

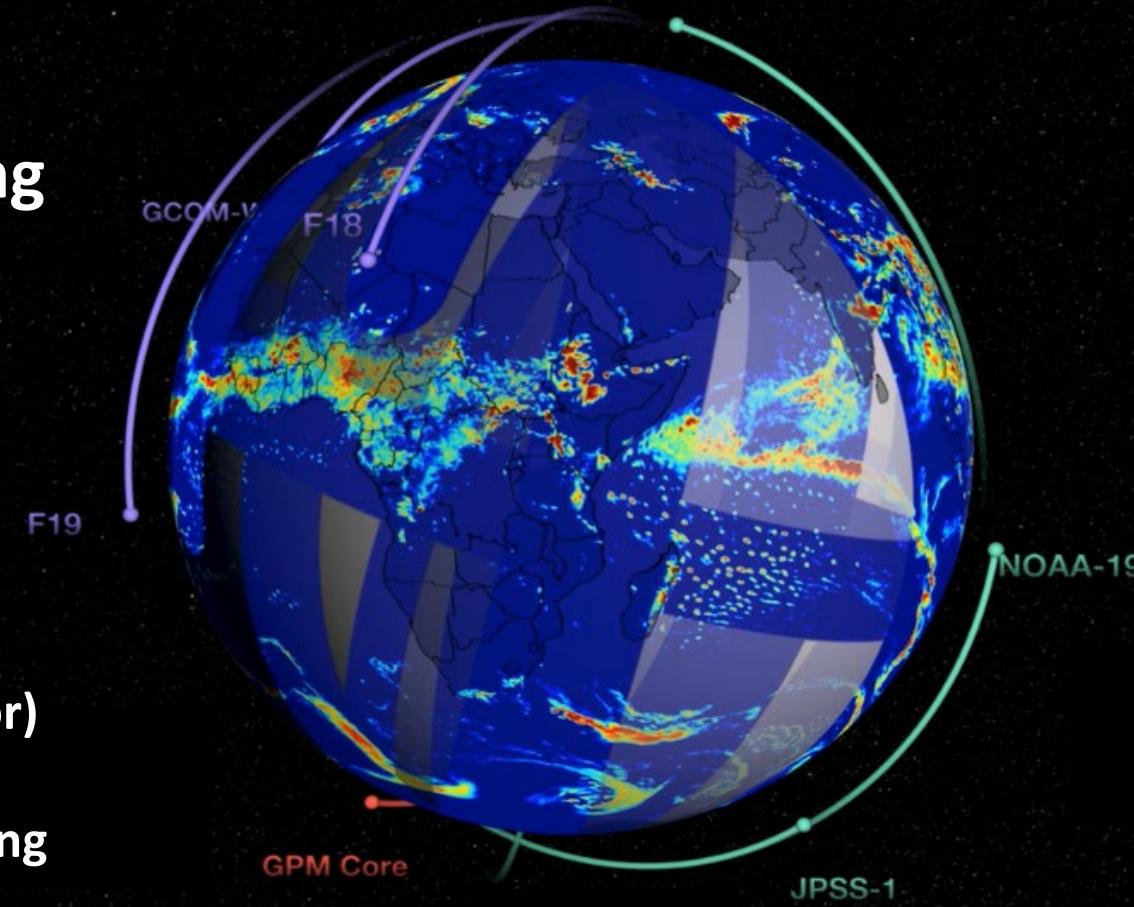
Large-Scale Benchmarking of a NRT Global Precipitation Ensemble: STREAM-Sat

Daniel Wright (Associate Professor)

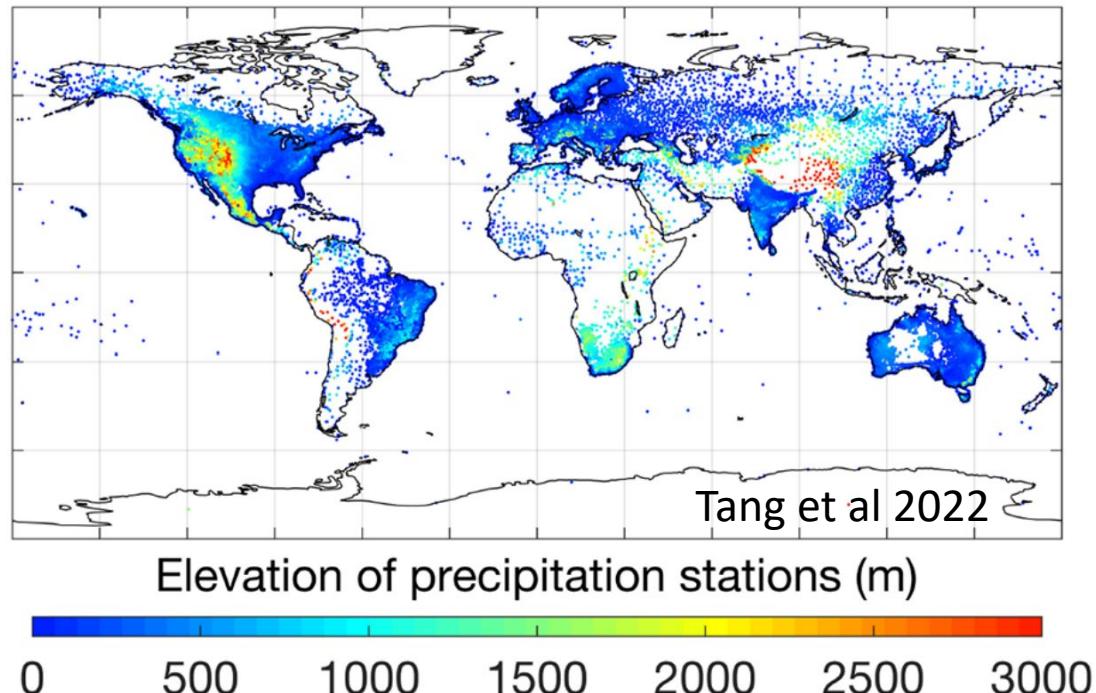
Kaidi Peng (PhD Student)

