

Improving over land precipitation retrieval by brightness temperature temporal variation

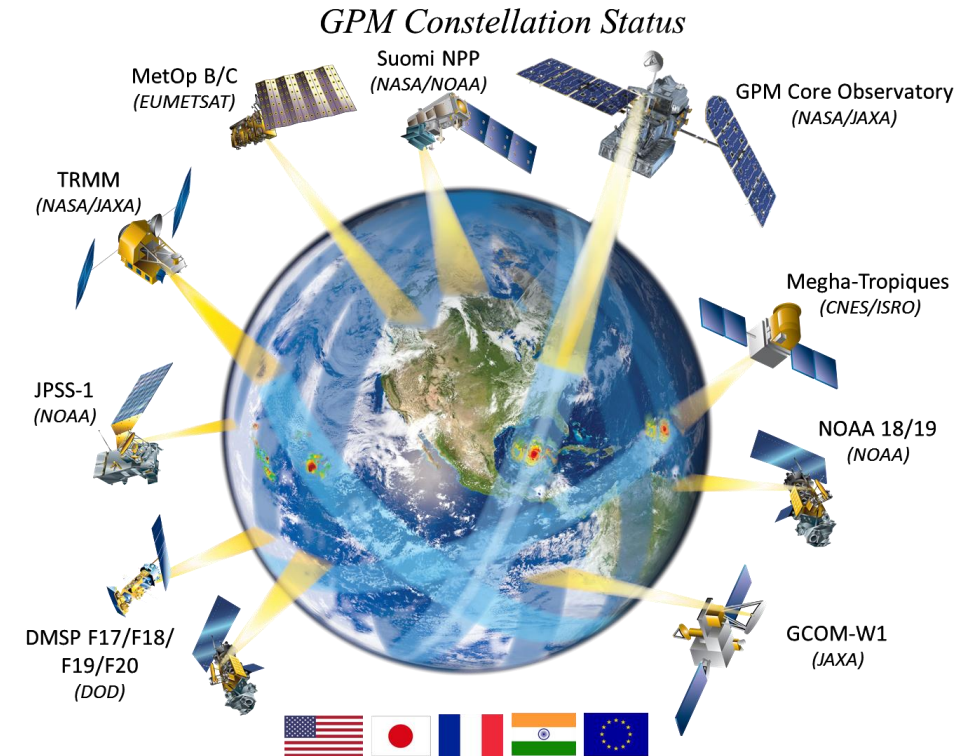
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Motivation:

- The primary precipitation signal over land is the brightness temperature (TB) depression at high frequency channels caused by the ice scattering.
- A common and serious issue is **the cold land surface contamination** (e.g., snow-covered land), which is particularly problematic for rainfall/snowfall retrieval in winter.
- To mitigate this problem, **this study proposes to use TB temporal variation (ΔTB)**, which is derived from eight polar-orbiting satellites, including **GPM, F17, F18, S-NPP, NOAA-18, NOAA-19, Metop-A and Metop-B**.
- MRMS precipitation data is taken as reference.



Basic idea: Make satellites “talk” to each other. When doing retrieval, one satellite should “consult” what happened previously from another satellite.

Why these 8 satellites:

GPM, F17, F18, S-NPP, NOAA-18, NOAA-19, Metop-A and Metop-B.

Because:

Radiometers onboard these satellites all have frequencies from ~89 GHz to ~183 GHz.

Satellite name	Radiometer name	Frequency	Frequency	Frequency	Resolution	ECT
GPM	GMI	89.0 (V/H)	166.0 (V/H)	183.3±3, ±7 (V)	6 or 7 km	—
<i>F17</i>	SSMIS	91.7 (V/H)	150.0 (H)	183.3±1, ±3, ±6.6 (H)	14 km	1826
<i>F18</i>	SSMIS	91.7 (V/H)	150.0 (H) ^a	183.3±1, ±3, ±6.6 (H)	14 km	1845
<i>SNPP</i>	ATMS*	88.2 (V)	165.5 (H)	183.3±1, ±1.8, ±3, ±4.5, ±7 (H)	14–45 km	1331
<i>NOAA-18</i>	MHS*	89.0 (V)	157.0 (V)	183.3±1, ±3 (H); 191.3 (V)	17–40 km	1833
<i>NOAA-19</i>	MHS*	89.0 (V)	157.0 (V)	183.3±1, ±3 (H); 191.3 (V)	17–40 km	1559
<i>MetOp-A</i>	MHS*	89.0 (V)	157.0 (V)	183.3±1, ±3 (H); 191.3 (V)	17–40 km	2129
<i>MetOp-B</i>	MHS*	89.0 (V)	157.0 (V)	183.3±1, ±3 (H); 191.3 (V)	17–40 km	2132

Previous work showed that high freq. channels provide most information for the light precipitation detection/retrieval.

