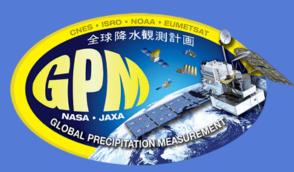


# GPM Level 1 Science Requirements: Science and Performance Viewed From the Ground

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### **1. GPM L1 Science Requirements for Precipitation Estimation**

The Global Precipitation Measurement (GPM) Mission Core satellite platform must meet Level 1 (L1) science requirements:

- GPM Dual-Frequency Precipitation Radar (DPR): *quantify rain rates* between 0.22 and 110 mm hr<sup>1</sup>. Demonstrate detection of snowfall at an effective resolution of 5 km.
- GPM Microwave Imager (GMI): *quantify rain rates* between 0.22 and 60 mm hr<sup>1</sup>. Demonstrate detection of snowfall at effective resolution of 15 km.
- Drop Size Distribution (DSD): GPM Core observatory radar estimation of  $D_m$  to within +/- 0.5 mm.
- Instantaneous *rain rate estimation* at 50 km resolution, *bias and random* error < 50% at 1 mm hr<sup>1</sup> and < 25% at 10 mm hr<sup>1</sup>, relative to GV

**Concept and Approaches** 

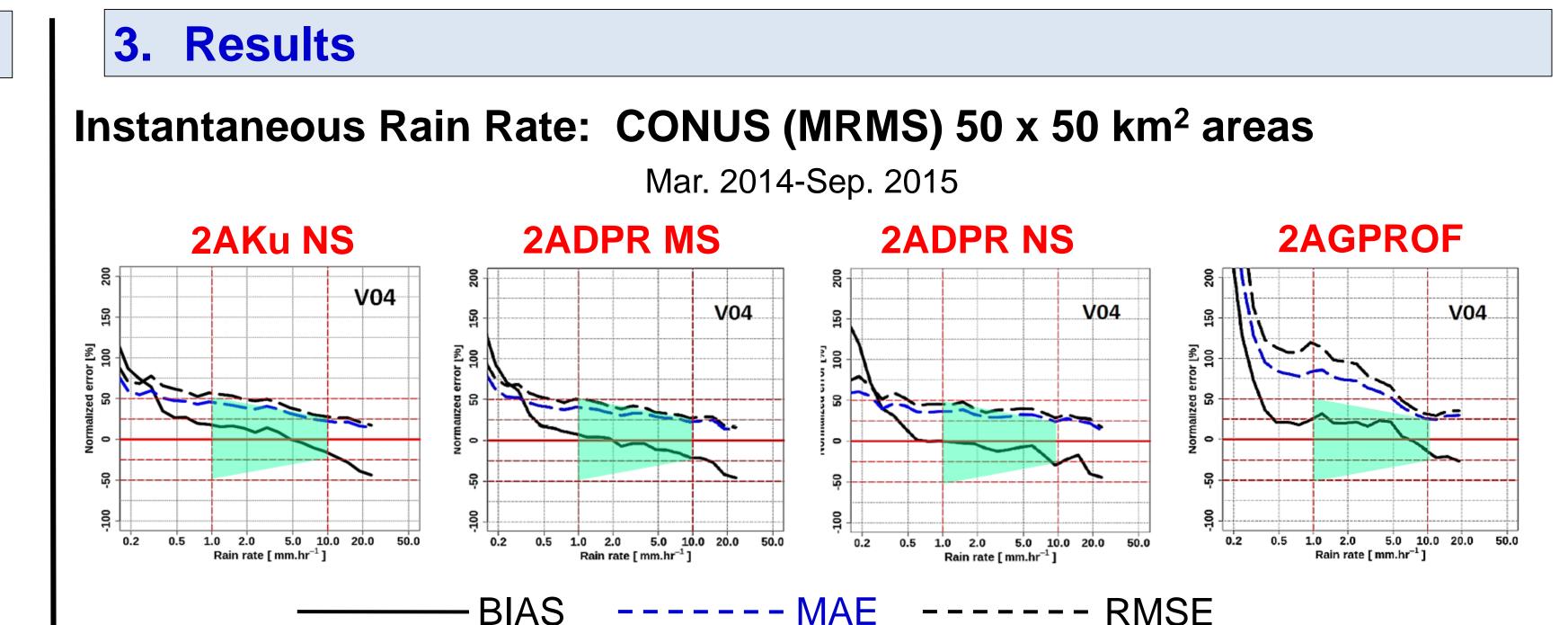
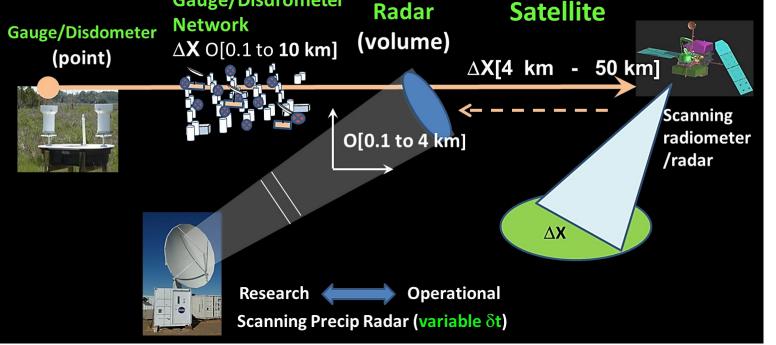