Civil and Environmental Engineering
Department

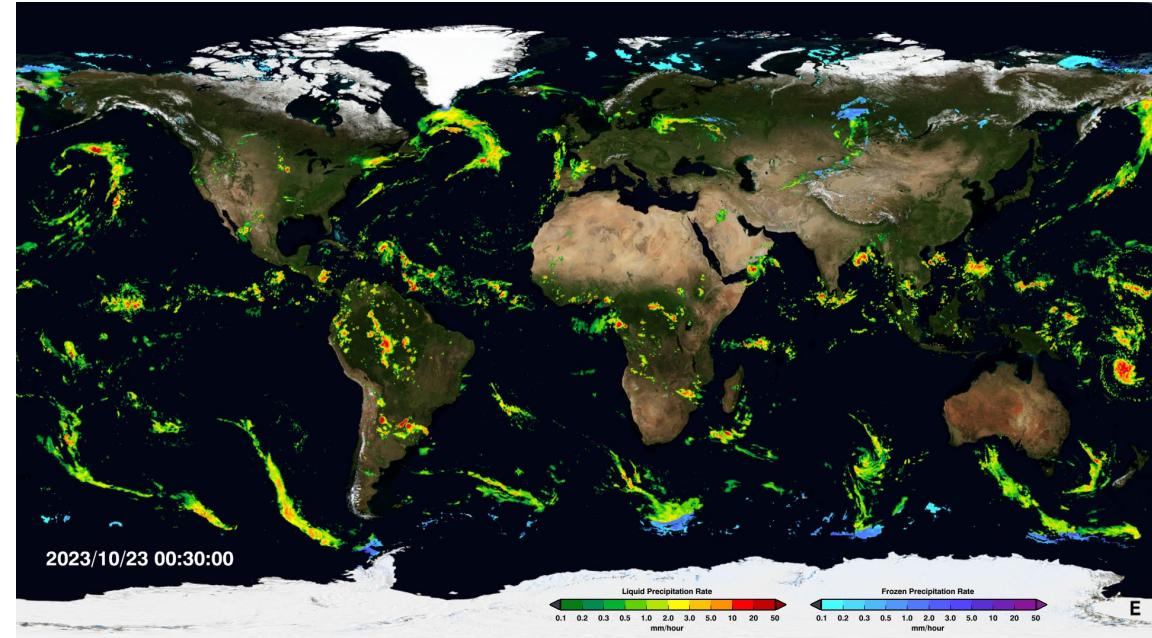
University of Wisconsin-Madison



Gauge network VS. Satellite



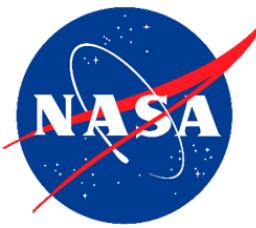
- Globally sparse
- Low availability & Long latency
- Usually high accuracy



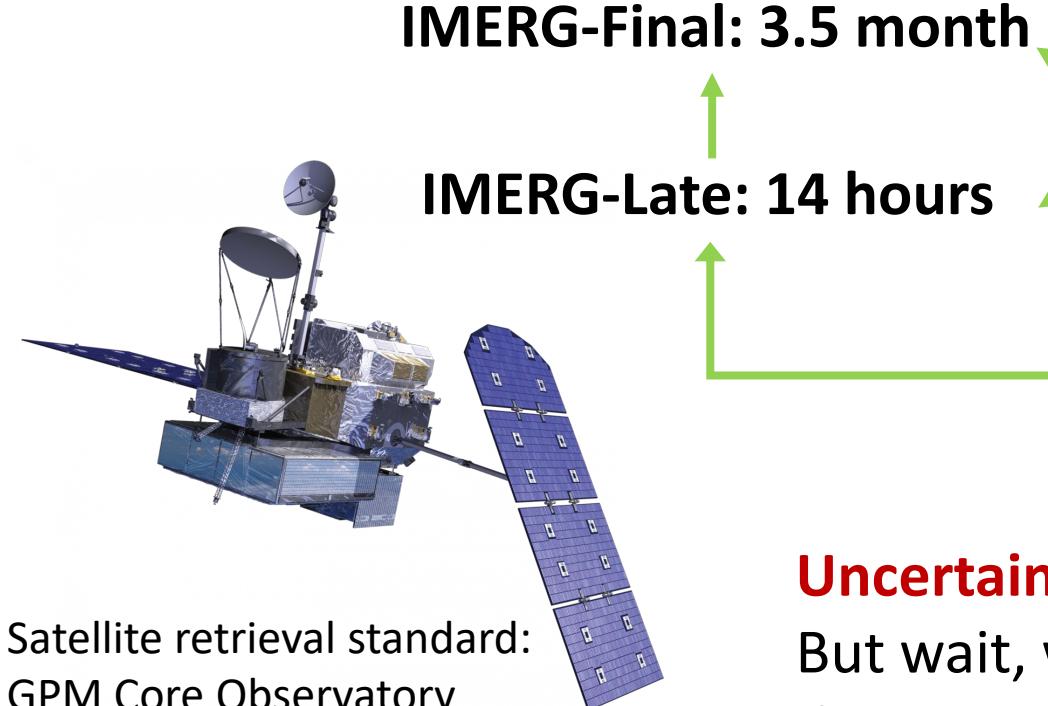
- Usually low accuracy
- High spatiotemporal resolution
- Global availability & Near Realtime

Others: Reanalysis (e.g., ERA5), Merged (e.g., MSWEP), Gauge-based (e.g., GPCP)
All has their own deficiency & strength

IMERG-Early



- The Integrated Multi-satellitE Retrievals for Global Precipitation Measurement (IMERG) to estimate precipitation globally



IMERG-Final: 3.5 month
IMERG-Late: 14 hours

Pros:

- Dates Covered: June 2000 - Present
- Spatial Resolution: ~10km / 0.1 Degree
- Minimum Latency: 4 hours

Cons:

- Accuracy: Overestimate and Underestimate



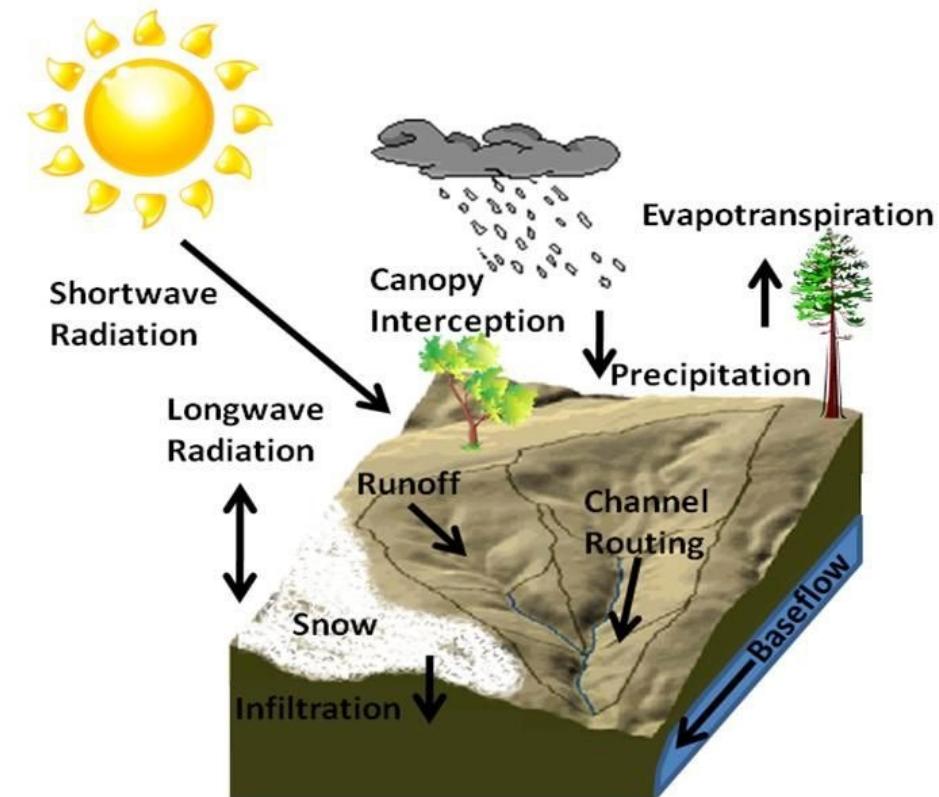
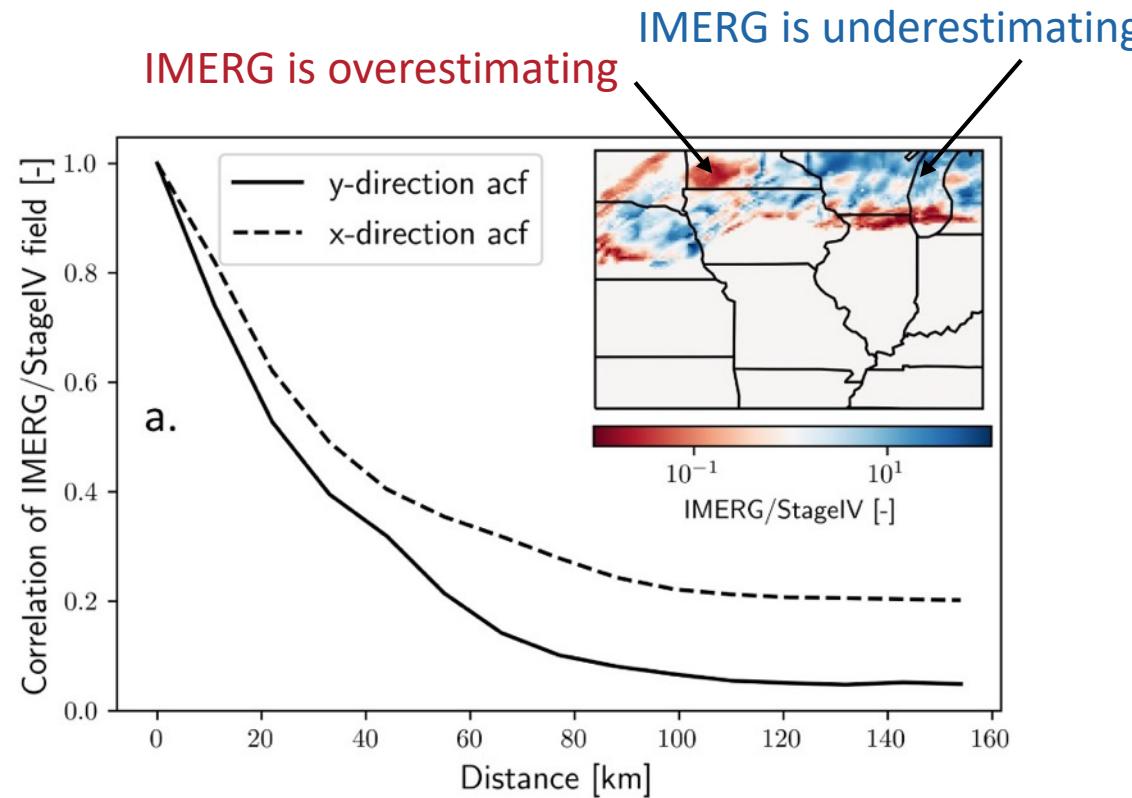
Uncertainty Estimation

But wait, what kind of attributes of IMERG error have been documented so far?

Multi-scale complexity... Magnitude & location dependent...

Satellite retrieval standard:
GPM Core Observatory
Satellite

IMERG Uncertainty Across Space/Time



Precipitation uncertainty is one of the main sources of uncertainty in hydrologic models

Properties that change over time (or space) = **nonstationary**

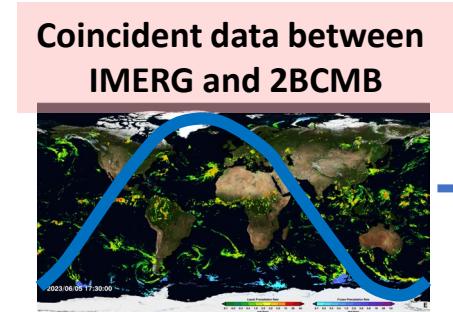
Properties that vary with direction = **anisotropic**

1. Grid scale error model
2. Space-time connection

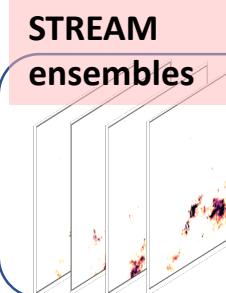
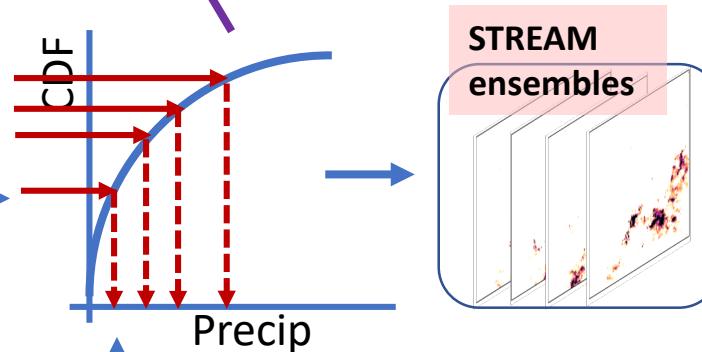
NRT Global Precipitation Ensemble: STREAM-Sat

