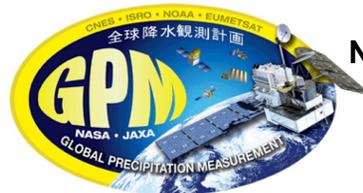


Precipitation top heights estimation in convective systems

Raúl Moreno (UCLM),
Francisco J. Tapiador (UCLM) and
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*The authors acknowledge support from projects
CGL2013-48367-P, CGL2016-80609-R,
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**NASA Precipitation Measurement Missions (PMM)
2018 PMM Science Team Meeting
Phoenix, Arizona, Oct 8-13, 2018**

Outline

- Introduction
- Correlation analysis
- Precipitation detection
- Precipitation estimation
- Conclusions

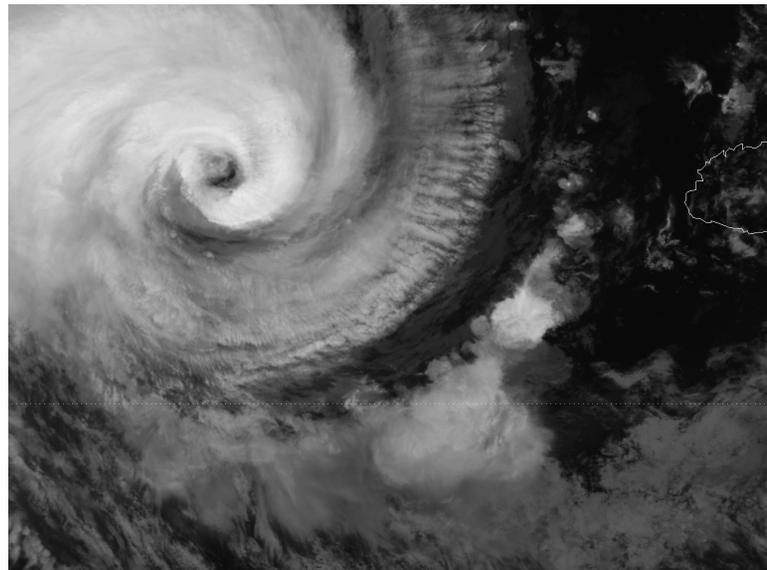
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Introduction

Convective storms

- Deep convective storms are one of the main causes of uncertainties in weather
- Difficult to measure

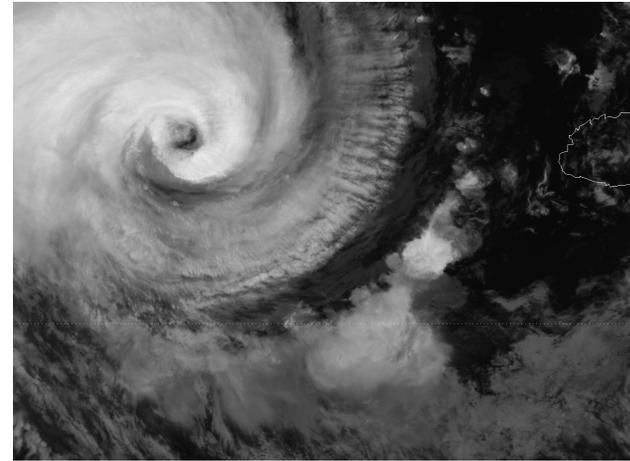


IR image of Tropical Cyclone Winston obtained from <http://eumetsat.int>

Introduction

Data sources for convective storms

- IR images
 - A lot of data is available
 - Poor sensitivity below convections
- Radar measurements
 - High sensitivity to convections
 - Limited number of coincidences
- Radiometers
 - Huge amounts of data
 - Measurements are not directly applicable (or are they?)



Outline

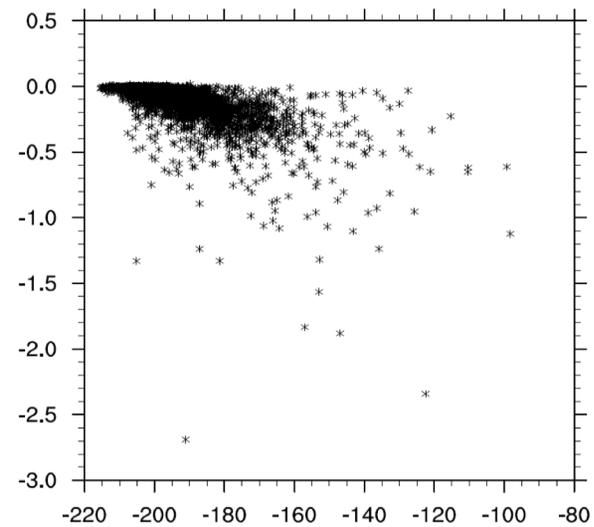
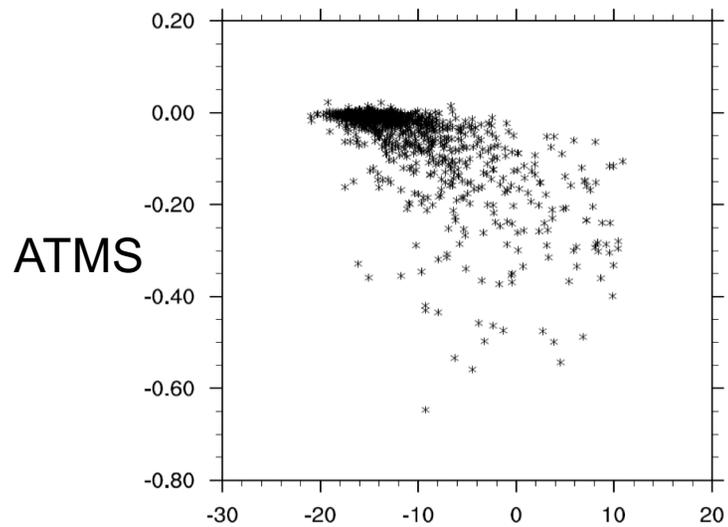
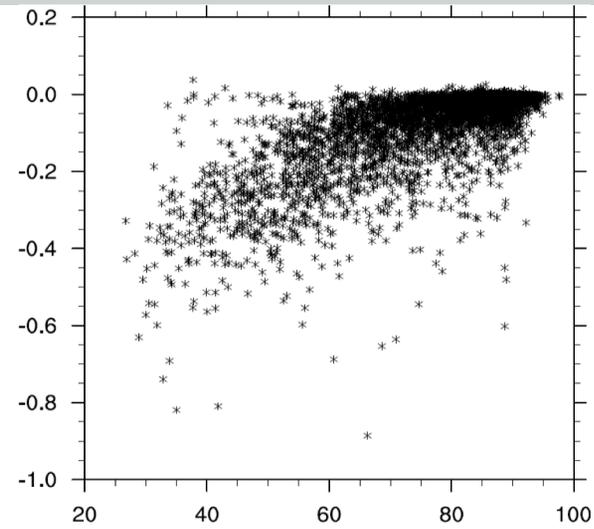
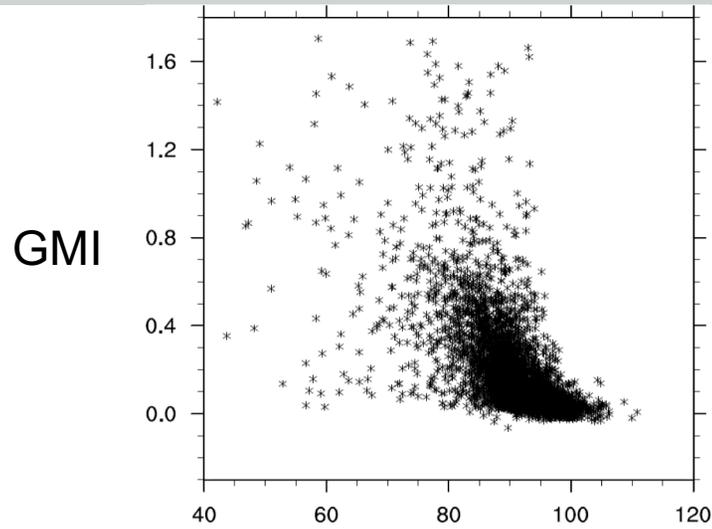
- Introduction
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Correlation analysis