Figure 5. Bias and random errors (MAE and RMSE) for footprints averaged over 50 km areas for Ku normal swath (NS), DPR Ku NS, DPR Ku/Ka matched swath (MS), and GPROF products. Green polygons outline requirement boundary for 1 and 10 mm/hr. Note departure of GPROF from L1 requirements in random error at light rain rates.

**Ocean:** Kwajalein Atoll (KWAJ) and Middleton Island AK (PAIH)



Radars bridge point to area FOV/swath

Gauges, disdrometers reference multiparameter radar retrievals

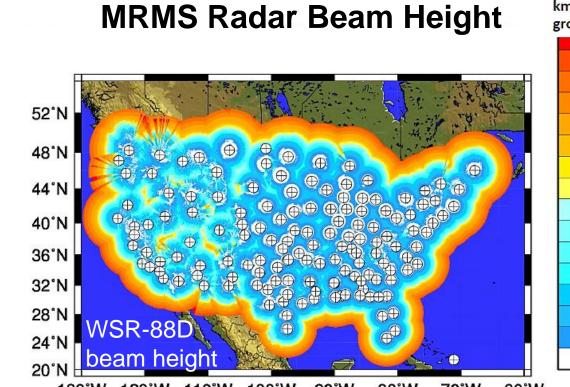
Figure 1. Radars as a bridge between scales

### Satellite FOV Footprint and Area Selection

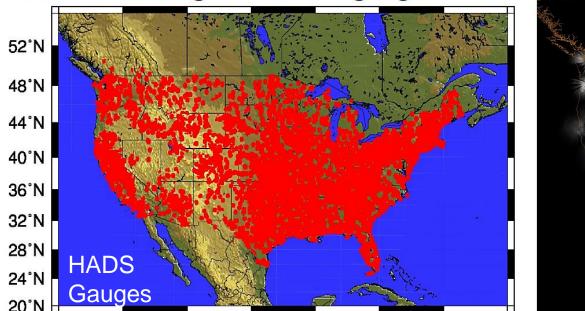
- 5 km DPR / 15 km GMI footprint "effective" resolution (FOV) assumed
- 50 km x 50 km averages (of footprints), but also computing footprint bias and scaled random error (5 km/15 km footprints to 50 km scale; Steiner et al., 2003 to mitigate sample numbers for rain rates > 10 mm/hr over 50 km scale)

### **Rain rate:**

(1) CONUS: Rain gauge bias-adjusted Multi-Radar Multi-Sensor (MRMS)



