The Global Precipitation Measurement (GPM) Mission for Science and Society

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Submitted 1 August 2016

Revised 18 April 2019

Capsule: The Global Precipitation Measurement (GPM) mission collects essential rain and snow data for scientific studies and societal benefit.

Abstract—Precipitation is a key source of freshwater; therefore observing global patterns of precipitation and its intensity is important for science, society, and understanding our planet in a changing climate. In 2014, NASA and the Japan Aerospace Exploration Agency (JAXA) launched the Global Precipitation Measurement (GPM) Core Observatory (GPM-CO) spacecraft. The GPM-CO carries the most advanced precipitation sensors currently in space including a dual-frequency precipitation radar provided by JAXA measuring the three-dimensional structures of precipitation and a well-calibrated, multi-frequency passive microwave radiometer providing wide-swath precipitation data. The GPM-CO was designed to measure rain rates from 0.2-110.0 mm h^{-1} and to detect moderate to intense snow events. The GPM-CO serves as a reference for unifying the data from a constellation of partner satellites to provide next-generation, merged precipitation estimates globally and with high spatial and temporal resolutions. Through improved measurements of rain and snow, precipitation data from GPM provides new information such as: details on precipitation structure and intensity; observations of hurricanes and typhoons as they transition from the tropics to mid-latitudes; data to advance near-real-time hazard assessment for floods, landslides and droughts; inputs to improve weather and climate models; and insights into agricultural productivity, famine, and public health. Since launch, GPM teams have calibrated satellite instruments, refined precipitation retrieval algorithms, expanded science investigations, and processed and disseminated precipitation data for a range of applications. The current status of GPM, its ongoing science, and future plans will be presented.

Introduction and Motivation

Water is essential to our planet, Earth. It literally moves mountains through erosion; transports heat in Earth's oceans and atmosphere; keeps our planet from freezing due to radiative impacts of atmospheric water vapor; causes catastrophes through droughts, floods, landslides, blizzards, and severe storms; but most importantly water is vital for nourishing all life on Earth. Precipitation as a source of freshwater links the Earth's water and energy cycles. Thus knowing when, where, and how precipitation falls is of paramount importance for science and society.

While there are areas of the world that have dense ground-based sensors for measuring precipitation in the form of rain gauges and radars, the vast oceans, less populated regions, and parts of developing countries lack adequate surface measurements of precipitation (Kidd et al. 2016). Satellites provide an optimal platform from which to measure precipitation globally. In 1997, NASA and the National Space Development Agency of Japan (NASDA), now the Japan Aerospace Exploration Agency (JAXA), launched the Tropical Rainfall Measuring Mission (TRMM) (Simpson et al. 1998, Kummerow et al. 1998, 2000), which operated until April 2015. The TRMM spacecraft had both a passive microwave multi-frequency imaging radiometer (provided by NASA) and a Ku-band radar channel (provided by NASDA) capable of generating three-dimensional views of precipitation structure (Kozu et al. 2001). TRMM's data continue to foster important scientific investigations such as Curtis et al. (2007), Adler et al. (2009), Shepherd et al. (2011), Liu et al. (2012), Houze et al. (2015), and Liu and Zipser (2015). In addition, TRMM has a large user community that has applied these data operationally to support decision making (Kirschbaum et al., 2016).

The Global Precipitation Measurement (GPM) Core Observatory (GPM-CO) spacecraft is an advanced successor to TRMM, with additional channels on both the Dual-frequency Precipitation Radar (DPR) and on the GPM Microwave Imager (GMI) with capabilities to sense light rain and falling snow (Hou et al. 2014, Hou et al. 2008). The GPM-CO, also a NASA-JAXA partnership, was launched in February 2014 and currently operates in a non-sun-synchronous orbit with an inclination angle of 65^{o}. This orbit allows the GPM-CO to sample precipitation across all hours of the day from the tropics to the Arctic and Antarctic circles and for observing hurricanes and typhoons as they transition from the tropics to mid-latitudes. GPM expands TRMM’s reach not only in terms of global coverage, but also through sophisticated satellite instrumentation, the inter-calibration of datasets from other microwave radiometers, coordinated merged precipitation data sets, reduced latency for delivering data products, simplified data access, expanded global ground validation efforts, and integrated user applications. Because of the application focus of GPM, the public release of precipitation products is required in near-real-time (1-5 hours after the observations are downlinked to the ground stations).

The GPM mission has several scientific objectives including (1) advancing precipitation measurements from space, (2) improving knowledge of precipitation systems, water cycle variability and freshwater availability, (3) improving climate modeling and prediction, (4) improving weather forecasting and four-dimensional (4D) reanalysis, and (5) improving hydrological modeling and prediction. More details about these scientific objectives can be found in Hou et al. (2014).

The GPM-CO well-calibrated instruments allow for scientifically-advanced observations of precipitation in the mid-latitudes where a majority of the Earth’s population lives. The central panel of Figure 1 shows the coverage of the GPM-CO, and several interesting precipitation events are shown in panels a-l. These examples indicate the breadth of GPM's observational capabilities through measurements of diverse weather systems, such as severe convection, falling snow, light rain, and frontal systems over both land and ocean. The measurements include surface precipitation rates available from GMI and 3-dimensional precipitation structure from DPR.