1. Grid scale error model:

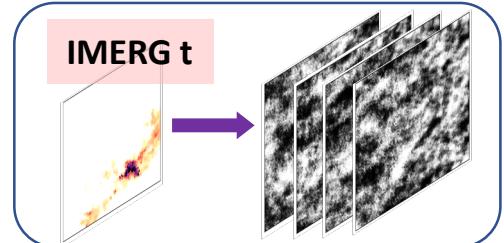
The distribution (CDF) of precipitation conditioned on observation



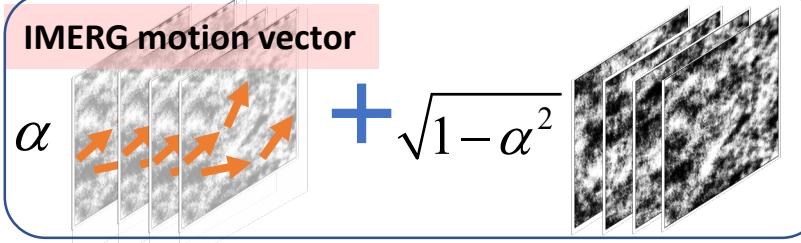
Train
CSGD



- ① SSFT to simulate noise that replicates spatial structure



$t \rightarrow t+1$



2. Space-time connection

No ground-based observations are used in STREAM-Sat



1. Grid scale error model

- The Censored Shifted Gamma Distribution (CSGD) error model

Fitting Distribution

$$\mu(t) = \frac{\mu}{\alpha_1} \log1p \left\{ \text{expm1}(\alpha_1) \left[\alpha_2 + \alpha_3 \frac{R_s(t)}{R_S} \right] \right\},$$

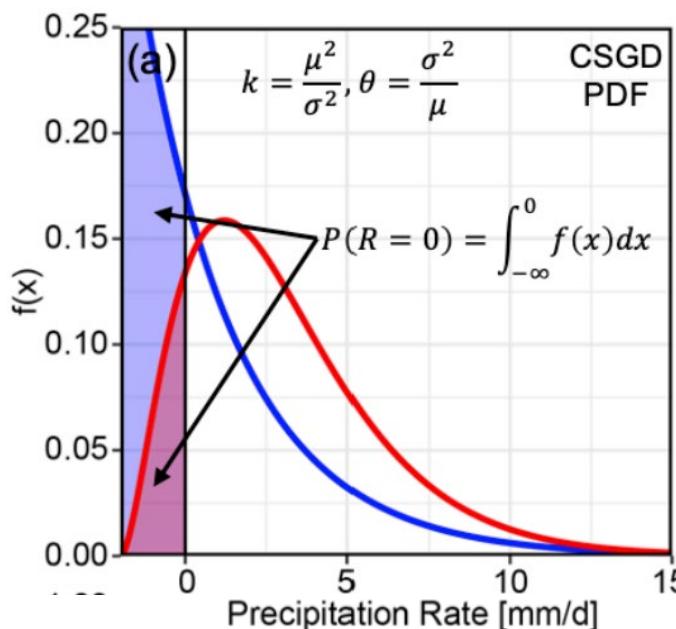
$$\sigma(t) = \alpha_4 \sigma \sqrt{\frac{\mu(t)}{\mu}},$$

$$\delta(t) = \delta,$$

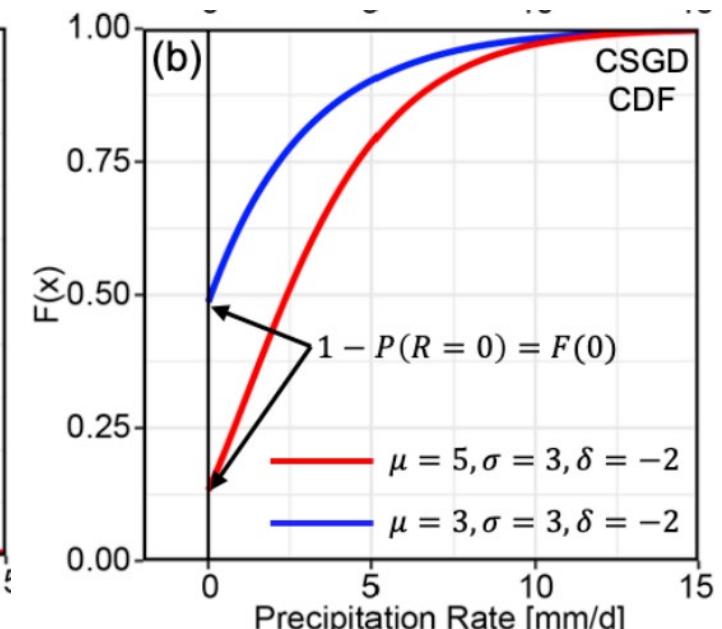
$$F_{\mu,\sigma,\delta}(x) = \begin{cases} F_{\mu,\sigma}(x - \delta), & \text{for } x \geq 0 \\ 0, & \text{for } x < 0 \end{cases}$$

Target Function

$$\text{CRPS}(F_{\mu(t),\sigma(t),\delta(t)}, o_t) = \int_{-\infty}^{\infty} \left[F_{\mu(t),\sigma(t),\delta(t)}(x) - \text{I}(o_t \leq x) \right]^2 dx$$