You, Y., N. Wang, R. Ferraro, and S. Rudlosky, 2017: Quantifying the snowfall detection performance of the Global Precipitation Measurement (GPM) microwave imager channels over land. J. Hydrometeor., doi: <http://dx.doi.org/10.1175/JHM-D-16-0190.1>

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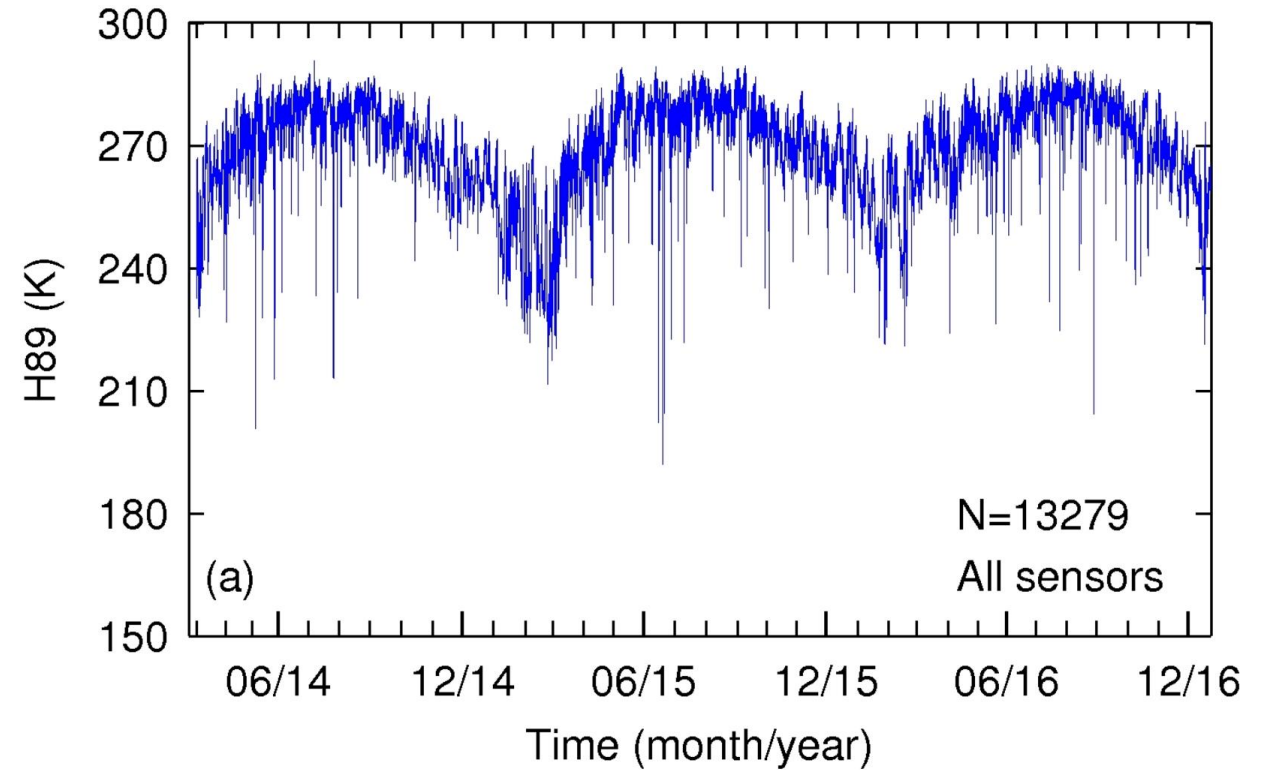
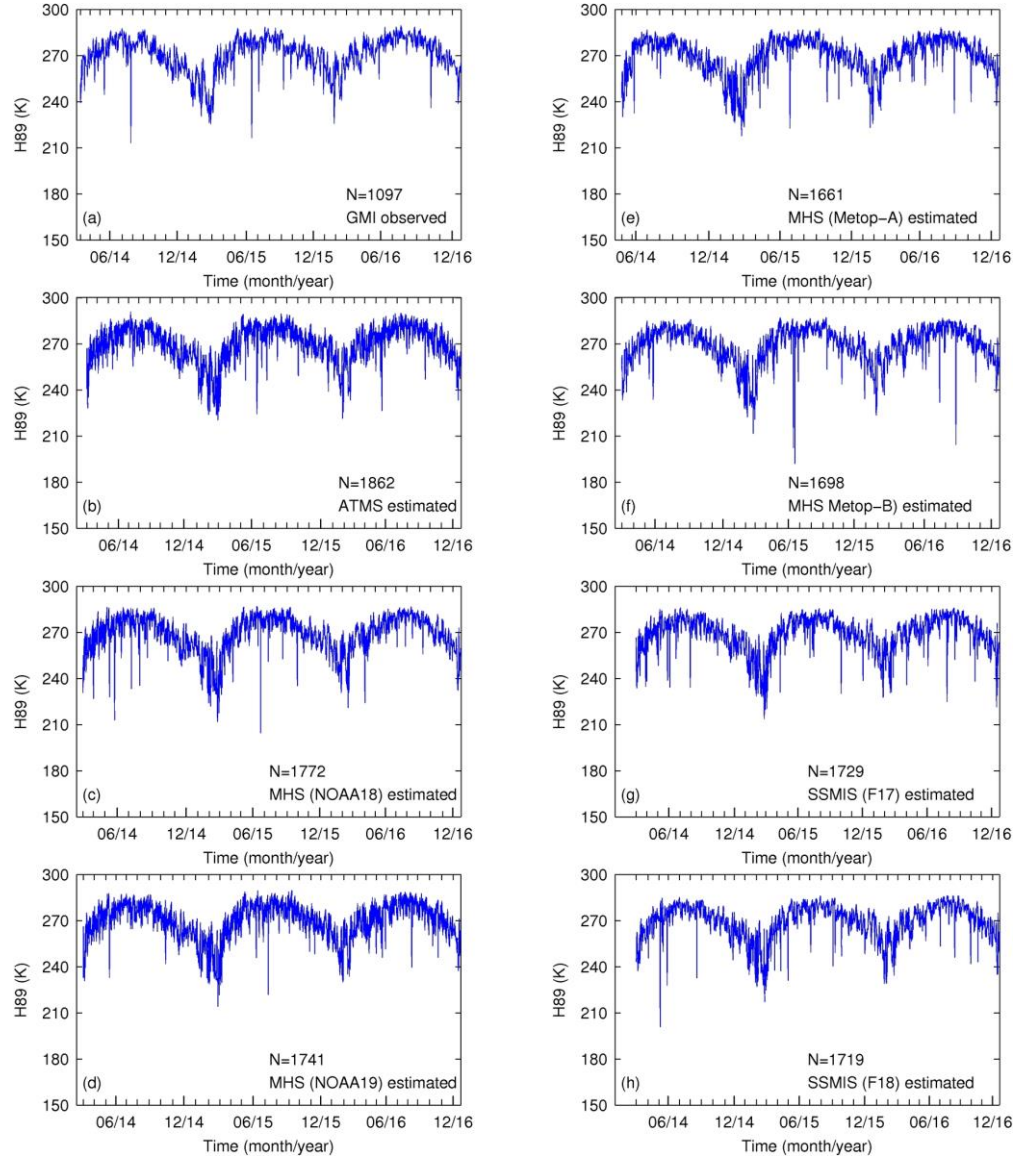
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“Convert” TBs from other radiometers to GMI