A founding concept of the GPM mission is the constellation of precipitation observations provided by national and international satellite partners of opportunity. International and national partnerships are formed independently by both NASA and JAXA for sharing satellite data, ground validation measurements, and scientific expertise (Hou et al. 2014). The GPM-CO serves as a calibrator to ensure unified precipitation estimates from all satellite partners at high temporal (0.5 to 3.0 hours) and spatial (5 to 15 km) scales (Hou et al. 2014). Such satellite precipitation datasets can be merged via algorithms and accumulated over time as shown in Figure 2. These GPM products allow for detailed investigations of how and where precipitation is distributed and how these patterns change over days, seasons, and years. These estimates are also used to model and estimate hazard impacts (e.g. floods and droughts), weather related disasters, agricultural forecasting, and famine warnings (Kirschbaum et al., 2016).

The GPM-CO instruments and constellation concept will be discussed in Section 2. Precipitation retrieval algorithms, data products, processing, and availability will be presented in Section 3. Section 4 will be devoted to early validation results. In Section 5, the paper will summarize how GPM data have been used over the past two years for selected scientific investigations and societal applications. Material presented herein is primarily from the U.S. Science Team. Nevertheless, it is important to note that the current and future successes of GPM are joint with our international partners, especially Japan. The paper will close with conclusions and next steps.

GPM Core Observatory and Constellation Configuration

An essential activity of the GPM mission is the use of the NASA-JAXA GPM-CO to unify and inter-calibrate data sets generated by constellation satellite partners and merge these into next-generation, high temporal resolution global precipitation estimates. Fundamental to the success of this activity is both the GPM-CO instrumentation and the constellation configuration.

GPM Core Observatory

The GPM-CO was launched February 28, 2014 at 3:37am JST (February 27, 2014 18:37 UTC) from Tanegashima Island, Japan. The prime mission lifetime (instrument design life) is 3 years and 2 months (for checkout) but fuel is projected to last well beyond that, potentially lasting 15 or more years if the instruments/spacecraft systems (e.g., batteries) do not fail and fuel requirements do not increase. The GMI and DPR together provide a powerful synergistic tool to assess precipitation micro- and macro-structure, intensity and phase globally at relatively high (regional) resolutions. The DPR with Ku-band (35.5 GHz) and Ka-band (13.6 GHz) channels provides three-dimensional (3D) precipitation (rain and snow) particle structure with vertical resolution of 250m, a horizontal resolution of ~5 km, and swath width of 125 km (Ka) and 245 km (Ku) (Hou et al. 2014). The DPR was extensively calibrated pre-launch (Kojima et al., 2013) and its performance meets mission requirements (e.g., Kubota et al. 2015, Kubota et al. 2016, Toyoshima et al. 2015). (See also the sidebar on GPM's Mission Science Requirements.)

The GMI is a 13 channel conically scanning microwave radiometer (see Table 1 and Hou et al. 2014 for details). GMI provides wide-swath (885 km) TB data to estimate surface precipitation at resolutions ranging from 5-25 km depending on frequency. Design requirements for GMI were driven both by requirements to build *a priori* databases to support Bayesian microwave precipitation retrieval algorithms (Kummerow et al. 2010, Kummerow et al. 2015) as well as to provide a reference radiance calibration standard for the GPM constellation (Hou et al. 2014). The design features needed to meet the requirements include a shroud over the warm load to eliminate solar intrusions, a robust reflective antenna coating to minimize emissivity issues, and the addition of noise diodes for a four point calibration of the window channels (Draper et al. 2013, 2015a, 2015b). The GMI instrument is meeting its performance requirements (Draper et al. 2015) and has already been deemed one of the best calibrated conically scanning passive microwave radiometers in space with brightness temperature accuracy for all channels within 0.4K and stability within 0.2K (Wentz and Draper, 2016).

GPM Constellation Configuration

The GPM mission encompasses the GPM-CO and a constellation of about 10 satellites (as of mid-2016) from national and international partners of opportunity [see Table 1 and Hou et al. 2014 for details]. These satellites are designed and operated for the partners' missions, but these agencies are willing to share their data with GPM for the purpose of producing next-generation unified global precipitation estimates. The constellation satellites bearing passive radiometers fly independent polar or non-sun-synchronous orbits allowing for multiple coincident overpasses with the GPM-CO.

For the constellation partner data, the first step toward unified precipitation estimates is the inter-calibration of brightness temperatures (TB) using GMI as the reference standard. This ensures that the observed TB are consistent among the sensors with expected differences after accounting for variations in the observing frequencies, bandwidths, polarizations, and view angles (see Wilheit 2013, Wilheit et al. 2015, Zavodsky et al. 2013, Zhang et al. 2011, 2016 and Table 1). Figure 3 shows the extent of coverage provided by single 98-minute orbits for each of the various radiometer types in the GPM constellation.

Sensor inter-calibration between GMI and the partner sensors involves several steps, as described in Wilheit (2013, 2015) and Berg et al. (2016). Multiple independent approaches are compared during these steps, which help to identify flaws or limitations of a given approach, thus increasing confidence in the results and providing a measure of the uncertainty in the resulting calibration adjustments. After adjustments, residual differences between GMI channels and those on the constellation radiometers are generally smaller than 1 K (Berg et al. 2016). This is a remarkable achievement that now allows the project to focus on the precipitation products rather than TB uncertainties.