Probability Density Function

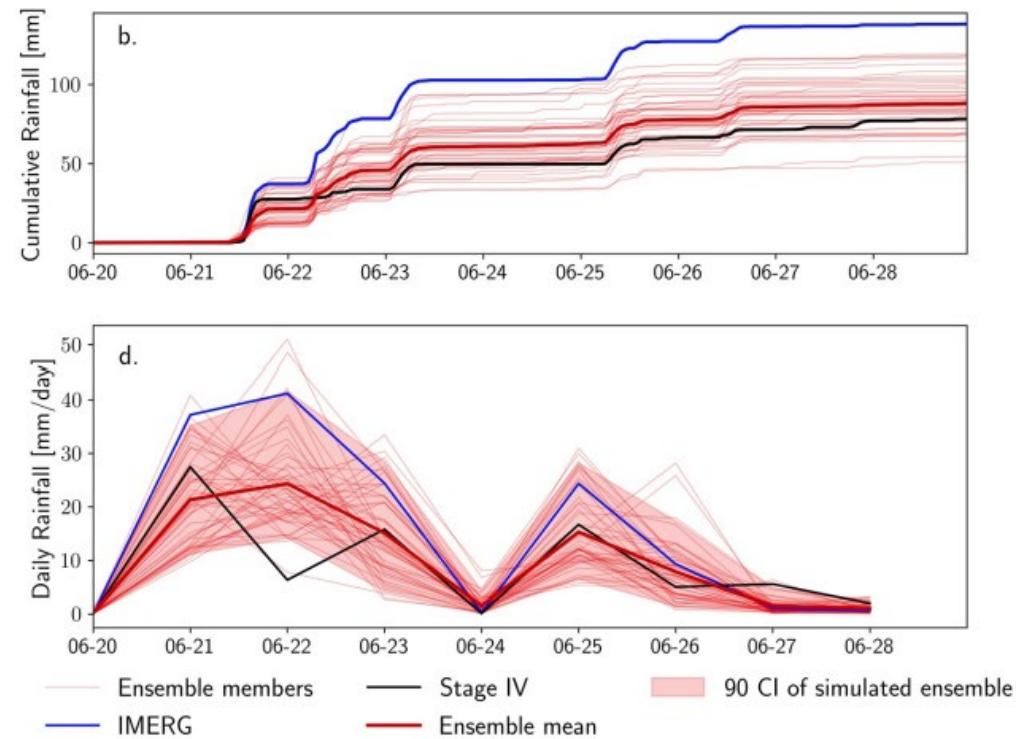
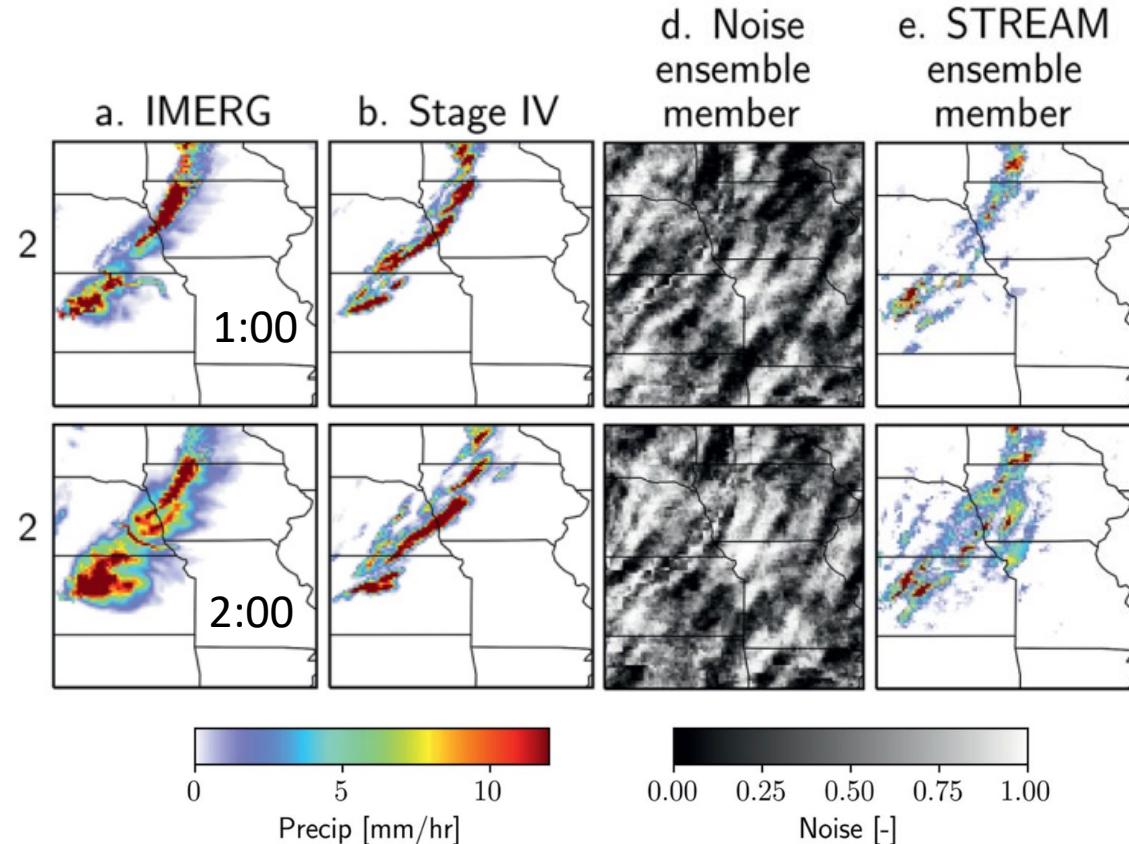