**MRMS** best observation areas ground level radar beam height < 2 km & gauge < 25 km 14.0



**MRMS** best observation area

Accumulation map (grey-scale) with 'best" area indicated by red box

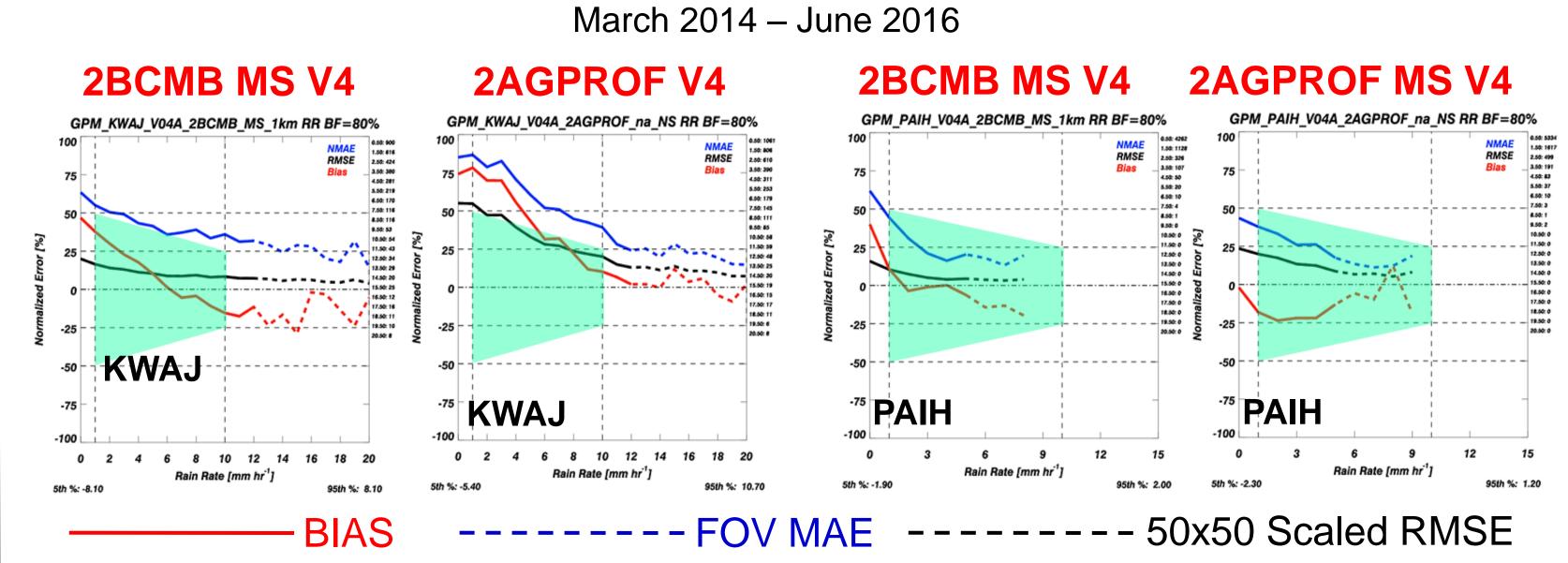


Figure 6. As in Figure 5 but for 2BCMB and GPROF algorithms only (left: KWAJ; right: PAIH). DPR and Ku NS swaths (not shown) similar or better than 2BCMB MS. Note: due to oceanic single radar sampling limitations, the bias and MAE traces are computed at footprint scale 5 km (15 km) for DPR (GPROF), with black line representing the RMSE scaled to 50 km. Dashed lines indicate rain rates for which sample numbers fall below ~30.

#### GPM Core observatory meets L1 rain rate science requirement based on Combined and DPR algorithm performance

# **DSD (D<sub>m</sub>) comparisons**

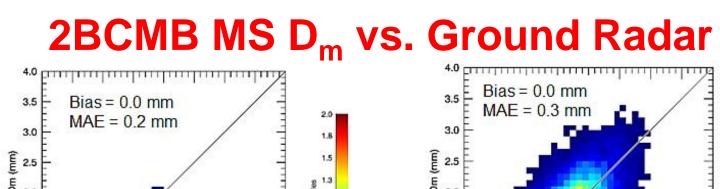


Figure 7. Example shown is for Validation Network comparison between the 2BCMB algorithm Version 3 (top) and Version 4 (bottom) radar-based estimates of  $D_m$  vs. DPR for stratiform (left column) and convective (right column) precipitation. 80-85% of total samples are stratiform-so, stratiform will weight final L1 result.