Future satellite inter-calibration tasks include understanding and quantifying the residual uncertainties in the estimated calibration differences due to the radiative transfer models and geophysical parameter retrievals and adapting to changes in the radiometer constellation. Updates in the GMI calibration algorithms and subsequent inter-calibration adjustments to the constellation sensors will occur during scheduled reprocessing of retrieval products. In addition, inter-calibrating TRMM's TMI and pre-GPM microwave constellation sensor data to GMI is necessary for generating a consistent long-term next-generation precipitation record that covers the TRMM and GPM eras.

Algorithms, Data Products, Data Processing and Data Availability

The GPM-CO data processing is a joint NASA/JAXA effort. NASA data processing is done at GSFC (Greenbelt, MD) in the Precipitation Processing System (PPS). JAXA data processing is carried out at the Tsukuba Space Center (Tsukuba, Ibaraki, Japan) in the Mission Operations System (MOS). The interconnected architecture of this joint mission ground system can be seen in Figure 4. Working with the GPM principal investigators and science algorithm developers, PPS maintains the operational science data processing system and ensures the timely processing of all GPM science instrument data [see Hou et al. 2014 for a table of GPM products]. During routine operations, raw instrument data (Level 0 data) is received in near-real-time by the PPS and processed using science algorithms to produce calibrated, swath-level instrument (Level 1, L1) data. JAXA's MOC processes DPR Level 1 products and their Level 3 merged satellite products. Additional algorithms are used to compute geophysical parameters such as precipitation rate at the swath-level resolution (Level 2, L2 data products). [For reference, a special collection of papers describing the L2 precipitation algorithms is appearing in the *Journal of Atmospheric and Oceanic Technology*.] At the final stage of processing, Level 3 (L3) algorithms produce gridded and accumulated geophysical parameters including products such as latent heating profiles (e.g., Tao et al. 2016). It is envisioned that Level 4 data products developed through model-assimilated precipitation forecast and analysis will be available in the future.

The GPM mission has both near-real-time (NRT) and research-quality production requirements. Both NASA and JAXA contribute key processing efforts to fulfill these latency requirements. The NRT products are produced using forecast or earlier forms of ancillary data.

NRT products include GMI TB, and precipitation estimates from GMI (denoted GPROF), DPR, and Combined Radar-Radiometer Algorithm (denoted CORRA) (Kummerow et al. 2015, Seto et al. 2015, Grecu et al. 2016). GMI products are available within an hour of data collection while DPR and CORRA are available within 3 hours of data collection. Another NRT product developed by the U.S. team is the Integrated Multi-satellitE Retrievals for GPM (IMERG) gridded retrieval that is a Level 3 NASA product (Huffman et al. 2015). JAXA produces an analogous product called Global Satellite Mapping of Precipitation (GSMaP) (Kubota et al. 2007, Aonashi et al. 2009, Ushio et al. 2009). IMERG uses the GPM-CO to inter-calibrate precipitation data from all constellation radiometers. Temporal and spatial gaps in the IMERG microwave precipitation estimates (e.g., as shown in Figure 3) are filled by morphing the estimates in between the microwave overpasses, and incorporating IR estimates with a Kalman filter where the gaps are too long (over about 3 hours) to produce 0.1^{o} x 0.1^{o} half-hour global products. The IMERG product is produced twice in NRT; once approximately 5 hours after data collection and again approximately 14 hours after data collection.

All of the NRT products are also processed as research products. The geolocation of the research products is more consistent as predictive ephemeris rarely needs to be used. Research products are produced by PPS when all the required high quality ancillary and geolocation data are received with the objective for accuracy, completeness, and consistency. These research products are available hours to months after data collection and are stable for long-term precipitation investigations. PPS generates and distributes all data from the instruments on the core satellite as well as Level 2 and Level 3 data from the partner constellation satellites. In addition to the standard HDF5 format files, a Geographic Information System (GIS; TIFF world files) product and ASCII text files are provided for selected product estimates. All GPM data are openly available and accessible from https://pmm.nasa.gov/data-access/downloads/gpm. JAXA's GPM products in general can be obtained from https://www.gportal.jaxa.jp/gp/top.html while the GSMaP multi-satellite merged data can be obtained from http://sharaku.eorc.jaxa.jp/. GPM data (Level 0-3) are periodically reprocessed as retrieval algorithms are improved. The at-launch Version 03 IMERG accumulation products are known to be high biased during heavy rain events and the next IMERG reprocessing to Version 04 (early 2017) is expected to address these high biases. GPM retrieval algorithms use the dual frequency channels of DPR and the high frequency channels of GMI and hence precipitation products from GPM are different than those from TRMM. Nevertheless, there are plans to reprocess inter-calibrated precipitation data (in winter 2017-2018) to produce a consistent long-term precipitation record that starts at the beginning of TRMM. GPM is meeting data latency requirements (as shown in the sidebar), on average, greater than 99% of the time. Recent PPS statistics show nearly 50 TB data downloaded by more than 1,000 unique users from all over the world in a single month.

Validation Efforts

GPM Ground Validation (GV) efforts include the direct statistical validation and verification of satellite estimates against high-quality ground measurements, and physical validation for algorithm improvement and hydrological models. Validating data is from both regular ongoing surface observations and focused field campaigns (Hou et al., 2014; see also https://pmm.nasa.gov/index.php?q=science/ground-validation). Major GPM validation efforts are: (1) Comparisons among satellite precipitation products, (2) comparisons against ground datasets, and (3) analysis for meeting mission requirements.