Cumulative Density Function

Real precipitation distribution conditioned on IMERG estimates

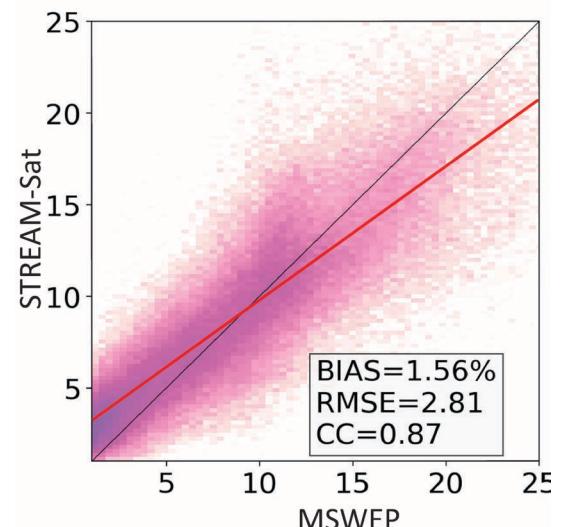
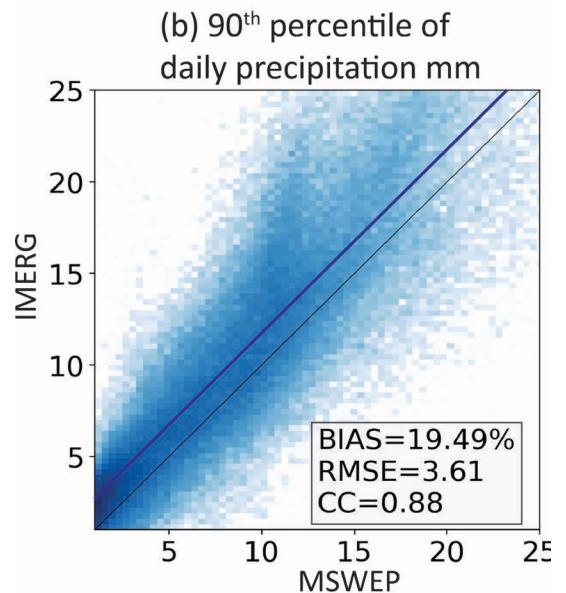
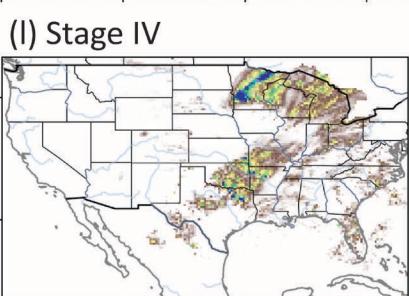
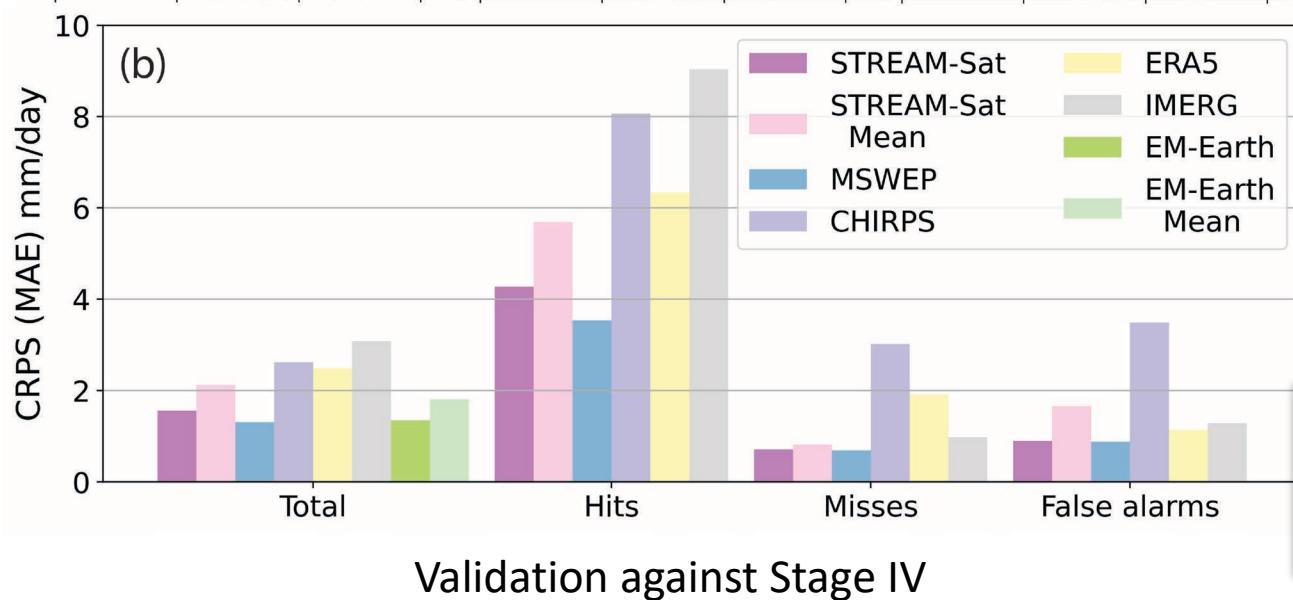
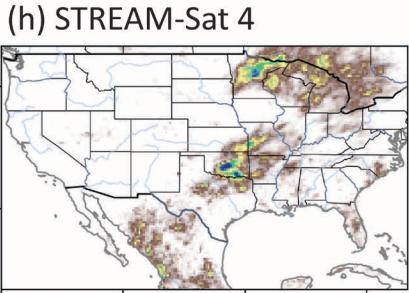
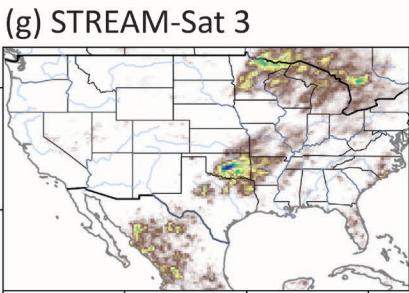
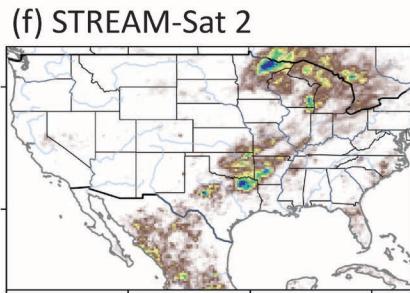
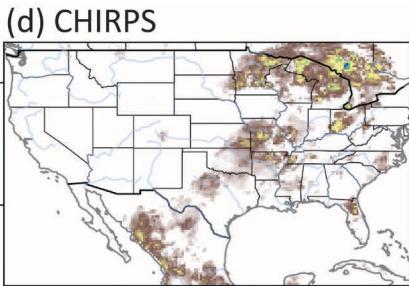
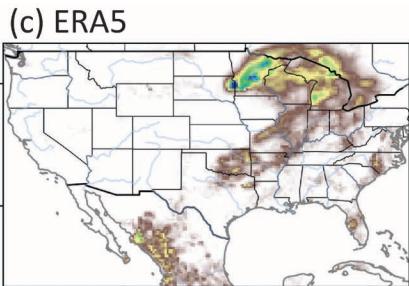
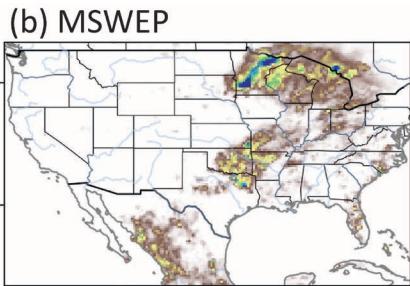
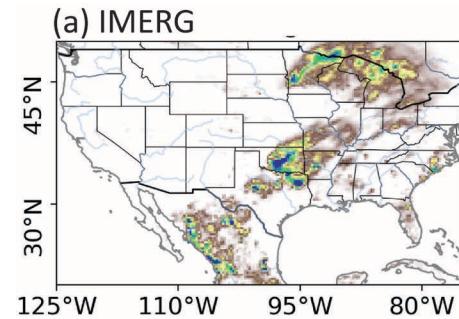
2. Space-time connection

- The Space-Time Rainfall Error and Autocorrelation Model (STREAM)