- **Simultaneous Conical Overpass (SCO)**: simultaneous measurements at a location from two different sensors at a similar frequency should be highly correlated.
 - (1) regress SSMIS(F17)-V91.7 against GMI-V89, using SCO pairs between F17 & GPM.
 - (2) apply this relation to all pixels.
- **Sounder (ATMS & MHS)**: SCO pairs dependent on scan position.
- By doing so, it is as if that we have eight sensors measuring TBs at GMI frequencies, which are 89.0 (V/H) 166.0 (V/H), 183.3 ± 3 (V), and 183.3 ± 7 (V).

You, Y., C. Peters-Lidard, J. Turk, S. Ringerud, and S. Yang, 2017: Improving over land precipitation retrieval with brightness temperature temporal variation. *J. Hydrometeor.* doi:10.1175/JHM-D-17-0050.1.

“Convert” TBs from other radiometers to GMI



- There are no obvious outliers observed when pooling data from all eight sensors together
- Our method can effectively convert TBs from other sensors to GMI channels.

TB temporal variation (ΔTB):

$$\Delta TB = TB_{t_0} - TB_{t_{-1}}$$

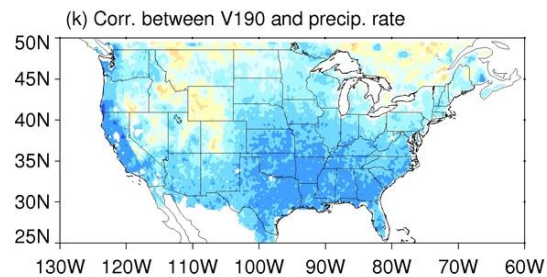
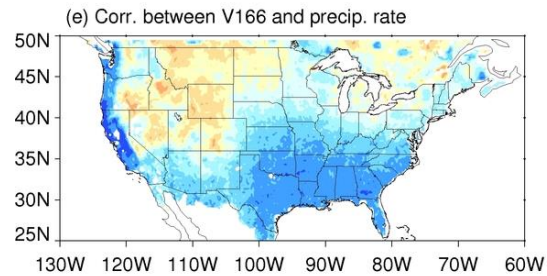
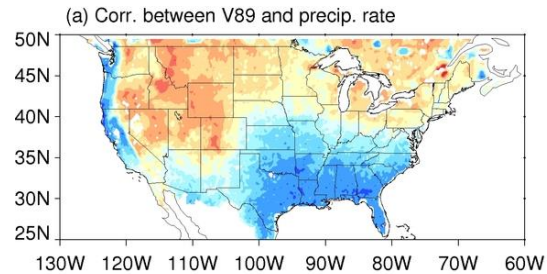
- TB_{t_0} is the current TB associated with precipitation.
- $TB_{t_{-1}}$ is the preceding TB at the same location without precipitation.
- ΔTB is not the difference between two temporally consecutive TB observations