One evaluation technique compares zonal means among the various GPM instrument algorithms and established precipitation estimates such as the Global Precipitation Climatology Project (GPCP) data sets (Adler et al. 2003) and, over ocean, the Merged CloudSat, TRMM, Aqua version 2 (MCTA2) data (Behrangi et al. 2014). Both GPCP and MCTA2 include a variety of input data sets selected for utility in precipitation estimation at both low and high latitudes. Figure 5 shows the global zonal means for 2015 for land and ocean (Figure 5a), ocean only (Figure 5b), and land only (Figure 5c). This figure illustrates that DPR, Ku, CORRA, and GPROF algorithm retrievals are in good agreement. The GPM zonal accumulations underestimate with respect to the MCTA at higher latitudes. This is most attributable to the fact that the DPR minimum detectable reflectivities correspond to minimum rain rates of approximately 0.2 mm h^{-1}. Since much of the higher latitude precipitation is light, and CORRA and GPROF are based on DPR estimates, GPM is low in the higher latitudes. A high latitude, light precipitation solution for GPROF is being implemented in the upcoming algorithm Version 05 release. The mean daily precipitation in mm day^{-1} for each of the algorithms is provided in Table 2. This table shows that IMERG annual precipitation is lower than the other algorithms while there are interesting differences among the diverse approaches over land. Land surfaces tend to complicate the retrieval process and the various algorithms use different approaches to mitigate surface (emissivity and clutter) issues.

Direct statistical GV of GPM rainfall rate estimates relies primarily on existing high-resolution, quality-controlled U.S. national radar network rain rate products such as the NOAA

National Severe Storms Laboratory/University of Oklahoma Multi-Radar/Multi-Sensor (MRMS) products (e.g., Zhang et al., 2016 and references therein). Currently, the MRMS system (http://mrms.ou.edu) incorporates data from all polarimetric WSR-88D radars (NEXRAD), a large number of automated rain gauge networks, and model analyses in the Continental U.S. (CONUS) and southern Canada. The system creates a gridded mosaic of quantitative precipitation estimates (QPE) products on a 0.01° x 0.01° grid at a 2-minute temporal resolution (Zhang et al. 2016 for most recent updates). Of particular value to GPM GV are MRMS radar-based gauge-adjusted QPE. Collectively, these MRMS products provide an independent and consistent reference for directly evaluating post-launch GPM precipitation products across a large number of meteorological regimes as a function of resolution, accuracy, and sample size (Kirstetter et al. 2012).

For continental scale verification of GPM products over CONUS all MRMS data coincident with GPM orbits are continuously processed and saved as a GPM GV dataset (http://wallops-prf.gsfc.nasa.gov/NMQ/index.html). In addition to standard MRMS quality control procedures (see Zhang et al. 2016), additional procedures to minimize radar uncertainties are employed to derive a high-quality precipitation reference at the satellite product pixel resolution (Kirstetter et al. 2012). Filtering out instances when the radar-gauge ratios are outside of the range 0.1-10.0 further refines the instantaneous gauge bias-corrected MRMS product. In addition only radar data with the best measurement conditions (i.e., no beam blockage and radar beam below the melting layer) defined by a Radar Quality Index (RQI) are retained. Gridded 0.01° MRMS products can then be matched to allow direct comparisons between the surface radar and satellite precipitation products (see Figure 6).

Independent comparisons of this GPM GV-MRMS reference data set with two dense, well-maintained, and data quality-controlled NASA rain gauge networks show that for c. 5 km footprint, 30 minute accumulations > 0.5 mm h^{-1}, biases are < 10% while normalized mean absolute errors (NMAE) are < 35-40%. These results are consistent with a quantitative assessment of the MRMS accuracy performed at its native resolution (Kirstetter et al. 2015b). Individual satellite radar matches are subsequently averaged to coarser 50 km grids, useful for quick look comparison products (cf. http://wallops-prf.gsfc.nasa.gov/NMQ/index.html) and for verifying GPM Level-1 science requirements (e.g., Figure 7). Here the increased spatial averaging of the footprints together with removal of outliers (5^{th} and 95^{th} percentile) maintains low-bias while further reducing random error in the MRMS data relative to the 5 km footprint scale mentioned above.

The GPM-GV MRMS reference dataset and its derivatives have revealed and quantified several aspects of satellite-estimated rainfall retrieval errors and uncertainties including comparisons of rainfall detectability and rainfall rate distributions (Kirstetter et al. 2014), separation of systematic biases and random errors (Kirstetter et al. 2012), regional precipitation biases (Chen et al. 2013), influence of precipitation sub-pixel variability and surface (Kirstetter et al. 2015b; Carr et al. 2015), and comparison between satellite products (Kirstetter et al. 2013, 2014; Tan et al., 2016a, b).

Figure 6 provides an example of comparisons to GPM Core satellite products for instantaneous sampling times (e.g., coincident swath and MRMS sample time) as a density-scatter plot for individual near surface DPR sensor footprint scales (effective resolution 5 km). Here it is important to note that the scatter of the data exhibited in Figure 6 is expected based on the instantaneous nature of the comparison at high spatial resolution (e.g., effective FOV), and the related intrinsic random error associated with matching associated precipitation estimates in time and space between MRMS and GPM L2 data swaths. Comparisons at this scale are best interpreted as a tool for evaluating the broader systematic bias behavior between GPM products using the GV as a third reference.