- Is it possible to infer water content W^{52} features from brightness temperatures T^6 ?
- Difficult to explain 52 dimension of W with only 6 of T
- We reduced the dimensionality using Principal Components Analysis (PCA) and conserved the most significant Principal Components (PC)
 - Linearly compared them in the next slide

Correlation analysis

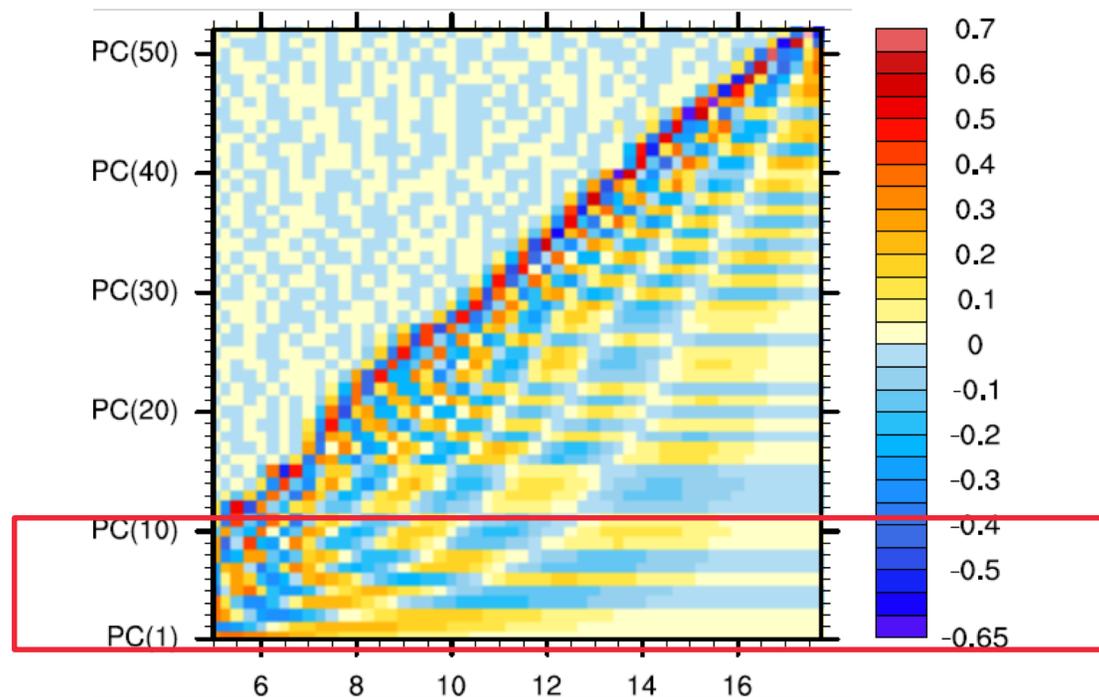
Correlation coefficients



Correlation analysis

Water content eigenvectors

- In the first 10 eigenvectors $PC(W)$ the weights for high precipitation is near zero
- High precipitation information is **lost** in the process



Correlation analysis

Top height correlation

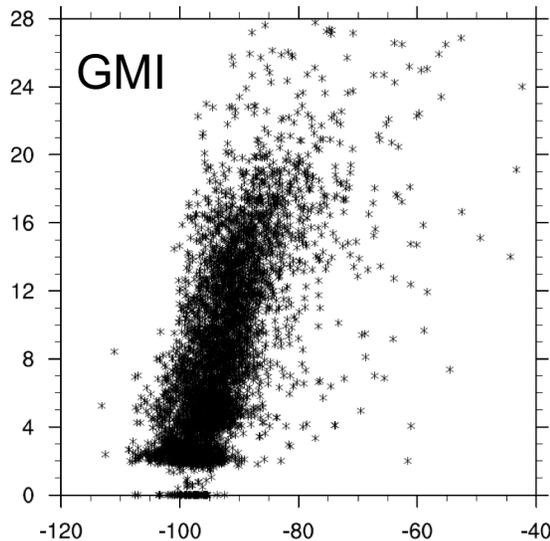
- There is a **weak** correlation between $PC(T)$ and $PC(W)$.
- What about the top precipitation height H ?

$$[h_{0.05}, h_{0.1}, h_{0.2}, h_{0.3}, h_{0.5},] \in H$$

- $h_{threshold}$ is the maximum height at which the water content is greater than the threshold (g/m^3)
- H is obtained from W
 - Can we use T to estimate H ?

Correlation analysis

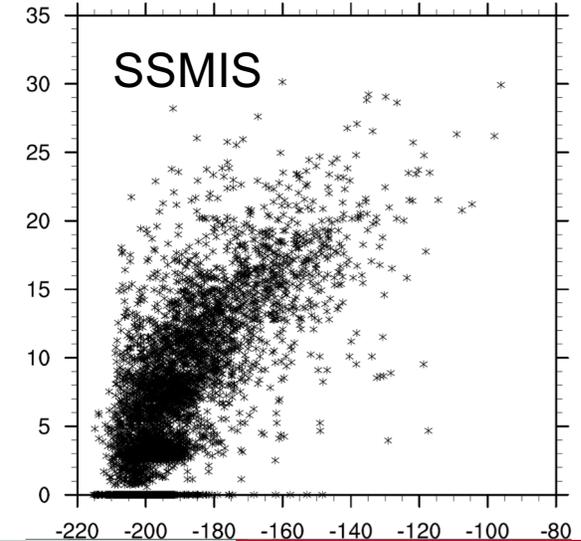
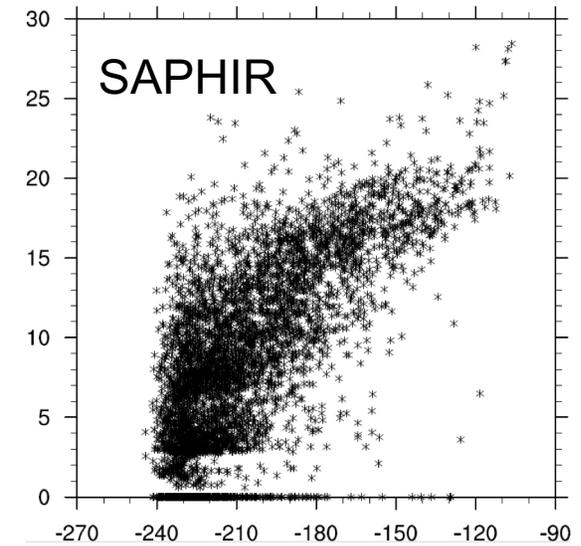
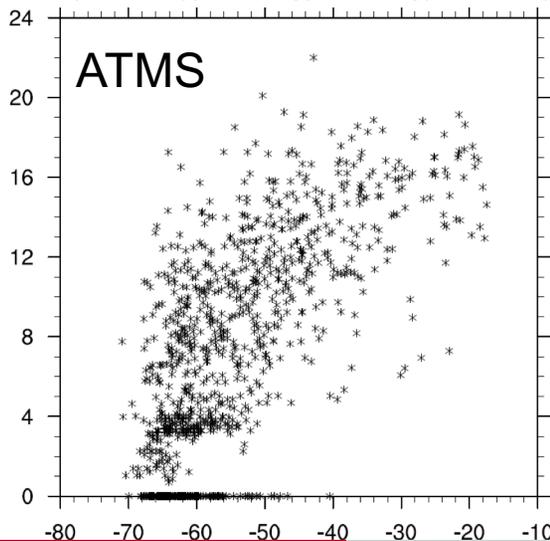
Top height correlation



- A strong correlation exists in this case

- *Conclusion:*

We can use brightness temperatures to estimate the top height of condensed water



Outline

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Precipitation detection

- We developed a **systematic** method for obtaining the maximum height of precipitation from radiometer measurements
- The method produces a precipitation classifier based on two steps:
 - **Precipitation detection** using Linear Discriminant Analysis or LDA. This method classifies any t into precipitation (P) or dry (D)
 - **Precipitation height estimation** using lookup tables based on Principal Components Analysis or PCA. This step gets the estimated height from t measurements