$$\Delta t = t_0 - t_{-1}$$

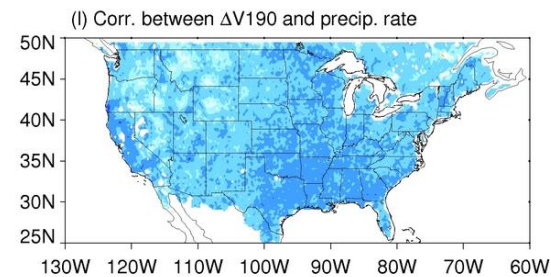
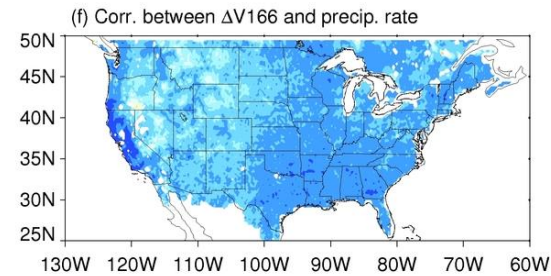
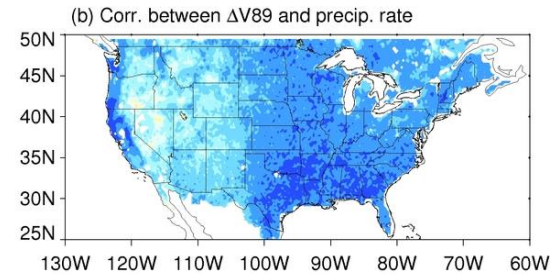
- Δt time difference between these two observations.
- The shorter the revisit time (Δt), the better the correlation between TB temporal variation (ΔTB) and precipitation intensity is.

Correlation from TB vs. from Δ TB

Corr. from TB

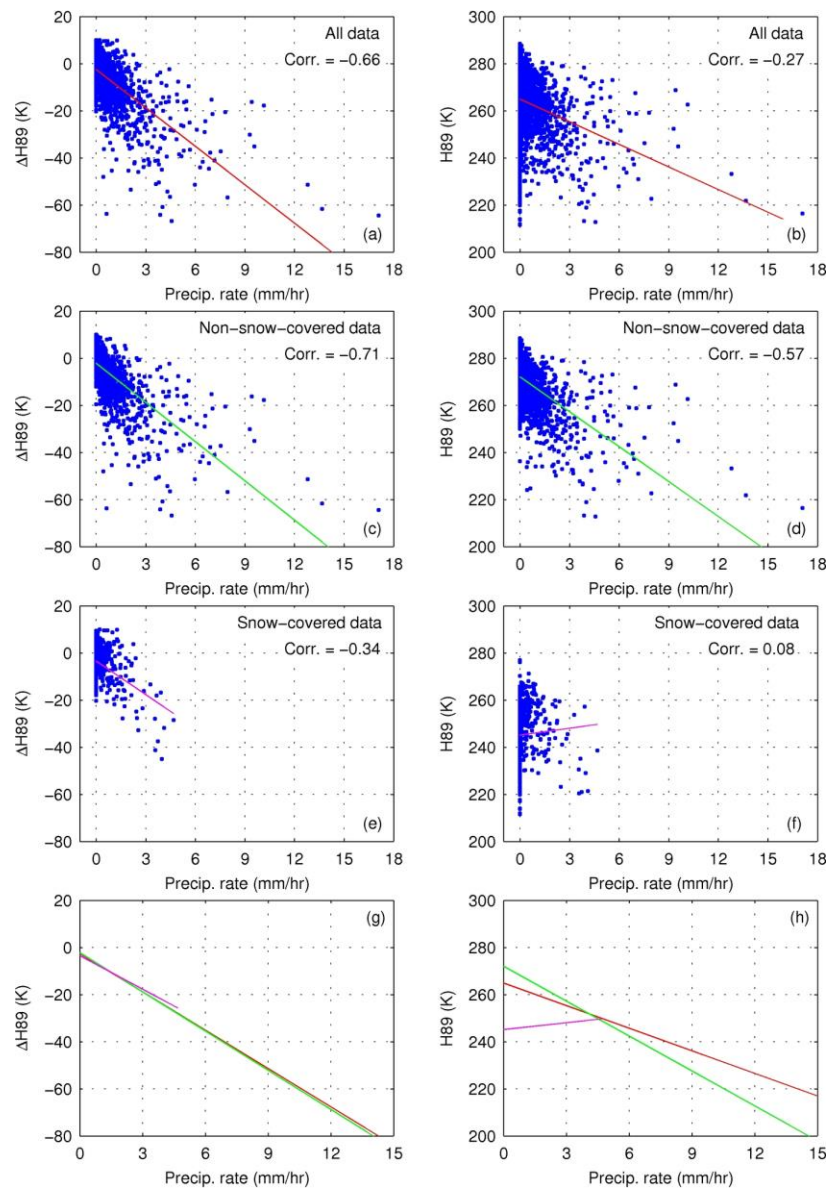


Corr. from Δ TB



- Δ TB correlates more strongly with precipitation rate.
- Especially, over Rocky Mountains and northeast CONUS.

