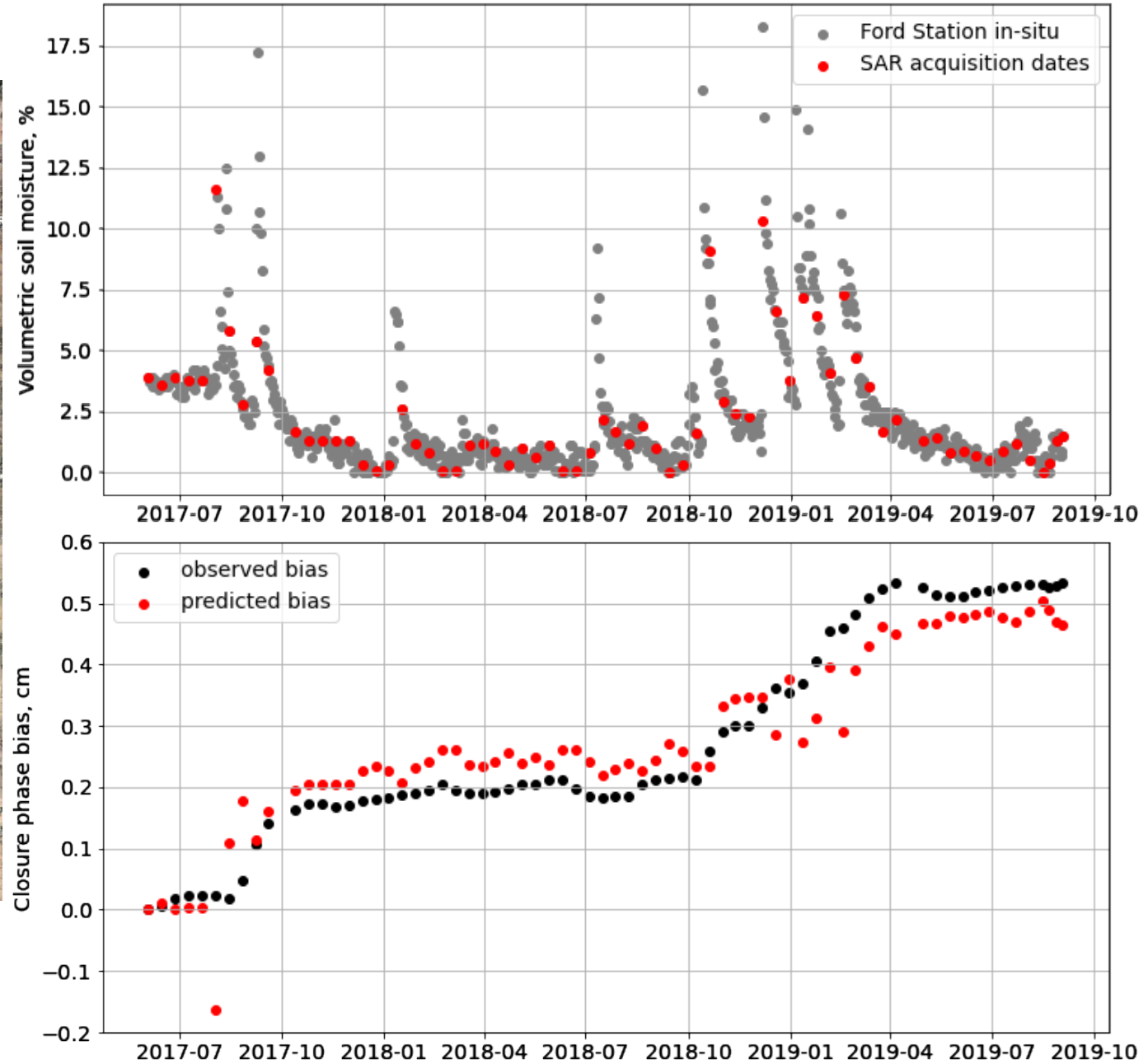
Forward model validated with in-situ measurements