Precipitation detection

Source datasets

- We created intersection databases (radar – radiometer) from the following instruments:
 - Dual-frequency Precipitation Radar (**DPR**)
 - GPM Microwave Imager (**GMI**)
 - **SAPHIR** humidity sounder
 - Advanced Technology Microwave Sounder (**ATMS**)
 - Special Sensor Microwave Image/Sounder (**SSMIS**)
- Millions of records were processed.
- All HDF5 instrument files were downloaded from NASA's PPS-STORM system

Precipitation detection

LDA detector training

- We divided the intersection database into a training database and a testing database
- Every record in both databases was classified as precipitation (P) or dry (D) record depending on two thresholds:
 - h_{thr} defines the **minimum height** to consider precipitation
 - w_{thr} defines the **minimum water content** to consider precipitation
- We used the training database to obtain a LDA discriminant x and its evaluation threshold v_{thr}
$$x \cdot t > v_{thr} \text{ implies } P, \text{ otherwise } D$$

Precipitation detection

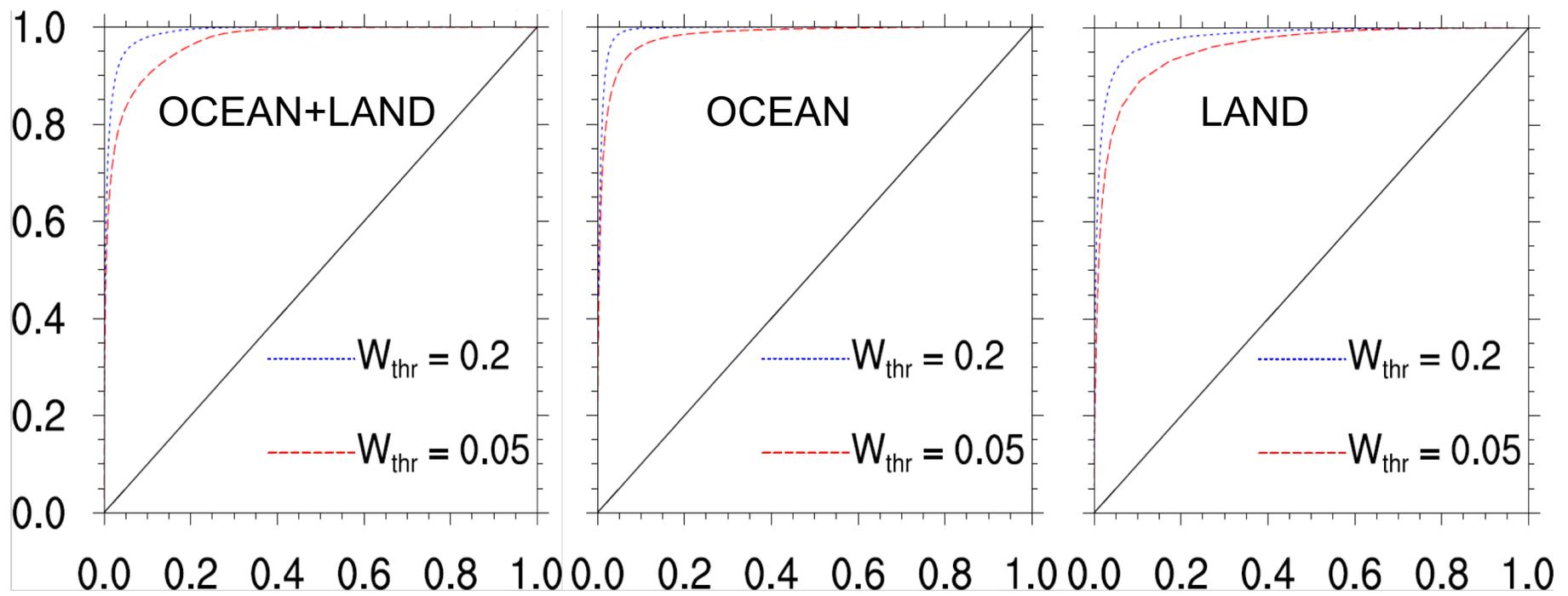
LDA detector testing

- The testing database was then used to obtain a *score* about how good the detection was
- *Receiver Operating Characteristic* (ROC) curves were obtained from the hits
 - Using different w_{thr}
 - Over land, ocean and both

Precipitation detection

Detector ROC GMI Ocean and Land

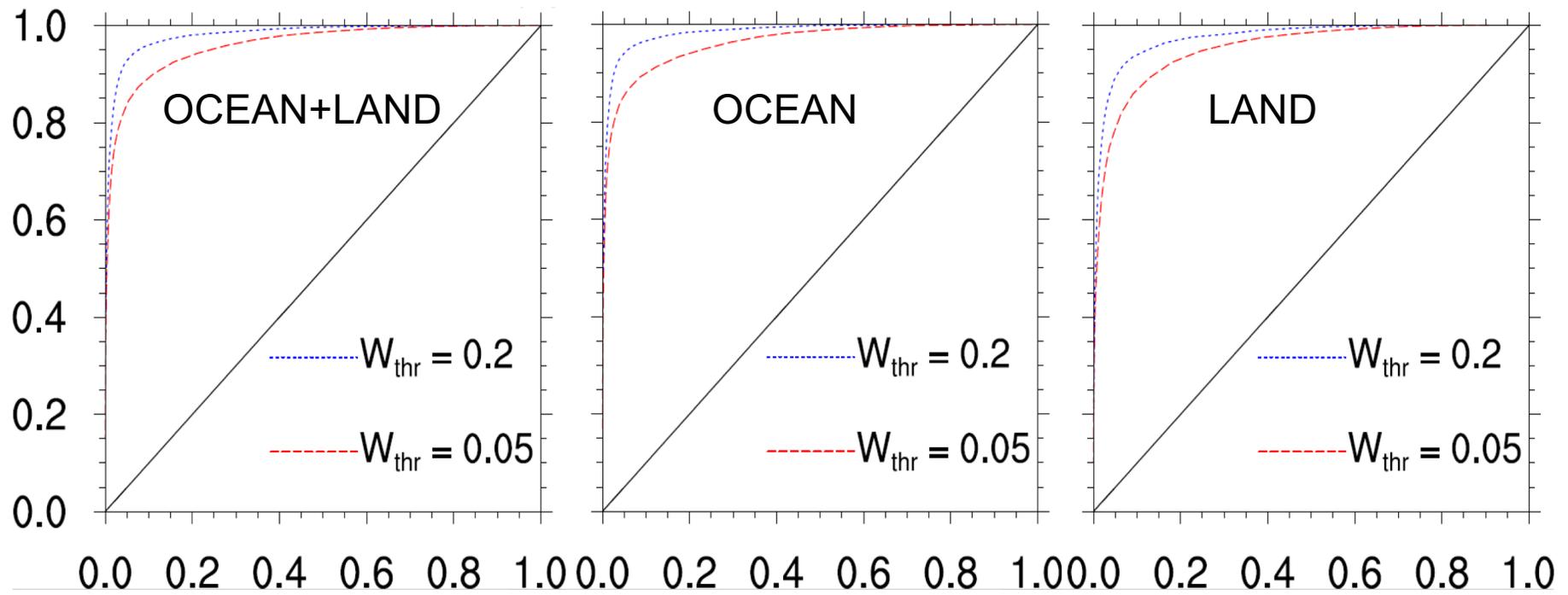
- OCEAN measurements were more accurate
- **Decreased** accuracy in LAND cases
- **Better detection** with higher w_{thr}