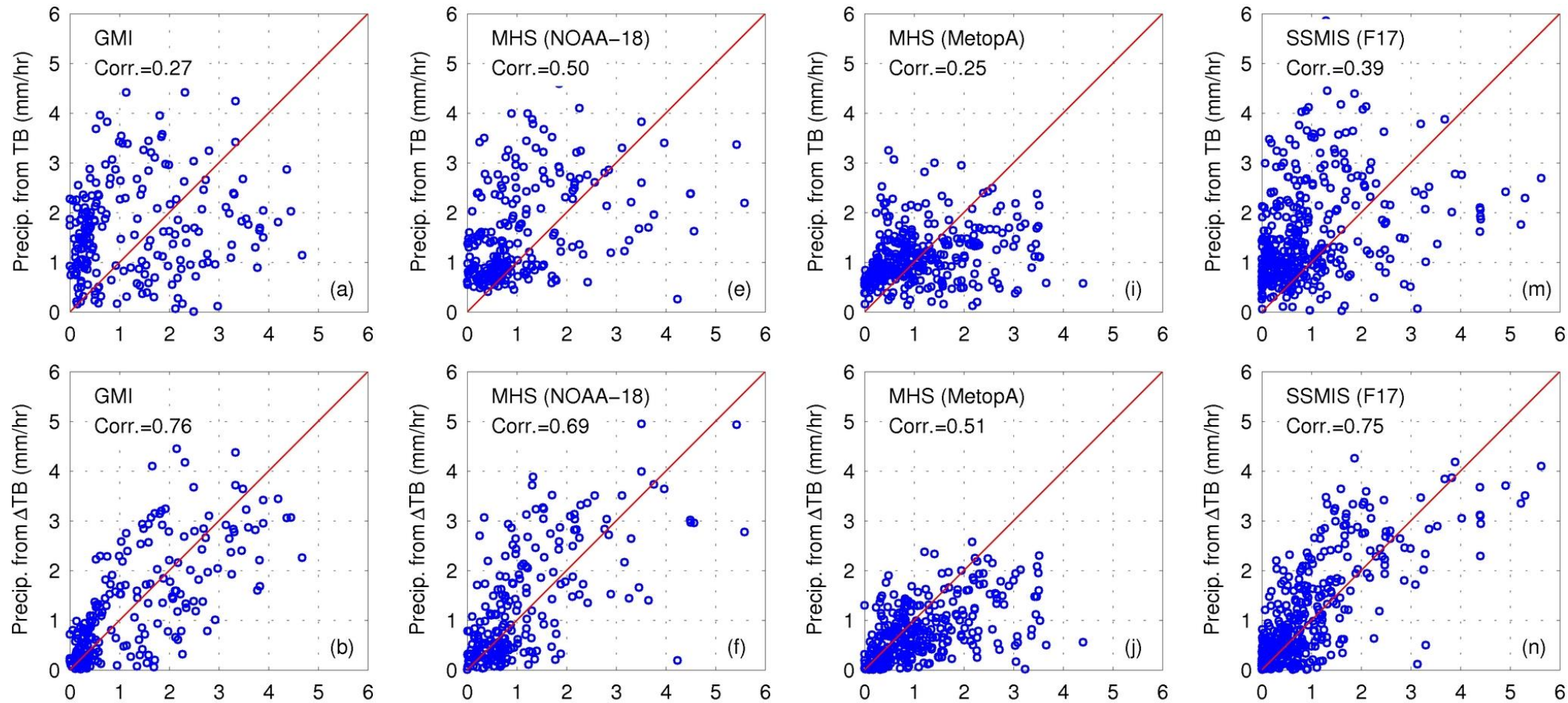
Why ΔTB is better



Because:

Surface contamination from snow cover is greatly mitigated by ΔTB due to the frequent re-visit (every 2hr) from these eight satellites.

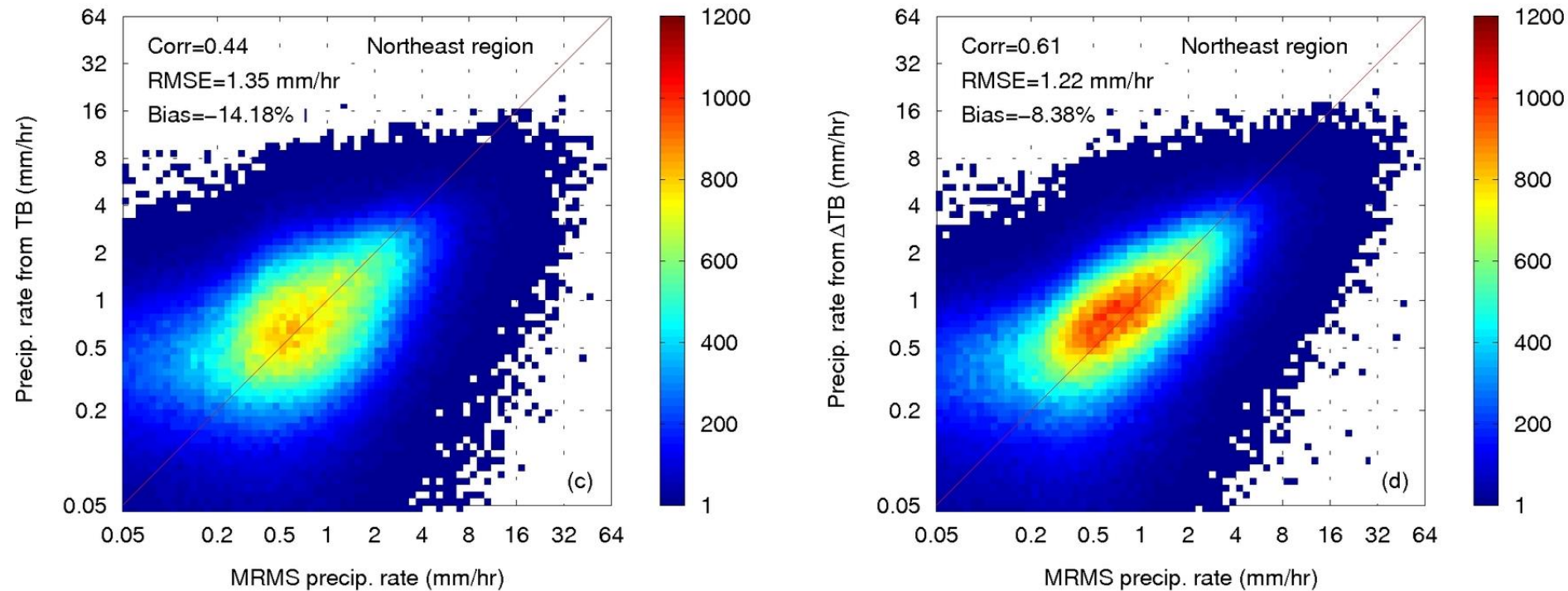
A blizzard case: Mid-Atlantic and Northeast on 23 Jan 2016