In Figure 6 good agreement between the GV MRMS reference and the near surface DPR-Normal Scan (NS) algorithm Version 04 is evident with a bias (defined as the mean relative error; MRE) and normalized mean absolute error (NMAE) of only -9.8% and 51.7%, respectively. The Normal Scan mode of DPR consists of retrievals using the Ku-band 245 km wide swath data. The agreement is particularly good for rainrates in the 1.0 – 10.0 mm h^{-1} range. Note that the minimum detectable signal of the DPR (~0.2 mm h^{-1}, in terms of rainfall) and partial beam filling are responsible for scatterplot differences at very low rain rates. Contingency statistics for DPR NS rain detection reveal that for ground "reference" rain rates > 0.2 mm h^{-1} (the lower requirement threshold specified for DPR rain detection based on radar sensitivity), yield a DPR Probability of Detection (POD) of 64%, False Alarm Rate of 9%, and Heidke Skill Score (HSS) of 37%.

GPM Mission Science Requirements (see sidebar) stipulate thresholds for detection, bias, and random error (Hou et al., 2013). For example, rain rate estimates should exhibit a bias and random error of ≤ 50% (25%) at rain rates of 1 mm h^{-1} (10 mm h^{-1}) for areas of 50 km x 50 km. Figure 7 is presented for the DPR Normal Scan (NS) product as a preliminary example of assessing bias and random error. For non-zero raining pixels in Figure 7, the bias in each reference rain bin is computed as the MRE in percent while for the random error the NMAE is computed with the systematic error (bias) removed. Figure 7 suggests that the above GPM Mission Science Requirements have been met for the DPR example shown and the method used. While these results are encouraging, work is ongoing to further test and refine methodologies for determining product-consistent lower rain rate thresholds for comparing GPM GPROF, CORRA, DPR, and MRMS datasets and for defining error types and to meet the other GPM Mission Science Requirements.

Initial Scientific Investigations and Applications

With two years worth of calibrated and validated precipitation estimates, GPM's data are being used for scientific studies (e.g., Liu and Liu 2016, Wentz and Meissner 2016, Panegrossi et al. 2016, and Prakash et al. 2016). Most of the science results are from investigations by members of the NASA Precipitation Measurement Missions science team (in 2016 consisting of 60 Principal Investigators from NASA centers and U.S. universities funded by NASA Headquarters while the Japanese PMM Science Team consists of 41 Principal Investigators). NOAA has a team of 16 investigators involved with GPM and more than 20 international no-cost teams also play important roles in GPM science and validation efforts. Herein, two scientific investigations are reported: falling snow retrievals and monsoon studies.

Scientifically, retrievals of falling snow from space represent an important data set for understanding the Earth's atmospheric, hydrological, and energy cycles. While satellite-based remote sensing provides global coverage of falling snow events, the science is relatively new and retrievals are still undergoing development addressing challenges such as those listed in Skofronick-Jackson et al. (2015). GPM's mission goal of estimating falling snow is demonstrated in an example from March 17, 2014, just 18 days after launch (Figure 1c). More generally, the GMI observed the average snow rate, maximum snow rate, and fraction of precipitation that fell as snow over the winter of 2014-2015 (Figure 8). While these snow estimates are not fully validated they do support the requirement that GPM detect falling snow. The high rates over the south-central states may not be representative of typical winter conditions, but may have resulted from the occurrence of several heavy snow events in mid-late February of 2015 when GMI had good overpasses. Of particular note for this period were the large snowfall rates along the west coast of Canada and southern coast of Alaska, where coastal topography may enhance local snowfall rates.

Looking elsewhere, the GPM mission can track the advance and retreat of India's annual monsoon and the tropical storms that impact India's populations. As shown in Figure 9, GPM observes the detailed structure of the copious monsoon precipitation as it marches from south to north across India over the seasons, with Tropical Cyclone Hudhud (Oct 2014) on the left and Storm Roamu (May 2016) on the far right of the timeline. Figure 9 shows the advance of the monsoon season from offshore in May to inland by July, and the retreat back to the Bay of Bengal from September to November over two years of GPM data. Over longer precipitation records, interannual variations due to the effect of large-scale oceanic or atmospheric patterns or to climate change may be identified, information that is crucial for societal applications and benefit.

Integrating satellite observations into land surface modeling systems is a critical component of how to resolve the state of the water cycle and stresses on the system during extreme events. The NASA Land Information System (LIS; Kumar et al. 2006, Peters-Lidard et al. 2007) runs operationally at the Short-term Prediction Research and Transition (SPoRT, 2016) Center (Jedlovec 2013, Zavodsky et al. 2013, Case et al. 2016) at NASA's Marshall Space Flight Center (Xia et al. 2012, Zhang et al. 2016, Vargas et al. 2015) to produce analyses and short term forecasts of soil moisture and other fields. LIS is a land surface modeling and data assimilation framework designed to integrate satellite observations, including GPM and the Soil Moisture Active Passive (SMAP) satellite data (Entekhabi et al., 2010) into the modeling infrastructure (http://lis.gsfc.nasa.gov/). The integration of GPM data within LIS, run operationally at SPoRT, can capture soil moisture changes. For example, LIS identified an extreme soil moisture increase the first week of October 2015 when a closed upper low over the Southeastern U.S. combined with a deep tropical moisture plume associated with Hurricane Joaquin, led to historic rainfall over the Carolinas. The SPoRT Center provided model outputs from LIS to Eastern Region NWS forecast offices in near-real-time. In other cases, these data are also used by a variety of end users experimentally for assessing drought, flooding potential, and situational awareness for wildfire and blowing dust. There is great potential in the future for using GPM estimates together with other space-based soil-moisture measurements from SMAP to improve weather and hydrological prediction.