Precipitation classifier

Detector ROC SAPHIR Ocean and Land

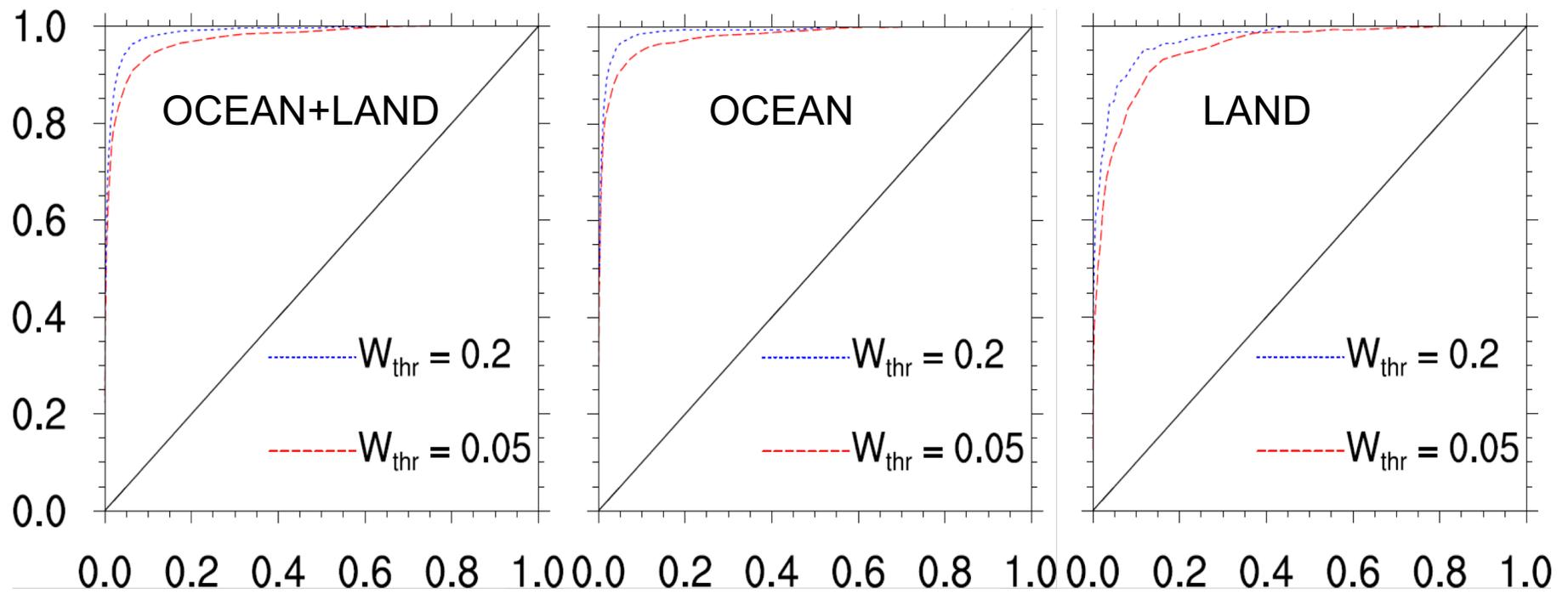
- Not much difference between OCEAN and LAND
- **Better detection** with higher w_{thr}



Precipitation detection

Detector ROC ATMS Ocean and Land

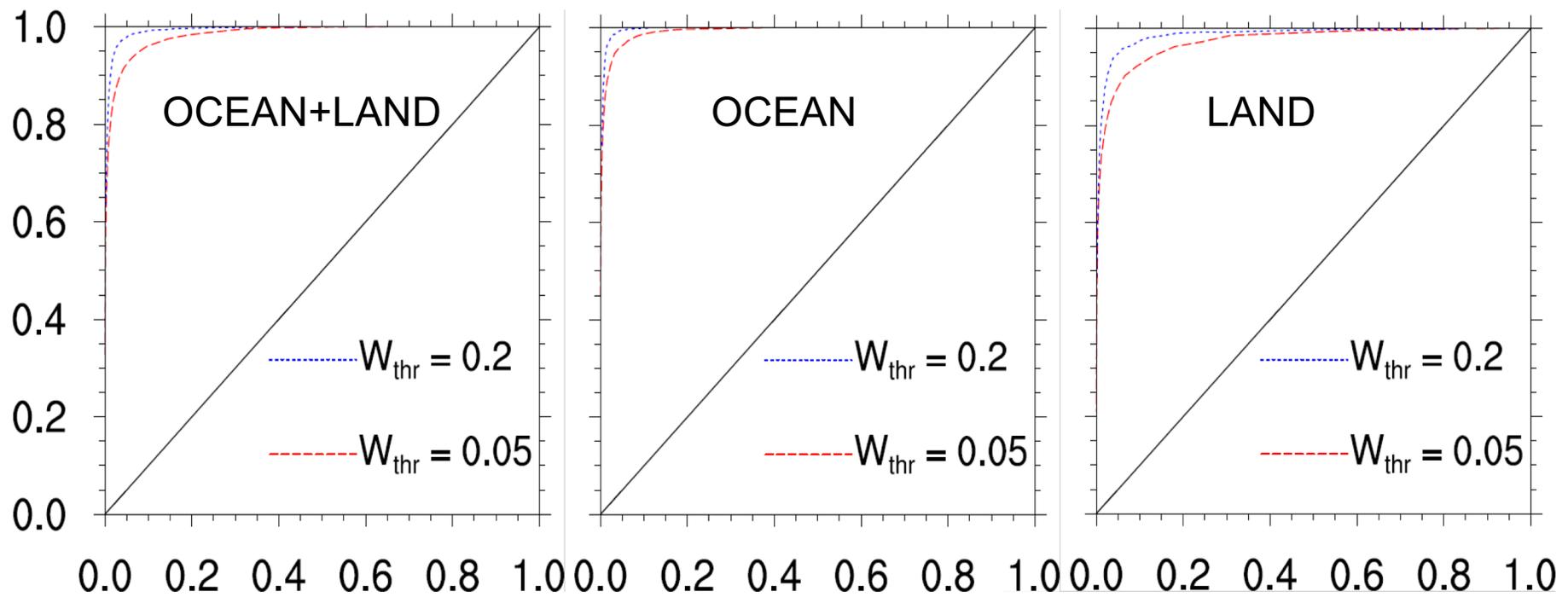
- LAND is performing **worse**
- **Better detection** with higher w_{thr}



Precipitation detection

Detector ROC SSMIS Ocean and Land

- LAND is performing **worse**
- **Better hit ratio** than other instruments
- **Better detection** with higher w_{thr}



Precipitation detection

Detection summary

- Over 85% of hit rate for all instruments
- **LAND** is a source of **uncertainty**
- SSMIS channels are very good for high precipitation detection.
- The higher the w_{thr} the better the precipitation detection
 - Caution: lower water content is undetectable

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Precipitation estimation

PCA classifier training

- We built **lookup tables** from the top height average values (expectation) in the training database
 - In every lookup table, every cell is **addressable** by $PC(t)$

$$[t_1, t_2, \dots, t_n,] \Rightarrow [PC_1(t), PC_2(t), PC_3(t)]$$

- The value of every cell is the expectation top height for the corresponding **Principal Components**

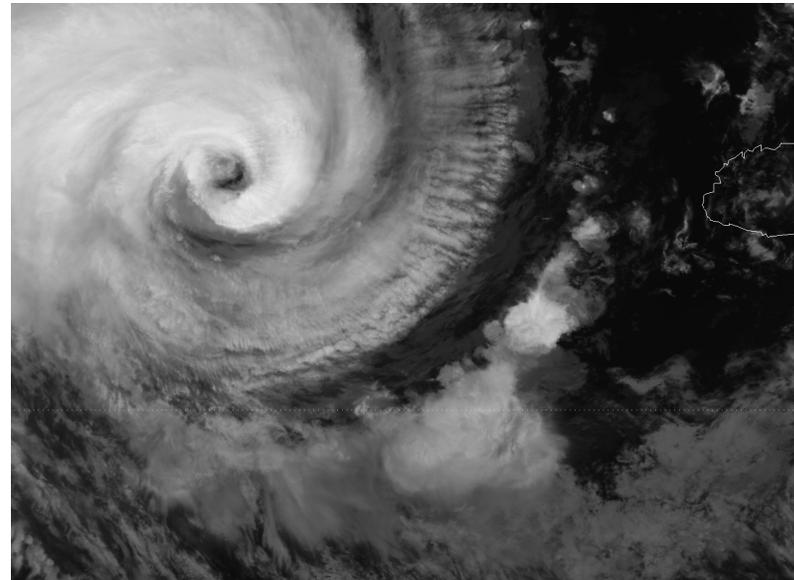
$$L(PC_1(t), PC_2(t), PC_3(t)) = E\{h(W)\}$$