- **Over-estimation for light precipitation (<2 mm/hr) is greatly alleviated for all sensors.**
- **Because: if ΔTB is 0, the retrieval can only be 0.**

Retrieval results:

- Using all channels from 89 to 183 GHz.
- Simple linear regression (2014-2015 training; 2016 validation).



- Largest improvement is at the lower end of the precipitation intensity.

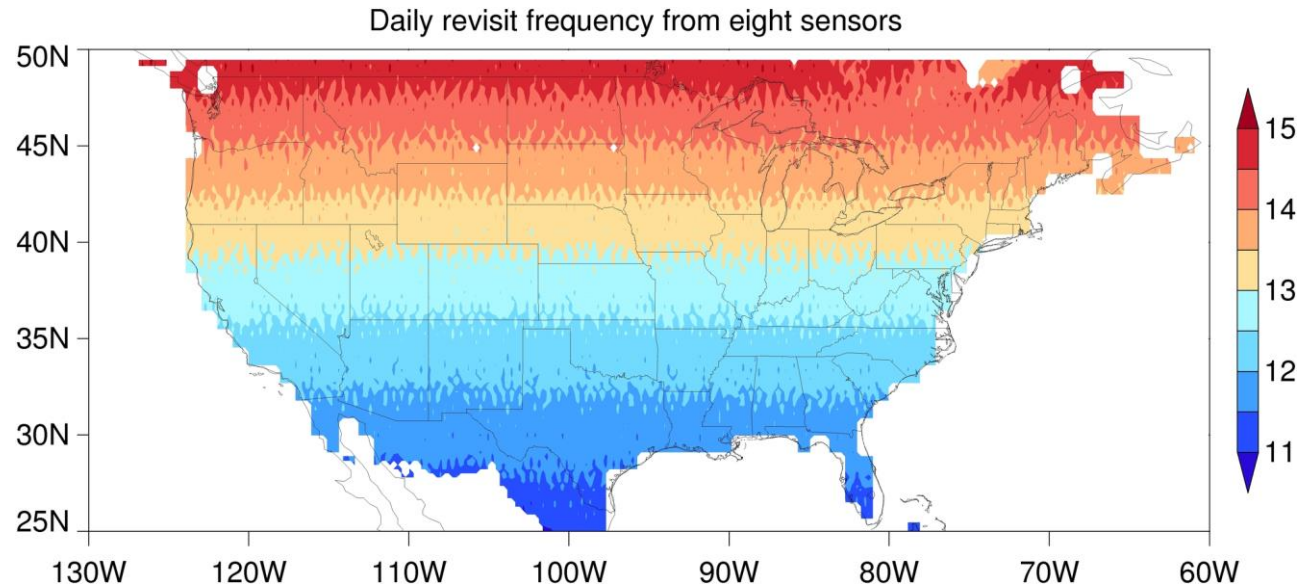
Summary:

- Δ TB correlates more strongly with precipitation rate than TB itself.
- Δ TB greatly mitigates snow-cover contamination.
- Largest improvement is found for the cold season precipitation.

Acknowledgement:

- Dr. Wesley Berg for SSMIS data status.
- Dr. Joe Munchak for reading/discussing the original draft.
- Dr. Eric Nelkin for providing the equatorial crossing time.