Case I – Arid environment

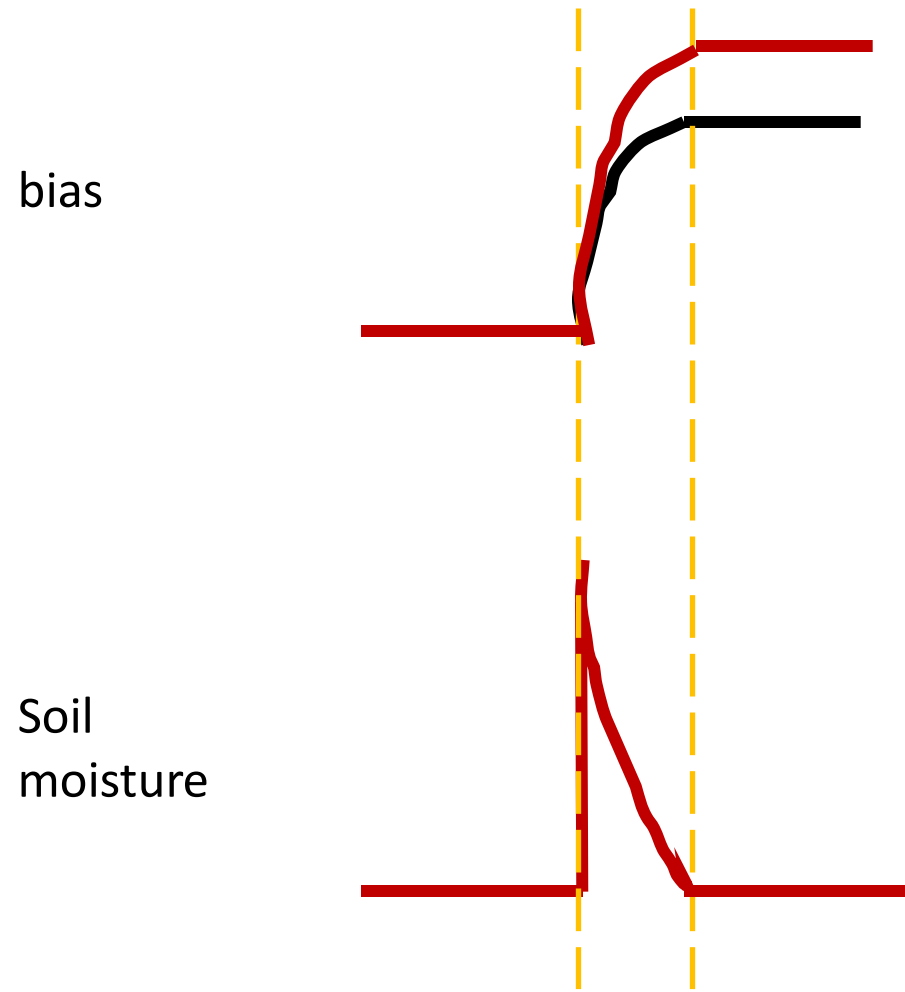


Closure phase bias, cm/year

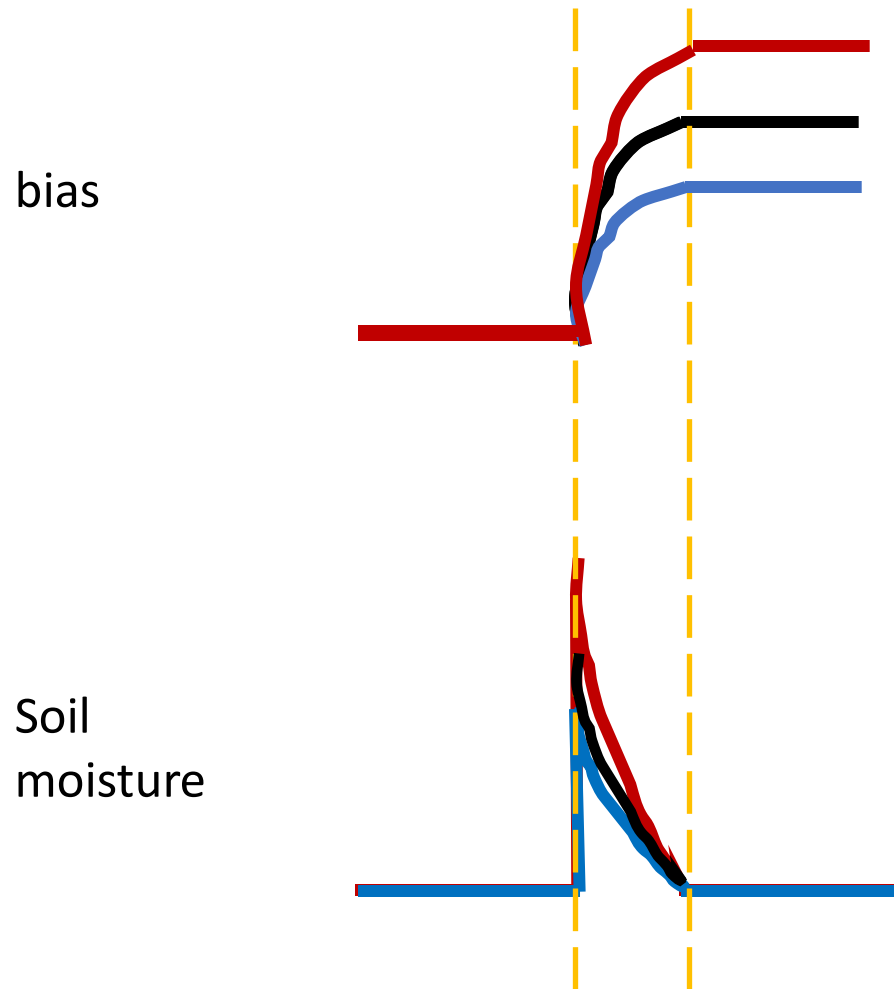
Pixel size : 300 meter



From closure phase bias to soil moisture: inversion scheme