The GPM suite of products contributes to a wide range of societal applications such as: tropical cyclone location and intensity, famine early warning, drought monitoring, water resource management, agriculture, numerical weather prediction, land system modeling, global climate modeling, disease tracking, economic studies, and animal migration; many of which were initially developed with TRMM data. Many of these applications require near-real-time data as well as longer-term, well-calibrated precipitation information. IMERG is starting to be used as an input for forecasts in other regions of the world, especially areas lacking adequate ground-based coverage. Selected applications are reported in Kirschbaum et al. (2016), Ward et al. (2015), Kucera et al. (2013), and Kirschbaum and Patel (2016).

Conclusions and Next Steps

The Global Precipitation Measurement mission provides unprecedented and highly useful global precipitation datasets. GPM's Core Observatory data are used to inter-calibrate a set of precipitation observations from constellation partner sensors. By merging GPM multi-satellite estimates with other IR satellite data, products with temporal resolutions down to 30 minutes and spatial resolutions as small as 0.1^{o} by 0.1^{o} are possible. Latencies, at 1-5 hours (depending on the product) after data collection, are vital for GPM's operational users. Research quality products (with accuracy requirements as indicated in the Sidebar GPM's Mission Science Requirements) are available later (12 hrs to several months) for intensive scientific studies ranging from diagnosing microphysical precipitation particle characteristics to assessing regional and global patterns of precipitation. The GPM mission provides indispensable precipitation data from micro to local to global scales via retrieved precipitation particle size distributions inside clouds, 5-15 km resolution estimates of regional precipitation, and merged global precipitation.

GPM's algorithms have been updated several times (currently on Version 04) with an additional update planned for 2017. After the release of Version 05, work will begin to reprocess Level 0-3 products back to the beginning of TRMM (1998) and also for partner satellite data sets to establish a long and consistent record of precipitation. Scientific studies and societal applications using GPM data are ongoing and growing rapidly. Knowing the horizontal and vertical structure of precipitation is important for improving weather forecasting and climate change models. The planned processing of a consistent precipitation record encompassing the TRMM and GPM era will be of high value to future generations of scientific studies and user applications. The consistent TRMM-plus-GPM record will generate interesting scientific insights and re-invigorate applications in hydrological/land surface modeling and numerical weather prediction. Going forward in time, GPM's prime mission lifetime lasts until May 2017 at which time GPM will move into Extended Operations. Current predictions suggest that the station-keeping fuel will last 15 or more years, implying that instruments or spacecraft systems (like the batteries) will likely be the life-limiting factors as long as the fuel requirements do not increase.

In quantifying precipitation, a key Earth system component, the GPM mission provides fundamental knowledge of the water cycle and compliments other NASA satellite missions such as the Gravity Recovery and Climate Experiment (GRACE), that measures changes in groundwater levels in underground aquifers (among other observations) (Tapley et al. 2004); the Soil Moisture Active Passive (SMAP) satellite (Entekhabi et al. 2010); Aquarius (while it was operating), that observed ocean salinity (Le Vine et al. 2010); and CloudSat, which measures the properties of clouds and light precipitation (Stephens et al. 2002). Integrated multidisciplinary scientific investigations can provide greater understanding of our complex Earth system. GPM has and will continue to provide valuable and freely accessible precipitation data for science and society.

Acknowledgments

This paper is dedicated to former GPM Project Scientist Arthur Y. Hou (1947-2013). Data are provided by NASA / JAXA. The climatological monsoon dates are from the Indian Meteorology Department. Imagery was generated by NASA Goddard Space Flight Center, including the monsoon visualization by Owen Kelley. PMM Science Team members are acknowledged for their many contributions to GPM science. We thank our three anonymous reviewers for thoughtful and paper-enhancing comments. Funding for this work was provided by NASA Headquarters, NOAA, and JAXA.

Sidebar 1: GPM's Mission Science Requirements

Prior to the GPM's launch in 2014, NASA formally documented Core Observatory requirements to be met within GPM's 3-year Prime Mission operations period in order for GPM to be deemed fully successful. Several of these requirements dealt with instrument performance or operational elements (e.g., orbit maintained to within ±1 km of operational orbital attitude) and will not be discussed here. Most of the requirements pertained to scientific accuracy and science data and are key to ensuring stable and validated precipitation products expected by both scientific investigators and application users. Specifically, these science requirements are:

- Measurements of the same geophysical scenes using both active and passive technique from 65^{o}N to 65^{o}S latitude with mean sampling time of 24 hours
- Using the DPR:
    - Quantify rain rates between 0.22 and 110.00 mm h^{-1}
    - Detect snowfall at an effective resolution of 5 km
- Using the GMI
    - Quantify rain rates between 0.2 and 60.0 mm h^{-1}
    - Detect snowfall at an effective resolution of 15 km
- Estimate precipitation particle size distribution (e.g., quantitative estimates of precipitation microphysical properties such as the mean median mass diameter of particle size distribution to within ±0.5 mm.)