- We obtained lookup tables for **every** instrument

Precipitation estimation

PCA classifier testing

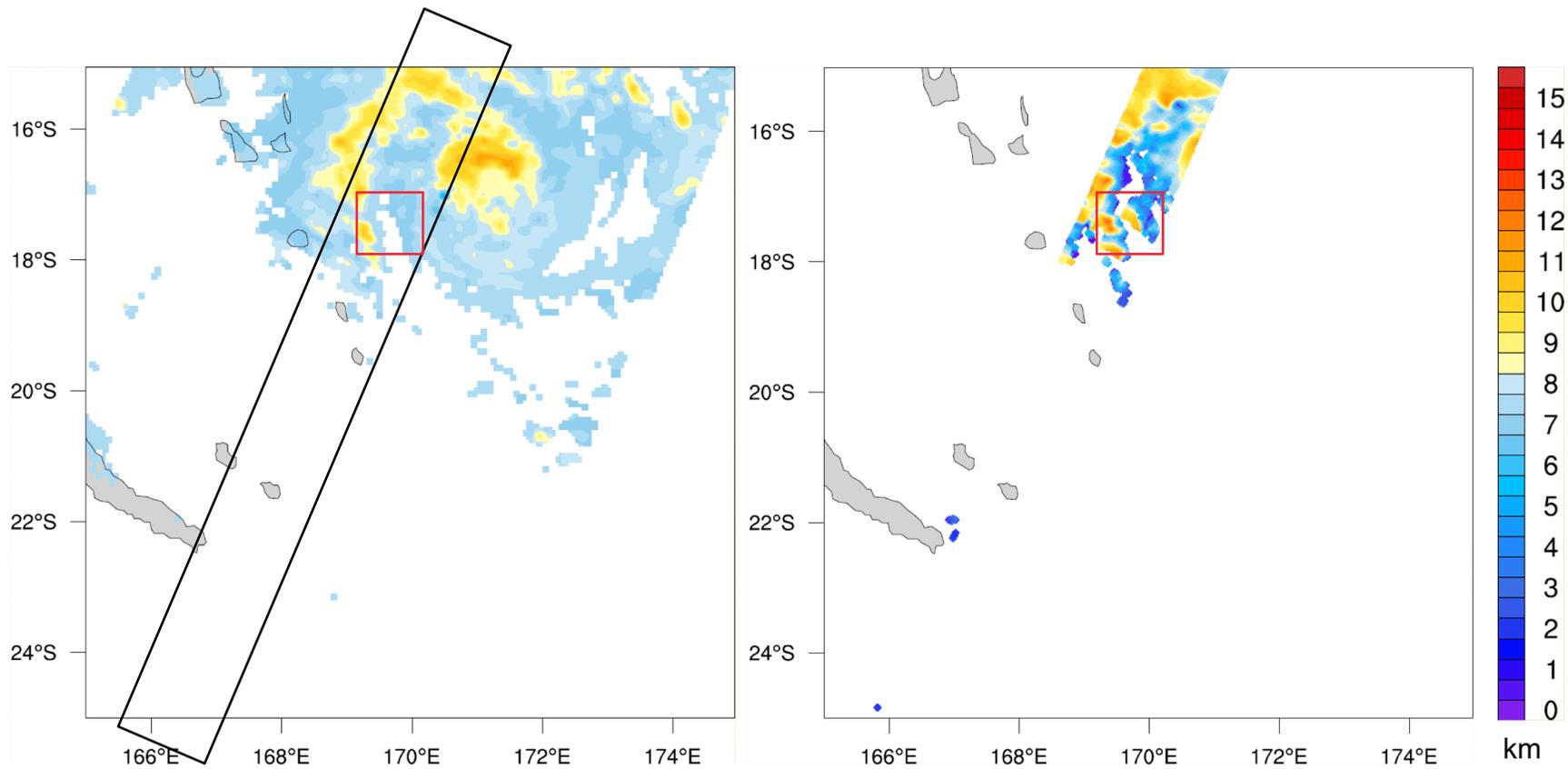
- We used some of the available data from Tropical Cyclone Winston
- Chosen dates:
 - 11 February 2016
 - 22 February 2016
- Chosen region:
 - Lat: -25 to 15 degrees
 - Lon: 165 to 175 degrees
- Parameters
 - $h_{thr} = 5 \text{ km}$
 - $w_{thr} = 0.05 \text{ g/m}^3$