From closure phase bias to soil moisture: inversion scheme



Step 1 – identify disturbance windows

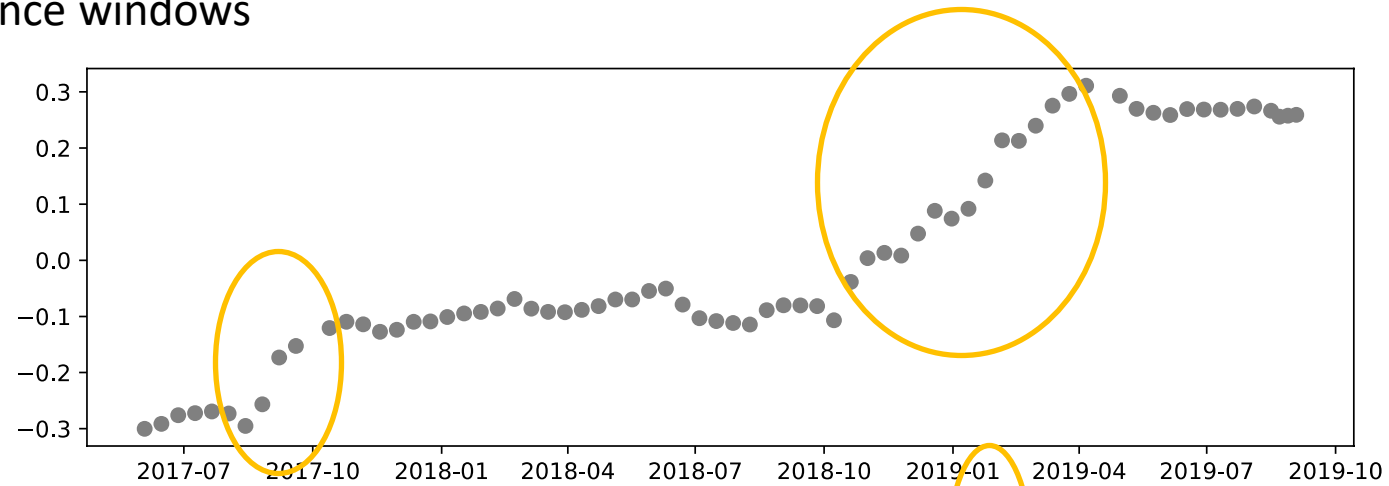
Step 2 – propose a soil moisture time-series with base value s_0 and an exponential decay model $s = s_0 e^{(-)t/\tau}$