Figure 2. Left: Beam height at lowest elevation angle; center: HADS gauges used in MRMS; right: optimal MRMS area for observational comparisons based on beam height and distance to nearest gauge.

- (2) Ocean- Tropical and mid-latitude: Dual-pol rain estimators (range < 100) km) for Kwajalein (tropics) and Middleton Island, Alaska (high-latitude).
- Rain rates estimated at 500 m height
- Scaled footprint RMSE for Ocean radars (mitigate sample number issue)
- Beam filling: pixels fill 80% of FOV, 50% > 0 mm/hr at 50 km;
- GPROF Radiometer estimate: Probability of Precipitation > 40%
- 5<sup>th</sup>/95<sup>th</sup> % outliers removed

## **DSD- Drop Size Distribution (D<sub>m</sub>):**

• Polarimetric radar retrievals of  $D_m$  applied to ~70 radars in U.S. network using GPM Validation Network software for geometric match to DPR overpasses

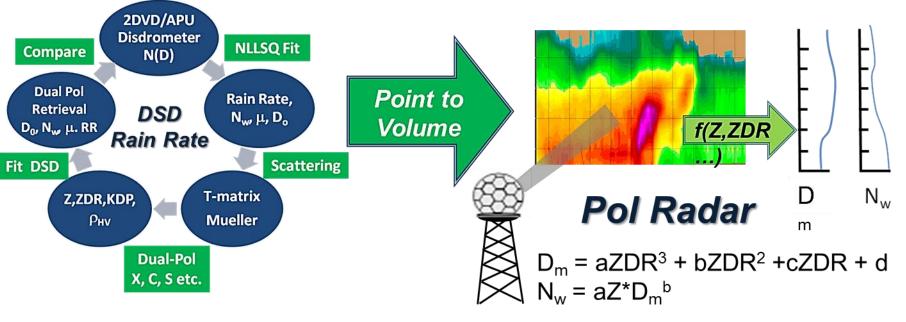
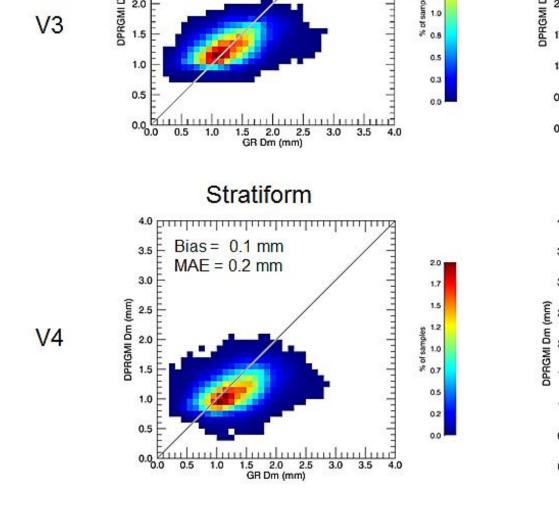


Figure 3. Polarimetric radar DSD modeling

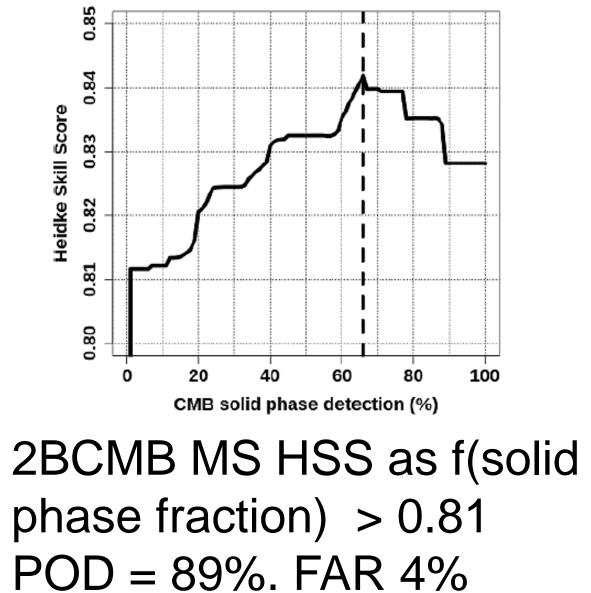
- Multi-regime 2DVD DSDs for 6 field efforts
- Robust ZDR-based retrievals Multi-Regime + Composite Empirical Relationships Ē 1.75 ഫ് 1.50