IR image showing the Winston the 21 of February 22:00

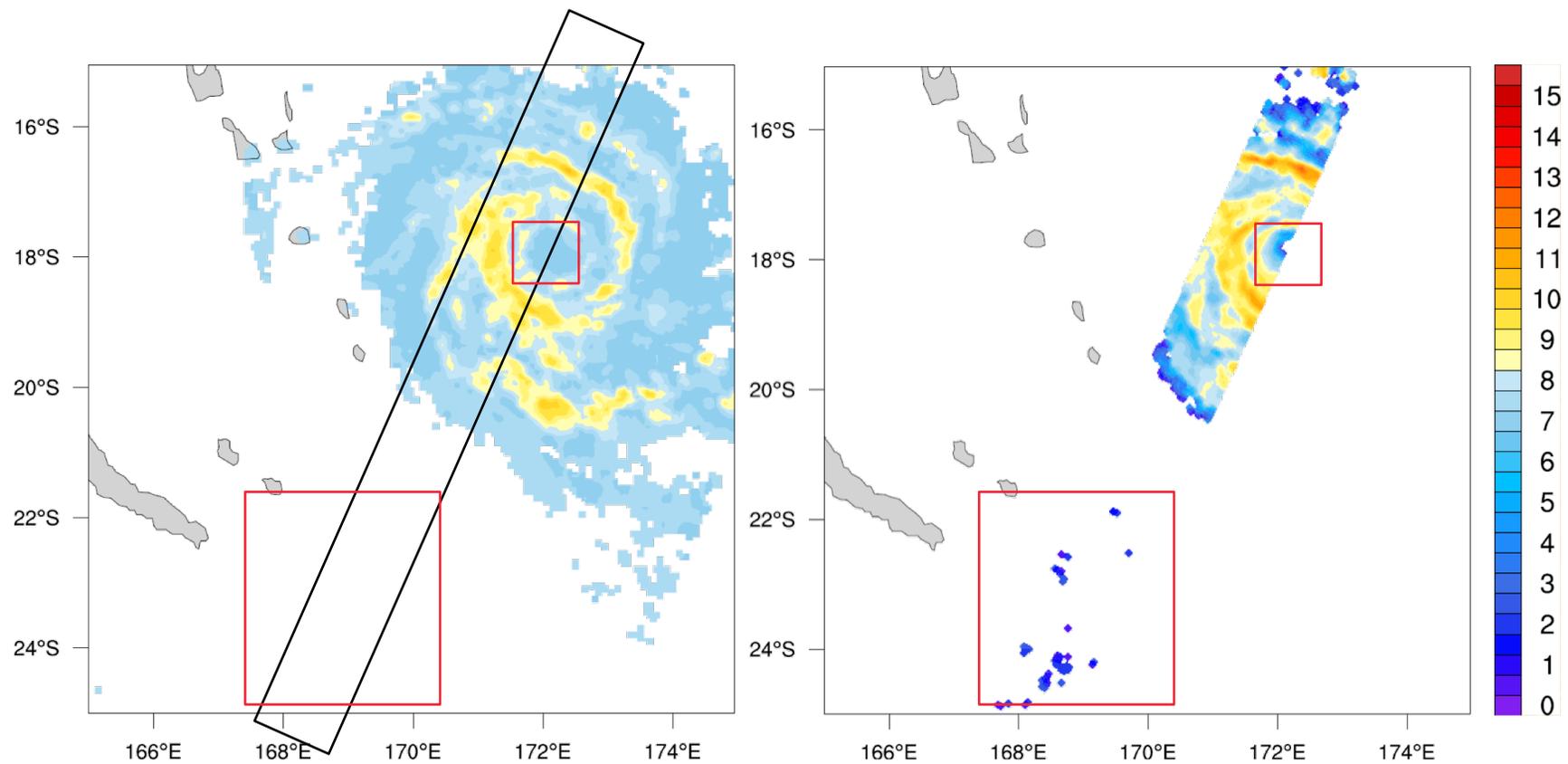
Precipitation estimation

GMI - 11 February estimation



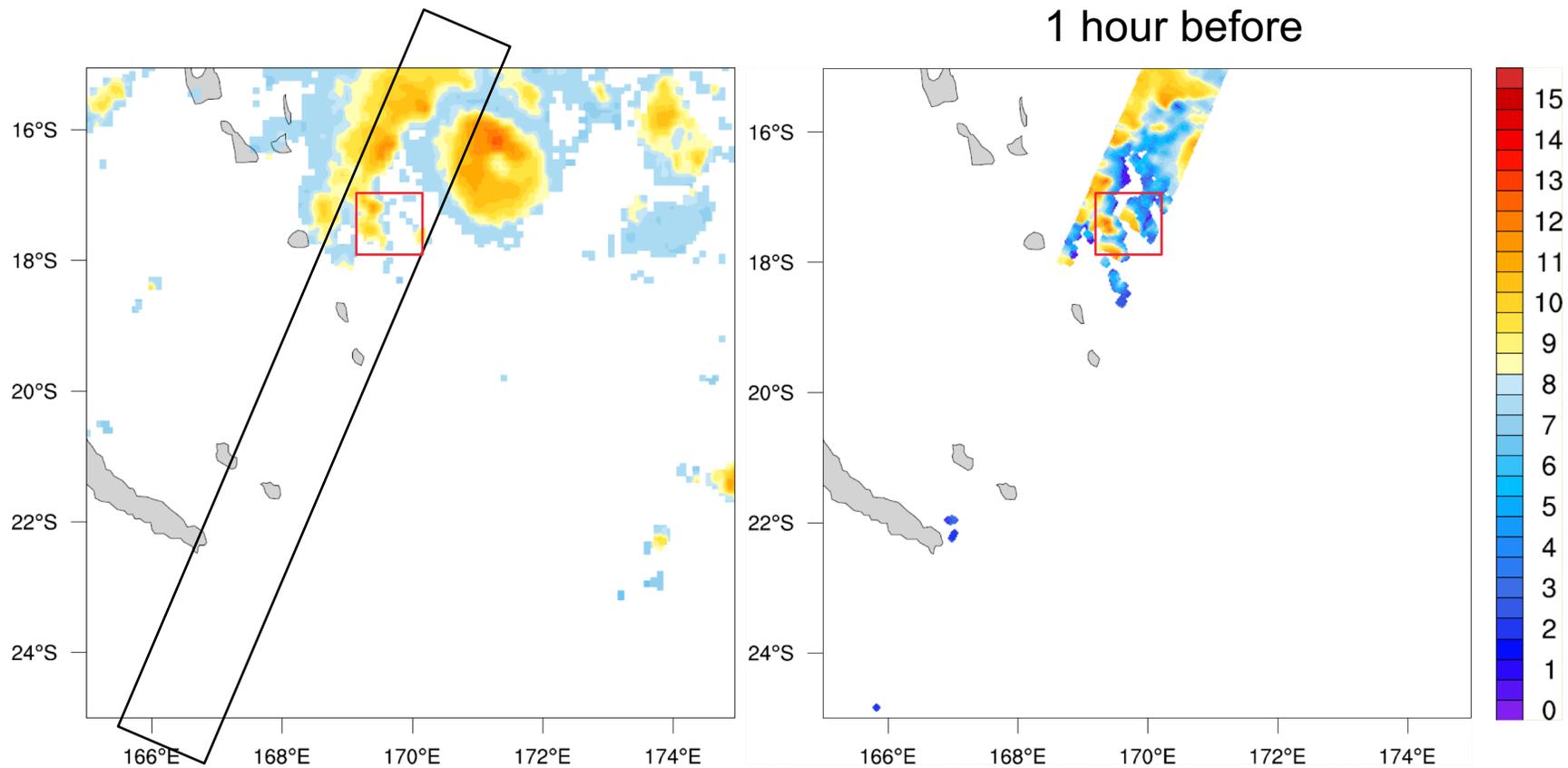
Precipitation estimation

GMI - 22 February estimation



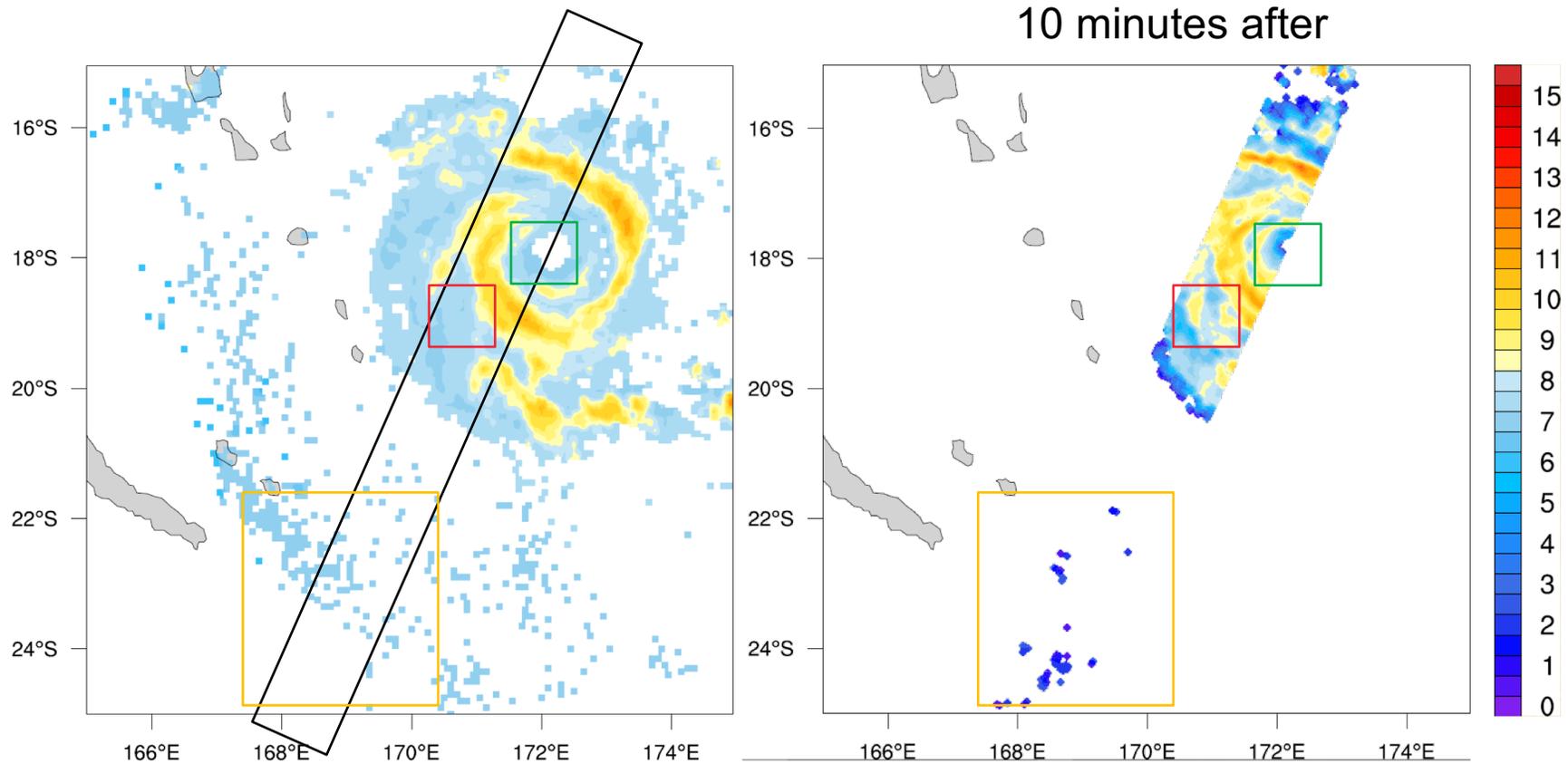
Precipitation estimation

SAPHIR - 11 February estimation



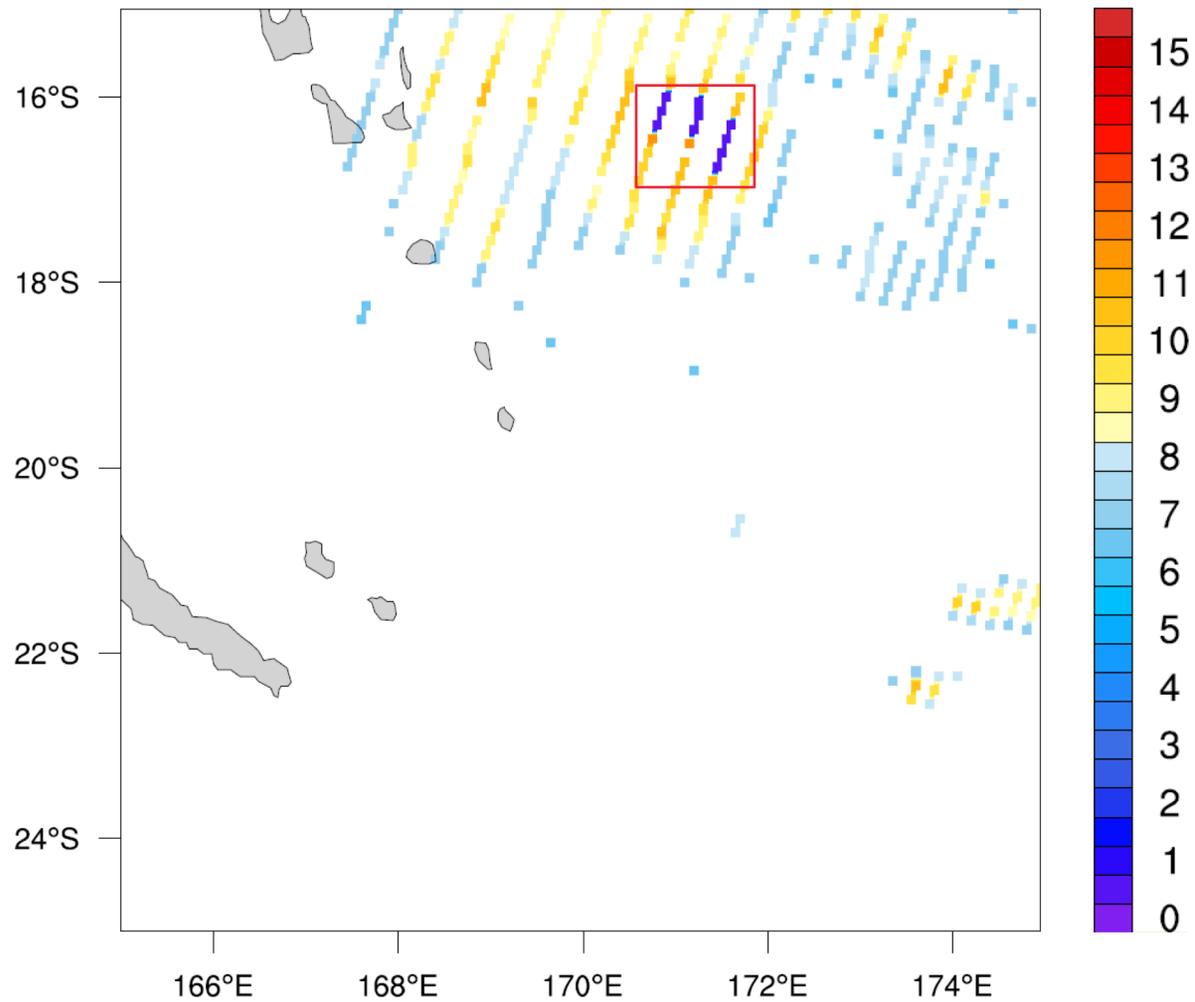
Precipitation estimation

SAPHIR - 22 February estimation



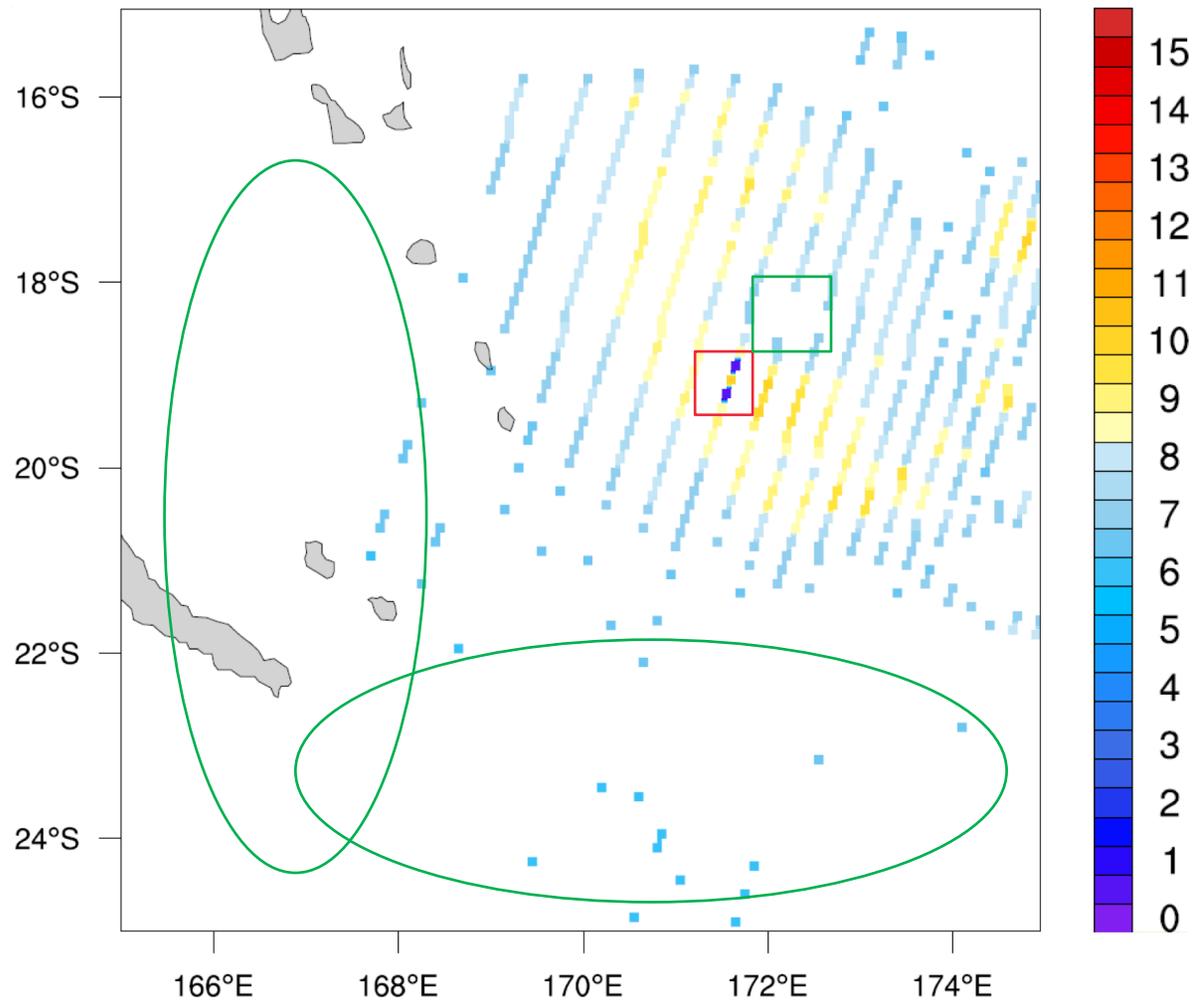
Precipitation estimation

ATMS - 11 February estimation



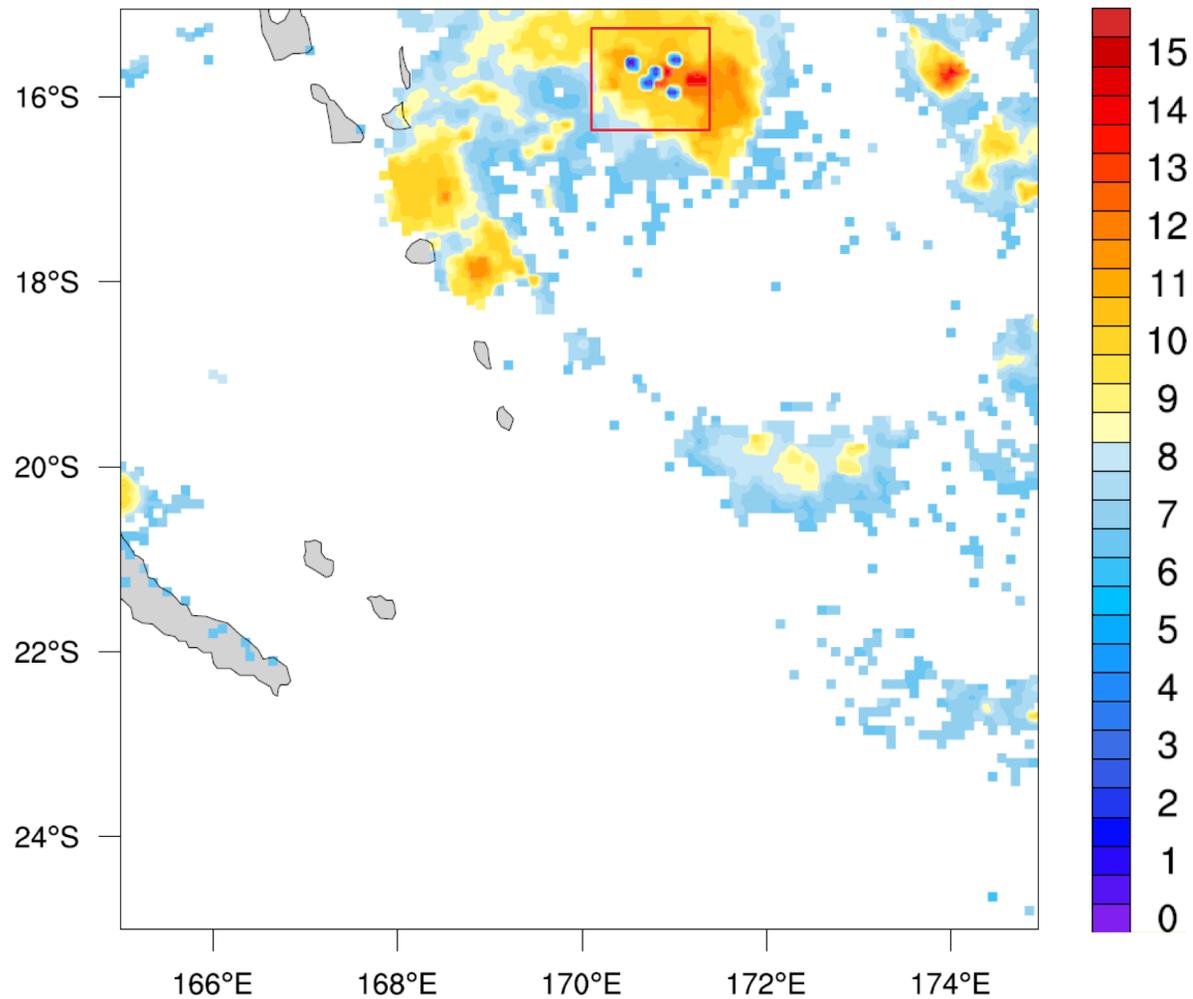