Step 3 -- solve for s_0 and τ that gives the best fit to the bias time-series

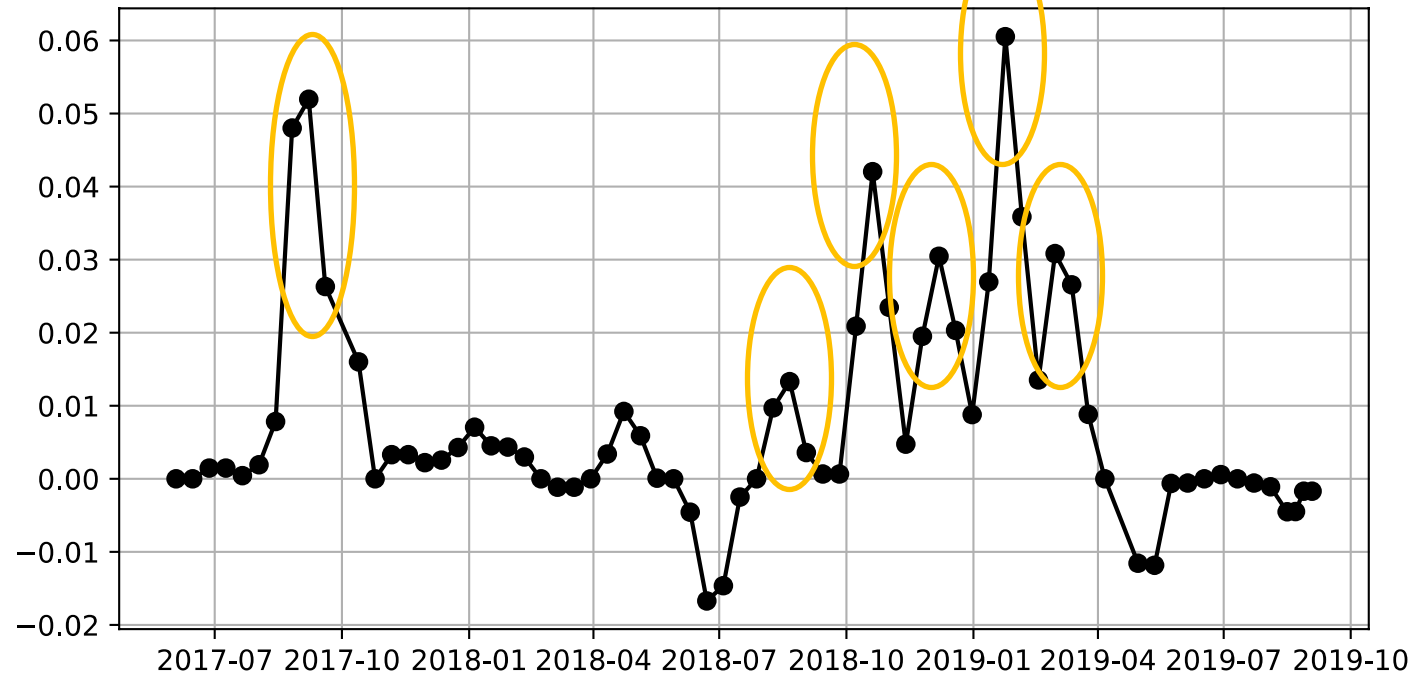
From closure phase bias to soil moisture: initial results