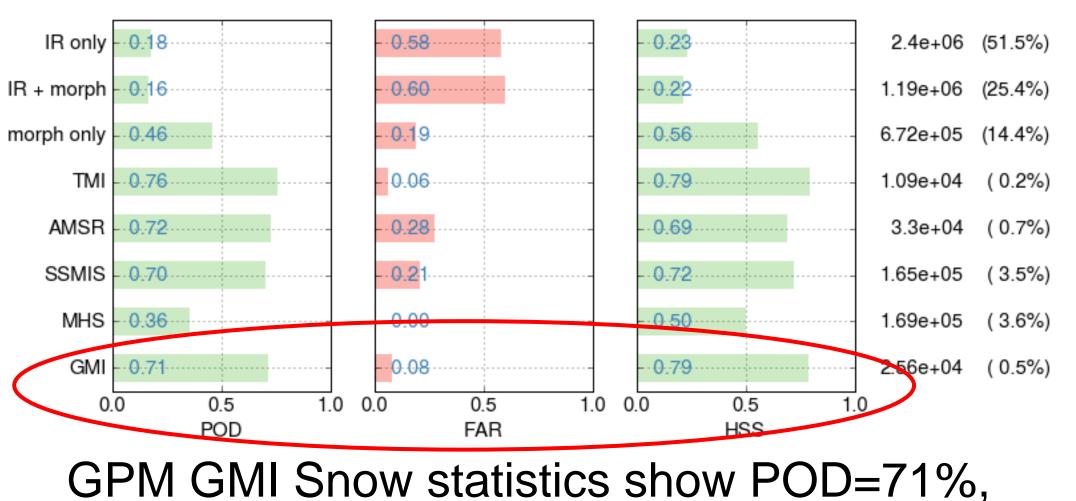


GPM bias + MAE in D<sub>m</sub> generally within 0.5 mm of GV for majority of sample.

## **Snow Detection at effective FOV (MRMS coincidences)**

Convective





FAR=8%, HSS = 0.79 (Courtesy, J. Tan, NASA GSFC)

• T-Matrix + fit N(D), and/or Rayleigh-Gans models + observed N(D) for pol variables • Fit  $D_m = f(ZDR)$  (polynomial) for each location and entire dataset

Cutoff at  $D_m < 0.5 \text{ mm}$  (No ZDR information) Sensitivity at large D<sub>m</sub>(>~3.5 mm) or ZDR(> 3-3.5 dB)

Parameterization error: Bias 1 - 10%, MAE 7 - 15%

Figure 4. Spectrum of  $D_m = f(ZDR)$  polynomial fits from different field campaigns/measurements

#### <u>Snow Detection: (Note: no liquid equivalent rate constraints!)</u>

 GPM Microwave platforms (e.g., GMI) in IMERG data files matched to MRMSdefined precip (snow) occurrence.

- L2 files, MRMS-defined snow with GMI POP 40%, <50 Liquid precip fraction (also Combined Alg.);
- DPR "phase near surface"; new "snow index" based for V5 (not shown) • Supplemental use of METAR or like databases (not shown)

We "demonstrate detection of snow", but determining lower detection threshold and accurately estimating snowfall rate, are outstanding problems.

### 4. Conclusion

• GPM meets Level 1 science requirements for rain estimation based on the strong performance of its radar algorithms. Changes in the V5 GPROF algorithm should correct errors in V4 and will likely resolve GPROF performance issues relative to L1 requirements. •L1 FOV Snow detection largely verified but at unknown SWE rate threshold (likely < 0.5 – 1 mm/hr liquid equivalent). Ongoing work to improve SWE

rate estimation for both satellite and GV remote sensing.