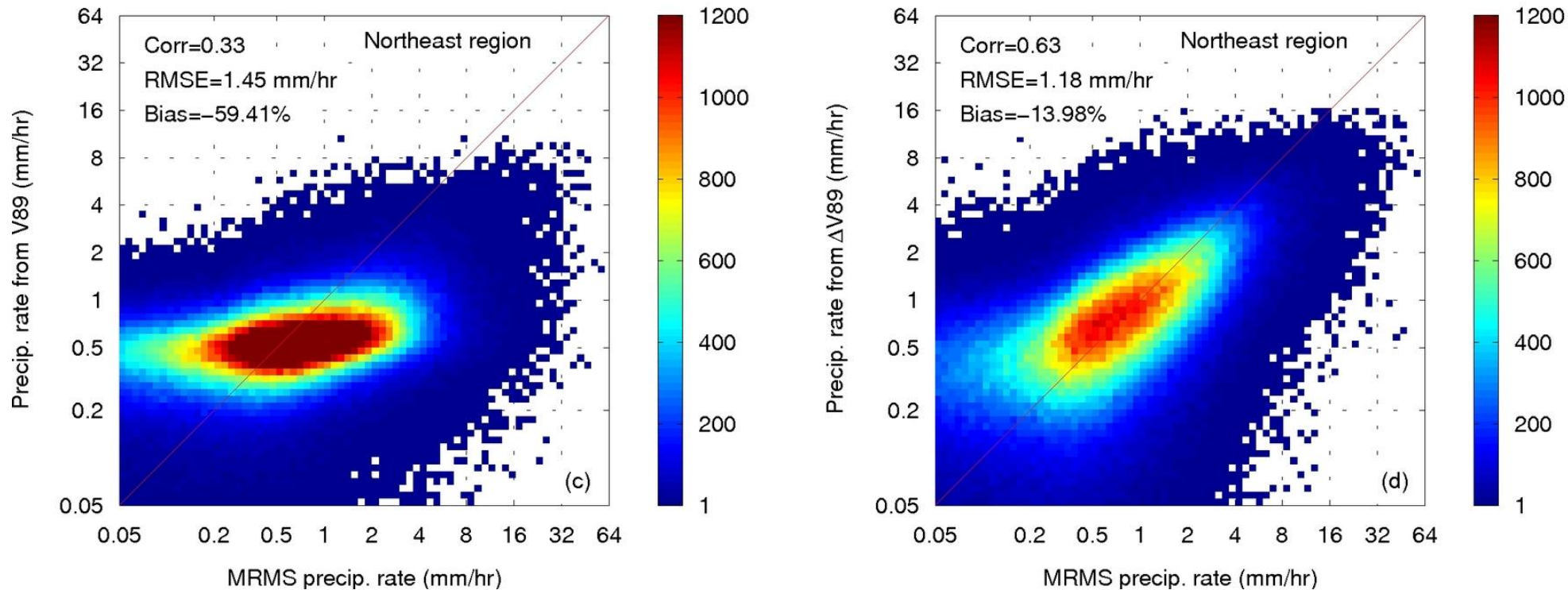
Daily re-visit frequency



- Over the targeted region, the daily re-visit frequency is from 10 to 16 times (on average, every ~ 2 -hr), depending on latitudes.
- The frequent revisit for a certain location enables us to find a non-cloudy background accurately, therefore ΔTB mostly contains the precipitation information, instead of surface and environmental contamination.

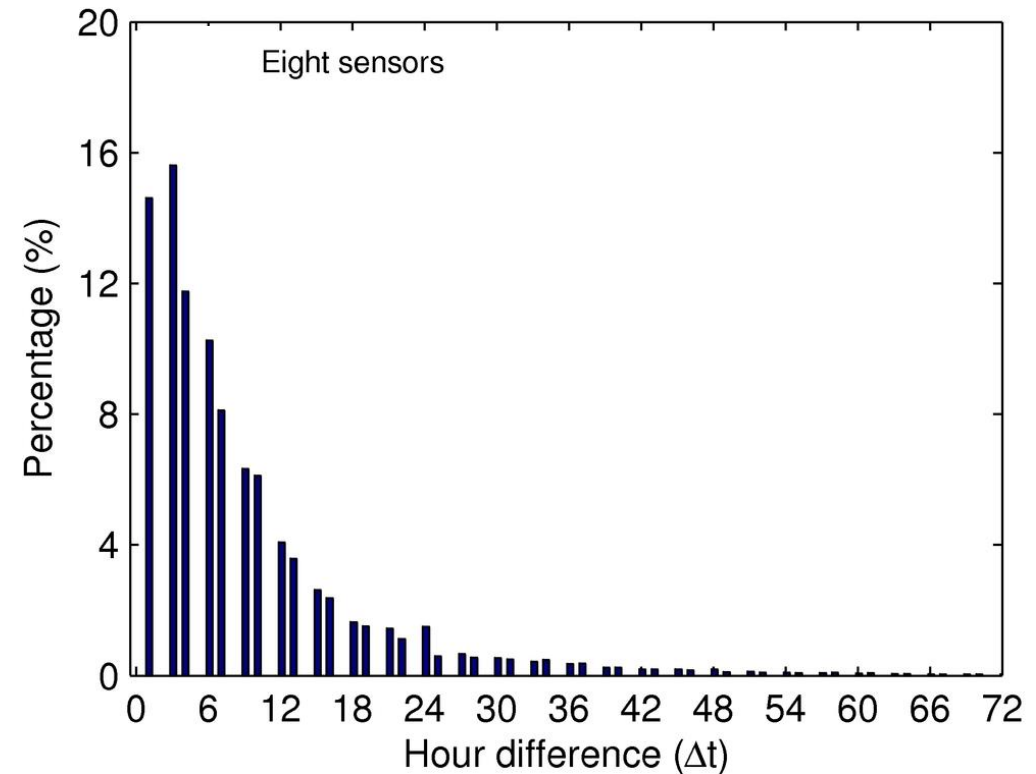
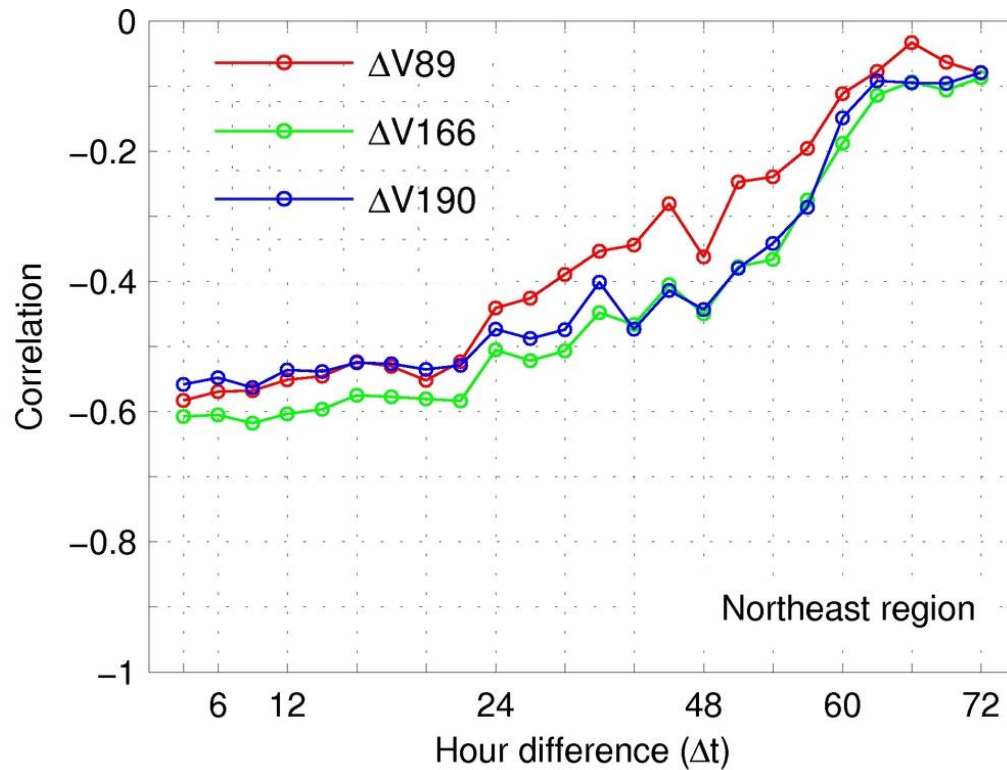
Retrieval results:

- Only using 89 GHz channel, AMSR2-type sensor (highest freq. ~ 89 GHz).
- Simple linear regression (2014-2015 training; 2016 validation).



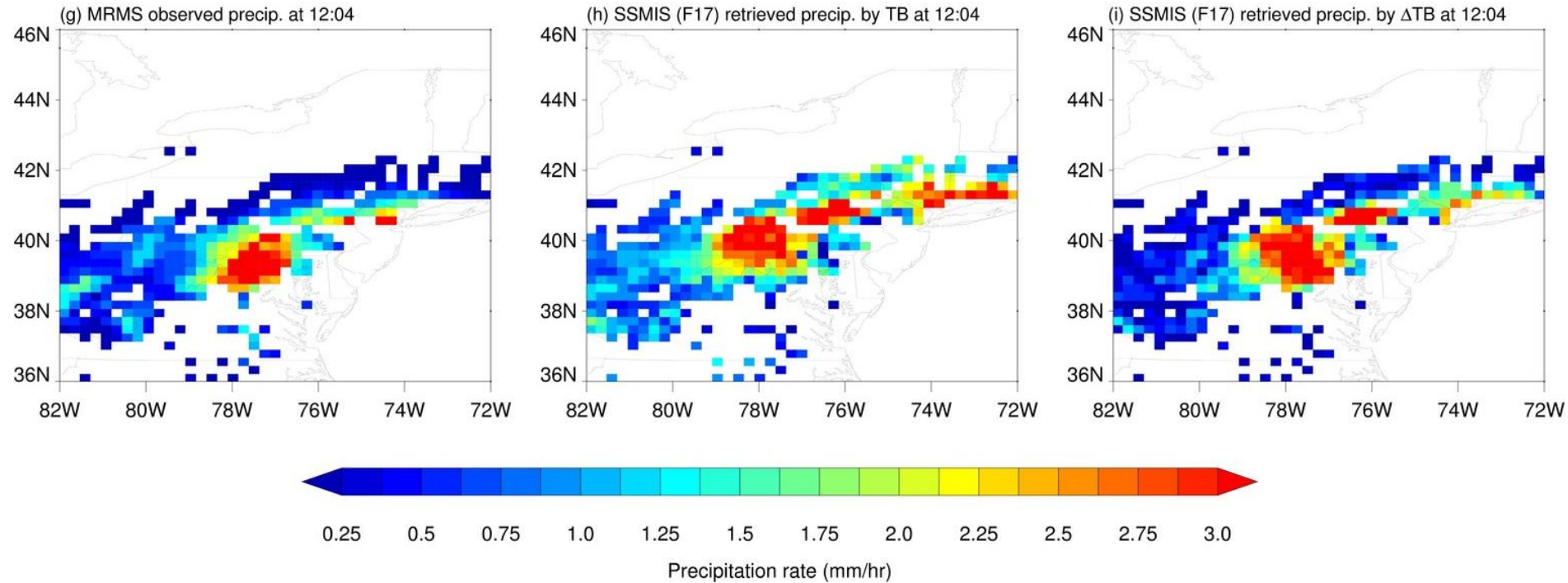
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Shorter ΔT , better correlation



- Correlation substantially weakens with longer ΔT over the Northeast region (37-47N, 65-80W).
- With these 8 satellites, almost all ΔT (94.6%) is less than 24 hrs.

A blizzard case: Mid-Atlantic and Northeast on 23 Jan 2016



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- We show later that the shorter the revisit time (Δt), the better the correlation between TB temporal variation (ΔTB) and precipitation intensity is.