Step 1 – identify disturbance windows

Closure phase bias



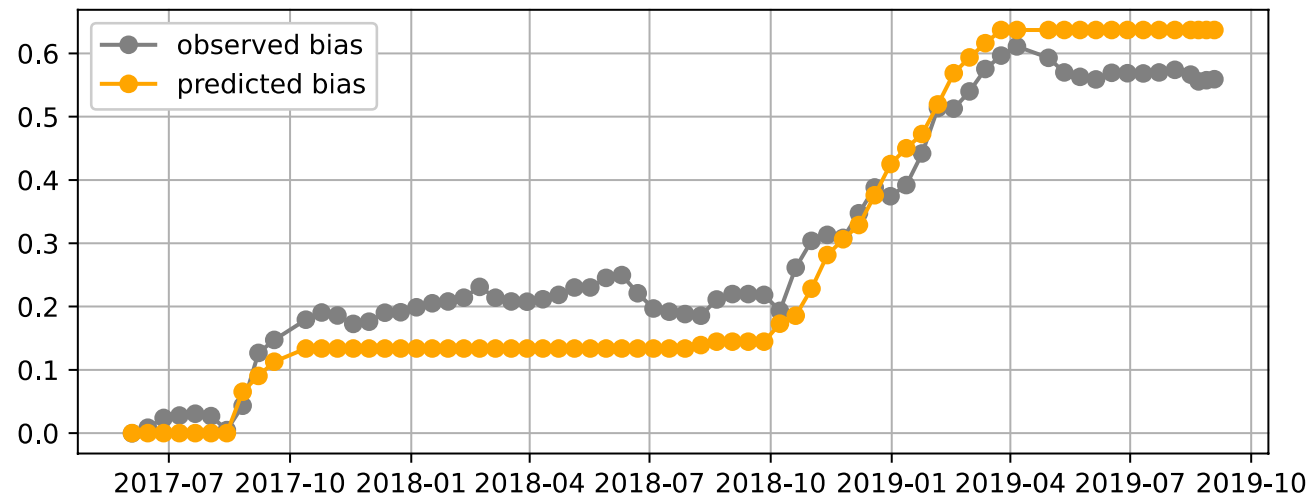
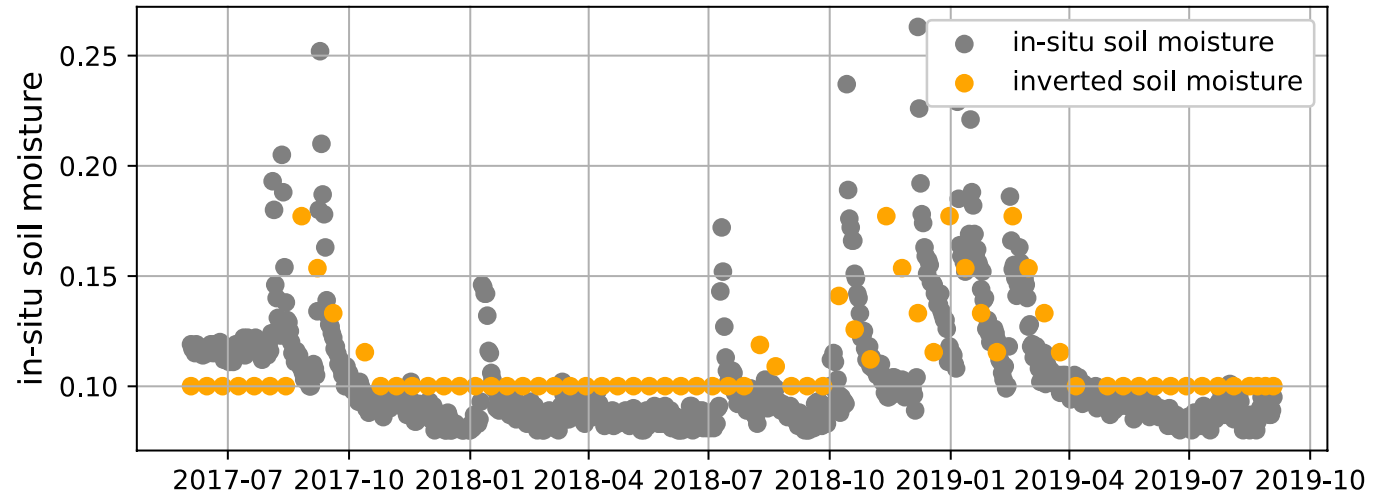
Closure phase bias gradient