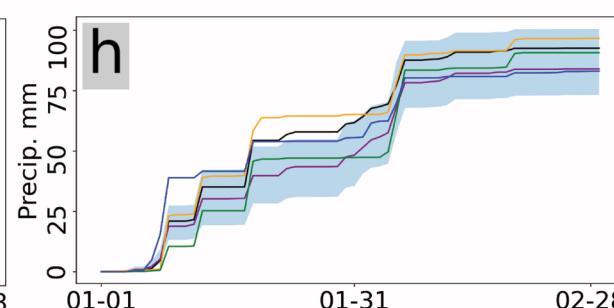
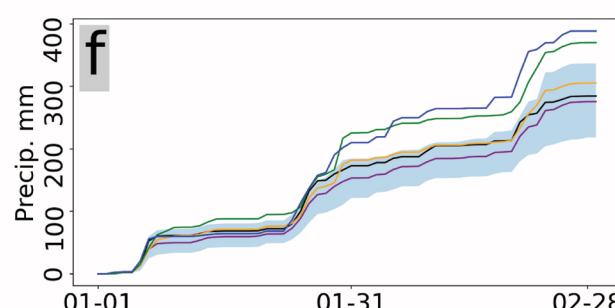
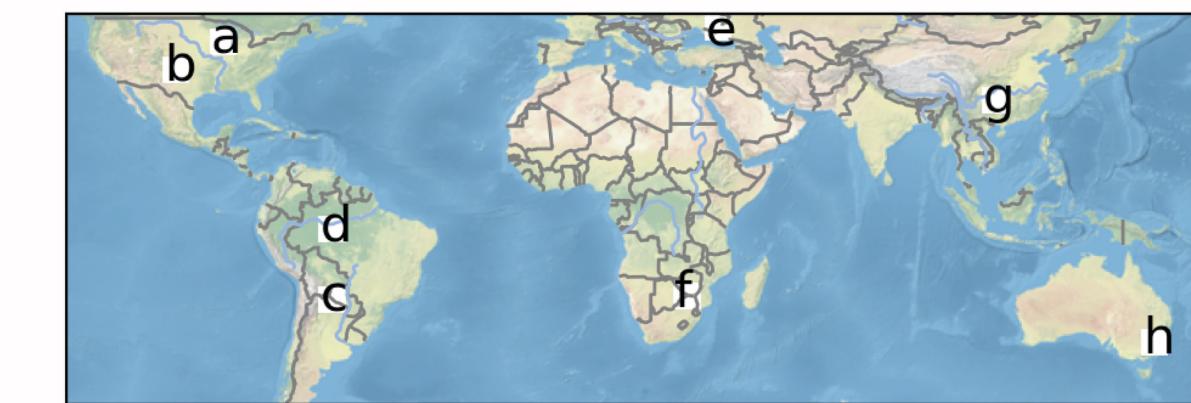
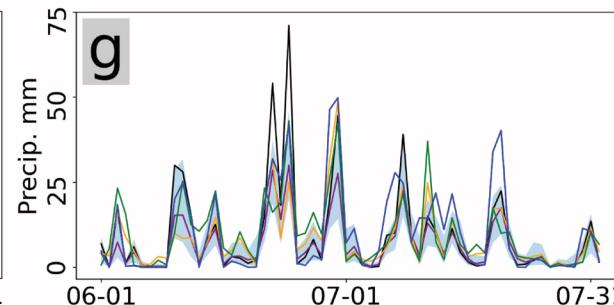
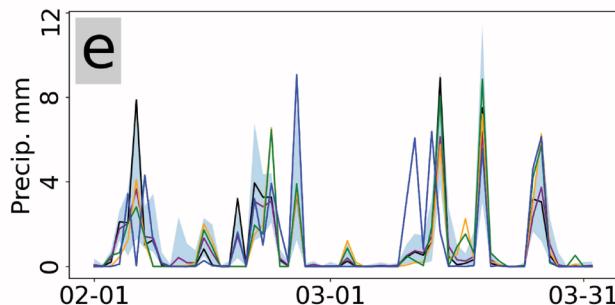
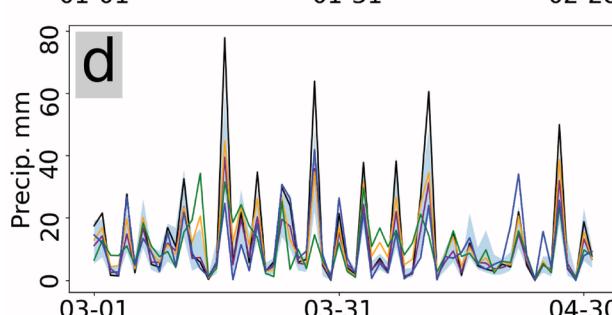
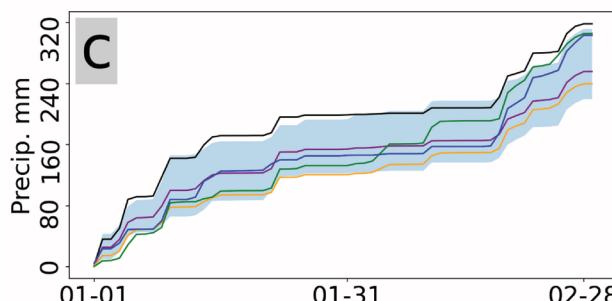
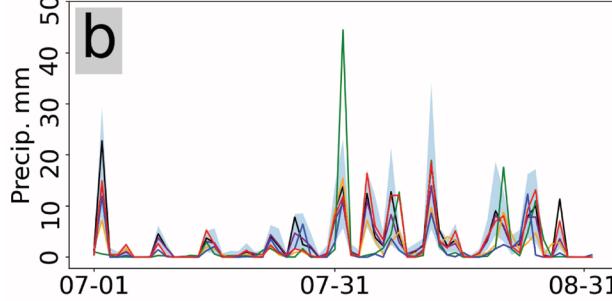
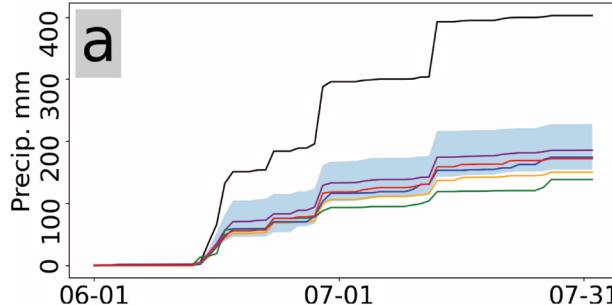


STREAM ensembles 1. replicate space & time autocorrelation; 2. resemble the real precipitation