- Provide calibrated ground-based precipitation measurements and associated error characterizations at 50 km horizontal resolution for comparison with space-based radar and radiometer measurements at designated ground validation sites within ground tracks of the GPM Core Observatory.
  - The biases in instantaneous rain rates between the ground-based and space-based estimates should not exceed 50% at 1 mm h^{-1} or 25% at 10 mm h^{-1}
  - The random errors between the ground-based and space-based estimates should not exceed 50% at 1 mm h^{-1} or 25% at 10 mm h^{-1}.
- In order to provide data in near-real-time for hurricane monitoring, numerical weather prediction, hydrological model forecast and other application and operational uses:
  - Combined radar/radiometer swath products will be available within 3 hours of observation time, 90% of the time, and
  - Radiometer precipitation products will be available within 1 hour of observation time, 90% of the time.

At the time of the writing of this article all science requirements have been shown to have been met but have not been documented in the literature. Several papers are being prepared on proving these requirements and will be included in the AMS Special Collection of GPM Publications.

Acronym List

AMSR2	Advanced Microwave Scanning Radiometer for Earth Observing System 2
ASCII	American Standard Code for Information Interchange
ATMS	Advanced Technology Microwave Sounder
CNES	Centre National d'Etudes Spatiales
ISRO	Indian Space Research Organisation
CONUS	Continental US
CORRA	Combined Radar-Radiometer Algorithm
dBZ	decibel relative to Z
DMSP	Defense Meteorological Satellite Program
DPR	Dual-frequency Precipitation Radar
EUMESAT	European Union Meteorological Satellites
FOV	Field of View
4D	Four-dimensional
GCOM-W1	Global Change Observation Mission - Water
GCPEx	Global Precipitation Measurement Cold Season Precipitation Experiment
GIS	Geographic Information System
GHz	Gigahertz
GMI	GPM Microwave Imager
GPCP	Global Precipitation Climatology Project
GPM	Global Precipitation Measurement
GPM-CO	Global Precipitation Measurement Core Observatory

GPROF Goddard Profiling Algorithm

GRACE Gravity Recovery and Climate Experiment

GSFC Goddard Space Flight Center

GSMaP Global Satellite Mapping of Precipitation

GV Ground Validation

HDF5 Hierarchical Data Format

IMERG Integrated Multi-satellitE Retrievals for GPM

IR Infrared

JAXA Japan Aerospace Exploration Agency

JPSS1 Joint Polar Satellite System-1

JST Japan Standard Time

LIS Land Information System

MAE Mean Absolute Error

MHS Microwave Humidity Sounder

MHz Megahertz

MOS Mission Operations System

MRE Mean Relative Error

MRMS Multi-Radar/Multi-Sensor

MSFC Marshall Space Flight Center

NASA National Aeronautics and Space Administration

NEDT Noise Equivalent Delta Temperature

NEXRAD Next-Generation Radar

NOAA	National Oceanic and Atmospheric Administration
NPP	NASA Postdoctoral Program
NRT	Near-Real-Time
NS	Normal Scan
NWS	National Weather Service
PMM	Precipitation Measurement Missions
PPS	Precipitation Processing System
QPE	Quantitative Precipitation Estimates
SAPHIR	Sounder for Probing Vertical Profiles of Humidity
SMAP	Soil Moisture Active Passive
SPoRT	Short-term Prediction Research and Transition
SSMIS	Special Sensor Microwave Imager/Sounder
TB	Brightness Temperature
TIFF	Tagged Image File Format
TMI	TRMM Microwave Imager
TRMM	Tropical Rainfall Measuring Mission
3D	Three-Dimensional
U.S.	United States
UTC	Coordinated Universal Time
WMO	World Meteorological Organization
WSR-88D	Weather Surveillance Radar 88 Doppler

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Table 1: Channel availability by frequency and polarization (V=Vertically polarized, H=Horizontally polarized) for the GPM constellation radiometers. GMI, TMI, AMSR2, and SSMIS are all conically scanning imagers while MHS, ATMS, and SAPHIR are cross-track scanning water vapor sounders. ATMS is currently operating on board Suomi NPP with a second copy to launch on board JPSS1 in mid 2017.

Sensor	Satellite	6-7 GHz	10 GHz	18-19 GHz	21-23 GHz	31-37 GHz	85-92 GHz	150-166 GHz	183 GHz
GMI	GPM		10.65 VH	18.7 VH	23.8 V	36.64 VH	89.0 VH	166 VH	183.31 V ±3, ±7
TMI	TRMM		10.65 VH	19.35 VH	21.3 V	37.0 VH	85.5 VH		
AMSR2	GCOM-W1	6.925 VH 7.3 VH	10.65 VH	18.7 VH	23.8 VH	36.5 VH	89.0 VH		
SSMIS	DMSP F16, F17, F18, F19			19.35 VH	22.235 V	37.0 VH	91.655 VH	150 H	183.31 H ±1, ±3, ±6.6
MHS	NOAA-18/19, MetOp-A/B						89 V	157 V	183.31 H ±1, ±3, 190.31V
ATMS	Suomi NPP, JPSS1				23.8 V	31.4 V	88.2 V	165.5 H	183.31 H ±1, ±1.8, ±3, ±4.5, ±7
SAPHIR	Megha-Tropiques								183.31 H ±0.2, ±1.1, ±2.8, ±4.2, ±6.8, ±11

Table 2: Area weighted mean annual precipitation in mm day^{-1} for each of the algorithms globally, over land, and over ocean from +/- 50 degrees latitude.