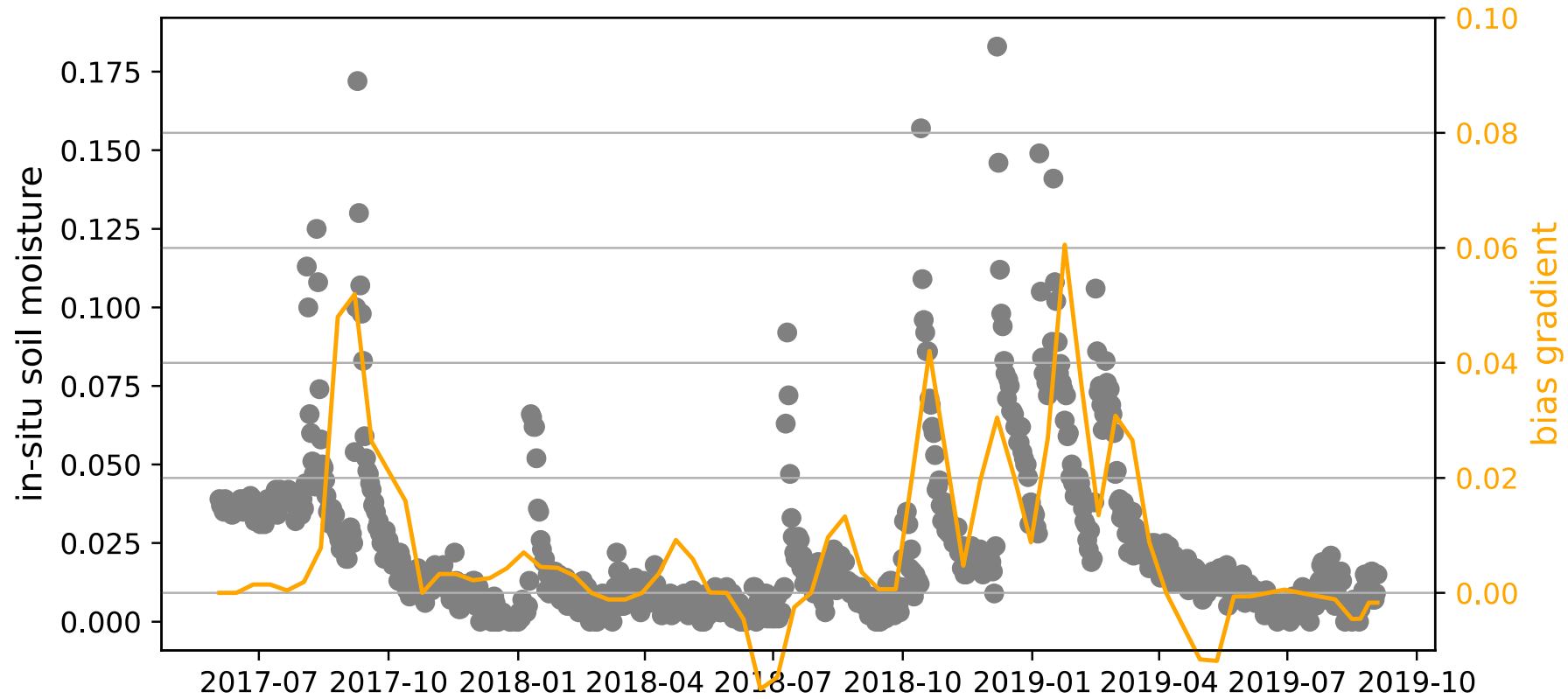
Six
disturbance
events
detected

From closure phase bias to soil moisture: initial results

Step 2 & 3 – find the best fit soil moisture time-series



From closure phase bias to soil moisture: bias gradient as a proxy