Precipitation estimation

ATMS - 22 February estimation



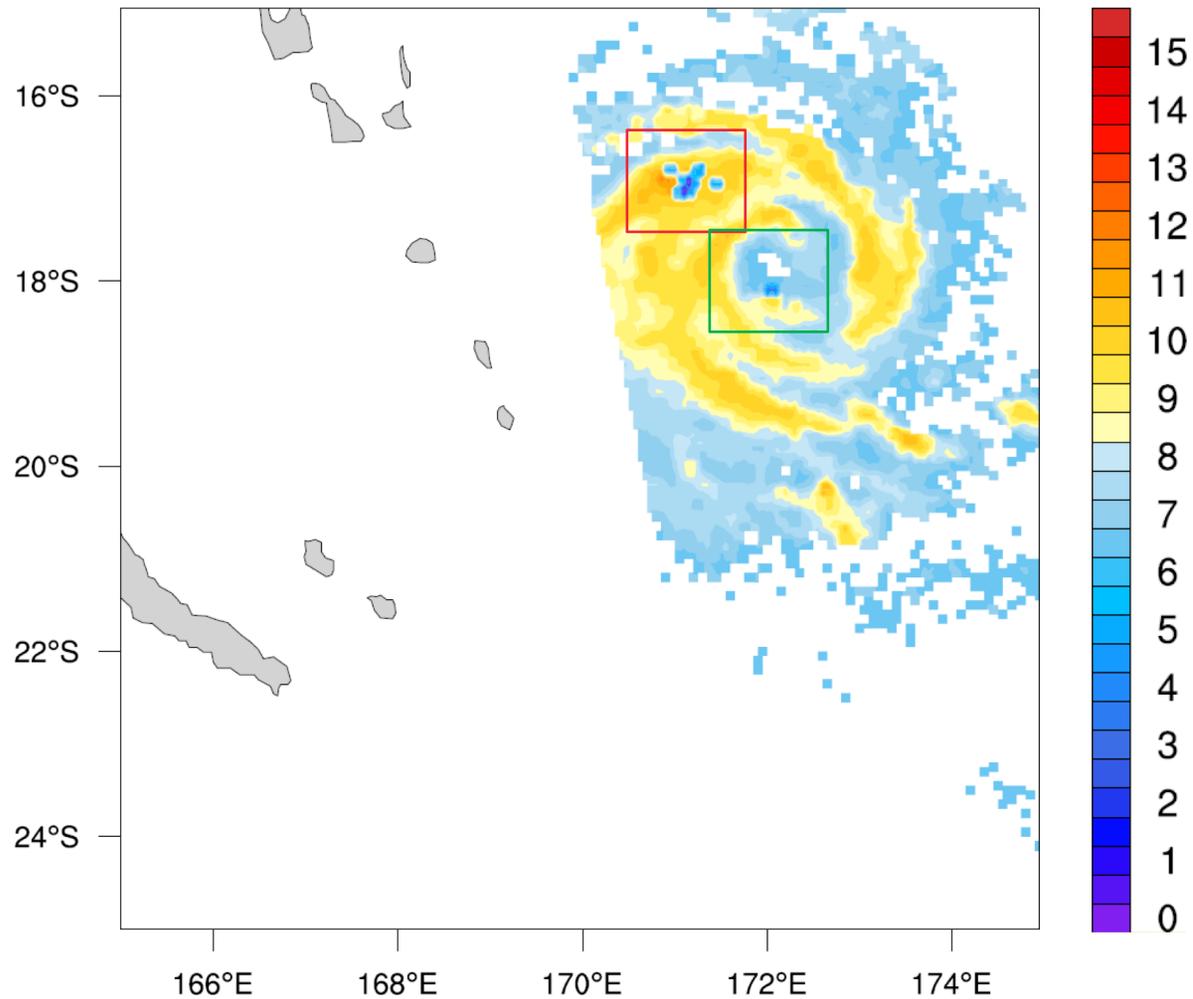
Precipitation estimation

SSMIS - 11 February estimation



Precipitation estimation

SSMIS - 22 February estimation



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Conclusions

Results

- We **cannot** infer **all** the water content information from brightness temperatures
- **Good estimations** of the maximum top height of precipitation can be obtained from just brightness temperatures
- The combined use of coincident DPR and the other instruments is critical to train the detection/estimation

Conclusions

Implications

- It is possible to build a **historical record** of convective precipitation
 - Radiometer information is available since 1998
- **Near-hour worldwide estimations**
- Improvement of convective system models
- Better estimations for **storm evolution**
- **Hazardous precipitation** detection in aviation weather
- Which radiometers channels are better for detecting convective precipitation
 - **New generation of instruments**

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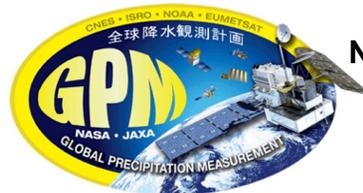
Thank you!

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