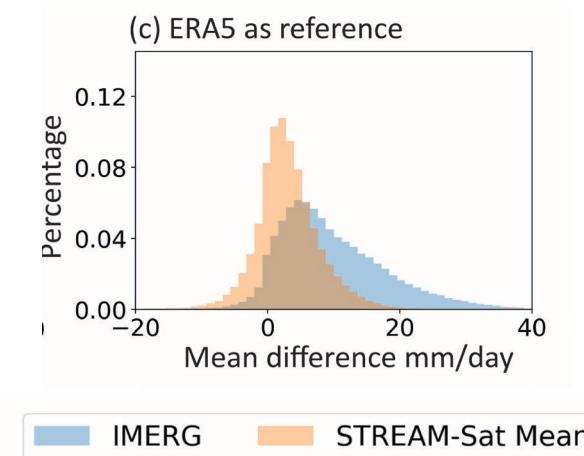
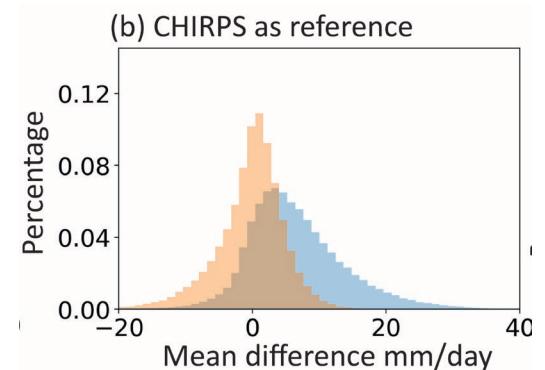
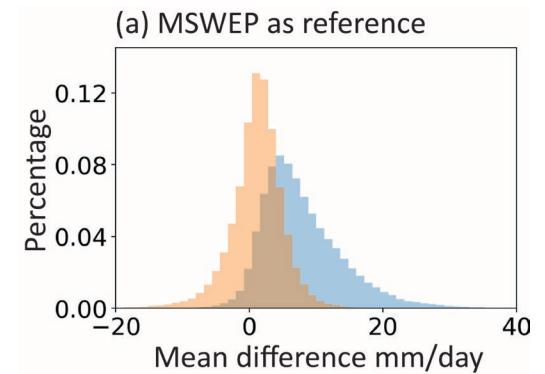
Evaluation: comparison with “peers”



- Great improvement compared to IMERG
- Its performance is close to, if not better than, its peers
- No gauge used



Precipitation time-series reported across the globe



90-quantile daily rainfall

Comparison with “peers”

| Name | Details | Resolution; coverage | Latency |
|-------------|--|--|--|
| IMERG Early | Satellite-based, Deterministic | 0.1°, half-hourly; global | 4 hours |
| MSWEP V2.8 | Merged, Gauge corrected, Deterministic | 0.1°, 3-hourly; global | available up to 2020 |
| CHIRPS V2.0 | Satellite-based, Gauge-corrected, Deterministic | 0.05°, daily; 50°S-50°N, land | The third week of the following month |
| ERA5 | Reanalysis-based, Deterministic | 0.25°, hourly; global | 5 days |
| EM-Earth | Reanalysis-based, Gauge-corrected, Ensemble | 0.1°, daily; global land except Antarctic | Available up to 2019 |
| STREAM-Sat | Satellite-based, Ensemble | 0.1°, half-hourly; 50°S-50°N, land | 4 hours |

STREAM-Sat Highlights:

- This method is to produce ensemble precipitation fields using only satellite data. It can be applied globally, in near-realtime, and at high resolution (here, 0.1° , half-hourly over land, inherited from IMERG).
- STREAM-Sat exhibits clear improvement in detection ability and intensity compared to IMERG Early.
- Compared to precipitation datasets relying on ground gauge network, STREAM-Sat shows consistent ensemble range and performance skills. STREAM-Sat could be more favorable in gauge-limited regions.
- While STREAM-Sat's accuracy (e.g., in terms of CRPS) is not superior in all respects over all other evaluation datasets, its unique combination of traits—i.e., it is probabilistic via its ensembles, it has low bias, it can be produced in near-realtime, and it has high resolution—gives it certain advantages over the other benchmark global precipitation datasets.
- The method is particularly valuable in ungauged regions with large and unknown meteorological uncertainties that merit ensemble approaches, while its near-realtime potential offers notable benefits in water-related disaster early warning.

Reference

Algorithm:

- Wright, D. B., Kirschbaum, D. B., & Yatheendradas, S. (2017). Satellite Precipitation Characterization, Error Modeling, and Error Correction Using Censored Shifted Gamma Distributions. *J Hydrometeorol*, Volume 18(No 10), 2801-2815. doi:10.1175/JHM-D-17-0060.1
- Hartke, S. H., Wright, D. B., Li, Z., Maggioni, V., Kirschbaum, D. B., & Khan, S. (2022). Ensemble Representation of Satellite Precipitation Uncertainty Using a Nonstationary, Anisotropic Autocorrelation Model. *Water Resources Research*, 58(8). doi:10.1029/2021wr031650
- Li, Z., Wright, D. B., Zhang, S. Q., Kirschbaum, D. B., & Hartke, S. H. (2020). Object-Based Comparison of Data-Driven and Physics-Driven Satellite Estimates of Extreme Rainfall. *Journal of Hydrometeorology*, 21(12), 2759-2776. doi:<https://doi.org/10.1175/JHM-D-20-0041.1>

Application:

- Hartke, S. H., Wright, D. B., Kirschbaum, D. B., Stanley, T. A., & Li, Z. (2020). Incorporation of Satellite Precipitation Uncertainty in a Landslide Hazard Nowcasting System. *Journal of Hydrometeorology*, 21(8), 1741-1759. doi:<https://doi.org/10.1175/JHM-D-19-0295.1>
- Hartke, S. H., Wright, D. B., Quintero, F., & Falck, A. S. (2023). Incorporating IMERG satellite precipitation uncertainty into seasonal and peak streamflow predictions using the Hillslope Link hydrological model. *Journal of Hydrology X*, 18. doi:10.1016/j.hydroa.2023.100148

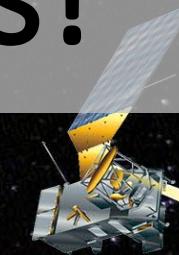


GPM Constellation

April 2019

Thanks!

NOAA 20
(NOAA)



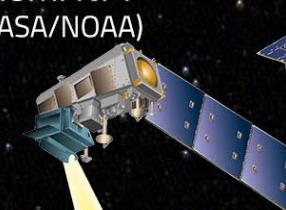
DMSP F17/F18
(DoD)



MetOp A/B/C
(EUMETSAT)



Suomi NPP
(NASA/NOAA)



GPM Core Observatory
(NASA/JAXA)

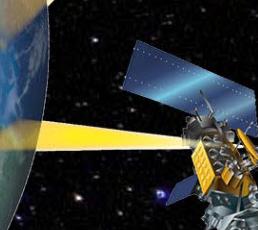
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