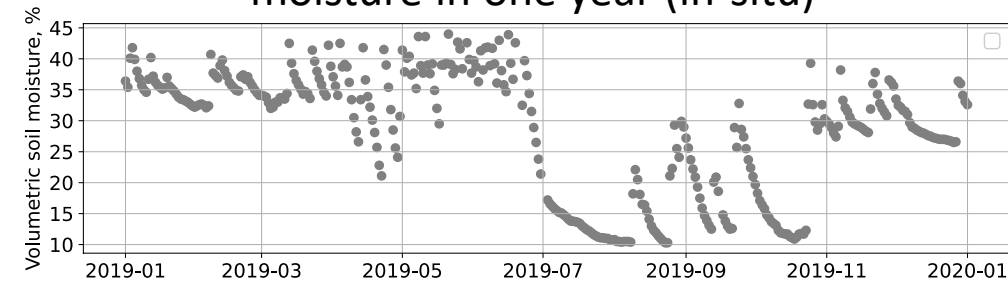
● Bias gradient

● In-situ soil moisture measurements

From soil moisture to closure phase bias – Numerical Simulation

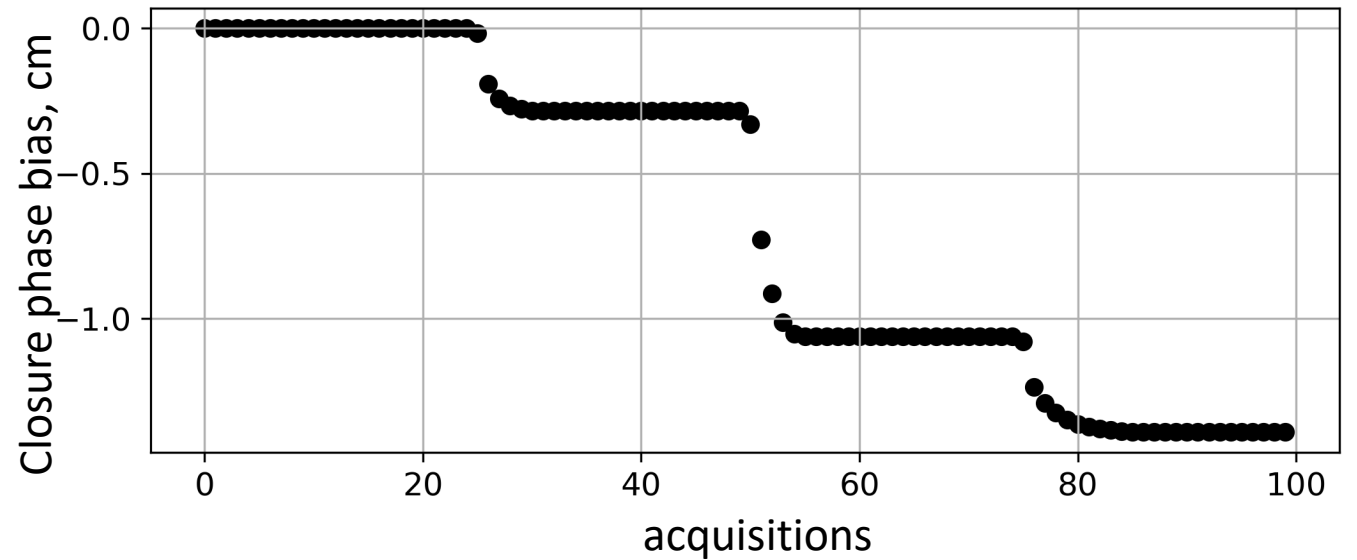
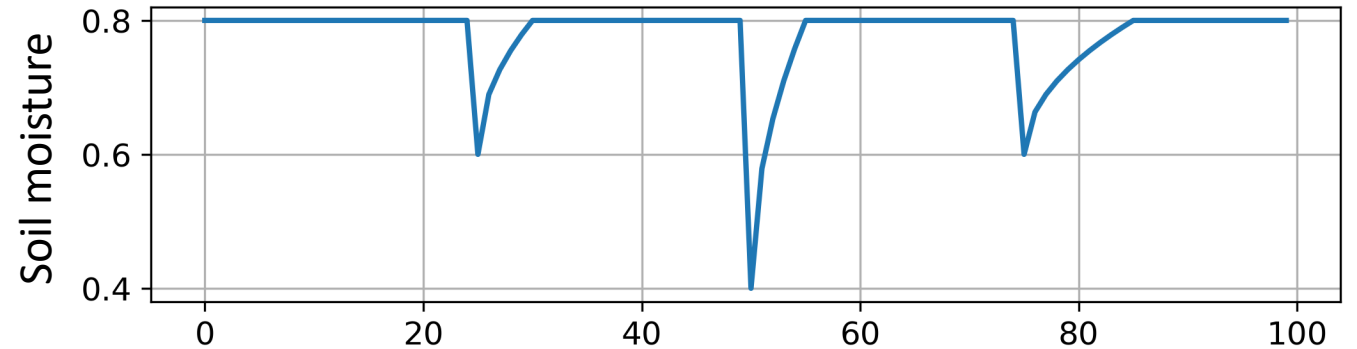
Case II – Wet environment

Example of wet environment soil moisture in one year (in-situ)



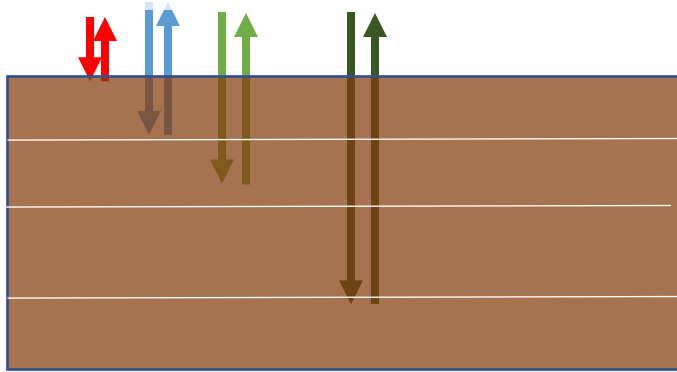
Key findings:

1. “Subsidence” in closure phase bias corresponds to soil moisture increase after irrigation.
2. subsidence magnitude depends on soil moisture increase rate and variation
3. Slower increase and greater variation result in larger subsidence



Modeling soil moisture from closure phase bias - Summary

MODEL

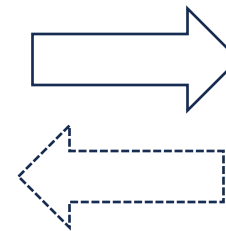
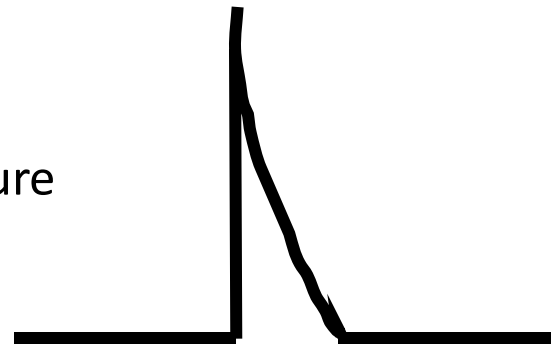


A discrete soil moisture model
soil moisture -> closure phase bias

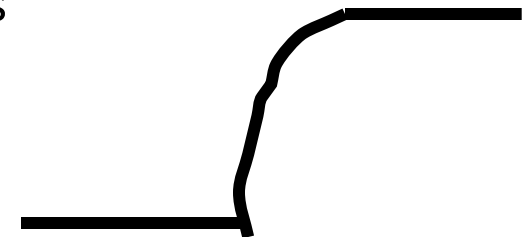
OBSERVATION



Soil
moisture



Bias



Modeling soil moisture from closure phase bias – Summary

Potentials

- Linking soil moisture variation to bias accumulation is straightforward relative to working directly with closure phase.
- It offers potential to detect drought onset, produce high-resolution soil moisture maps and scale with global time-series products

Limitations

- Limited temporal resolution (limited by SAR acquisition intervals).
- Inversion scheme has yet to be tested with extensive data.