	Global Mean Daily Precipitation	**Oceanic Mean Daily Precipitation**	**Land Mean Daily Precipitation**
DPR	2.51	2.77	1.72
GPROF	2.86	2.99	2.36
Ku	2.81	3.03	2.05
CORRA	2.83	2.85	2.77
IMERG	2.48	2.44	2.39
GPCP	2.95	3.15	2.43
GSMaP	2.74	2.83	2.12

Figure 1: GPM-CO GMI composite brightness temperatures and example precipitation event cases. Center panel: composite 89 GHz brightness temperatures averaged over 24 months showing the latitudinal extent of the GPM-CO measurements. Example precipitation cases (a) A North Pacific frontal system from GMI, (b) Severe storms in Texas from GMI, (c) winter storm over the Eastern U.S. as observed in 3D from the DPR, (d) North Atlantic winter storm from GMI, (e) Typhoon Fantala as observed in 3D from the DPR, (f) Typhoons Chan-Hom and Nangka in two successive orbits from GMI, (g) a South Pacific frontal system from GMI, (h) a South Atlantic frontal system from GMI, (i) a line of convection in Africa in 3D from the DPR, (j-k) Sumatra land/sea convection day and night from GMI, and (l) an Australian weather system from GMI.

Figure 2: Integrated Multi-satellitE Retrievals for GPM (IMERG) accumulated precipitation totals from 4-11 August 2014. The IMERG retrieval algorithm has not yet been developed for pole-to-pole retrievals. The large accumulation near Japan is Typhoon Halong. The accumulation also shows a major storm over the North Sea near Europe, the origins of Hurricane Gonzalo on the western coast of Africa, and a deep tropical depression that produced floods across northern India. IMERG gridded products are produced every 30 minutes with 0.1^{o} x 0.1^{o} grid boxes, currently covering the latitude band 60°N-S.

Figure 3: Precipitation estimates are shown for a single orbit of each of the GPM constellation radiometer types for January 1, 2015. The conically-scanning window-channel radiometers are shown on the left and the cross-track scanning water vapor sounding radiometers are shown on the right. The constellation radiometers include a) TMI and GMI on board the NASA TRMM and GPM satellites, b) ATMS on board NOAA's Suomi NPP satellite, c) AMSR2 on board

JAXA's GCOM-W1 satellite, d) SAPHIR on board the CNES-ISRO Megha-Tropiques satellite, e) SSMIS on board the DMSP F16, F17, F18 and F19 satellites, and f) MHS on board the NOAA-18, NOAA-19, and EUMETSAT MetOp-A and Metop-B satellites.

Figure 4: GPM mission operations data and communication system. GPM-CO satellite data are downlinked in near-real-time via the NASA Tracking and Data Relay Satellite System (TDRSS) to White Sands, New Mexico, where the GPM Mission Operations Center retrieves it, ensures its integrity and passes it to PPS. Partner data, ancillary information and validation measurements are also processed by mission operations.

Figure 5: Zonal precipitation averages (in mm day^{-1}) for the full annual cycle in 2015. The five estimates are: GPM DPR (dual-frequency radar in red), GPM GPROF (GMI passive radiometer in blue), GPM Ku (single-frequency radar in green), GPM CORRA (DPR+GMI in orange), IMERG (GPM merged with constellation estimates in purple), GPCP global estimates (in light blue), and MCTA2 estimates over ocean (in black, covering the years 2007-2010). The GPCP is Version 2.3, MCTA is Version 2, IMERG is Version 03, and the other GPM products are Version 04.

Figure 6: Density scatterplot of DPR-Normal Scan V04 versus reference MRMS precipitation (mm h^{-1}) at the footprint scale over the period June 2014 - August 2015. The 1:1 line (solid line) is displayed as well as the detection limit for the DPR (0.22 mm h^{-1}). The data shown focuses on

the conditional case of satellite footprint and reference mean precipitation rates both nonzero (> 0.01 mm h^{-1}), and a precipitation type of liquid only.

Figure 7: Conditional DPR V04 bias (MRE; solid black line) and random error (mean absolute error; dashed black line) versus the MRMS reference precipitation rate (mm h^{-1}) at 50 km resolution over the period June 2014 - August 2015 and normalized by the bin mean rain rate. Points falling outside of the 5%-95% inter-quantile range (outliers) were not included in this comparison. The dashed red lines indicate the GPM Mission Science Requirements 50% (25%) at the specified precipitation rates of 1.0 (10.0) mm h^{-1}.

Figure 8: The (a) average and (b) maximum liquid equivalent snowfall rates, and (c) fraction of precipitation that was identified as falling snow (and not liquid rain) from December 2014 – February 2015 from the GMI GPROF (Version 04) retrieval algorithm.

Figure 9: GPM depicts characteristics of India’s monsoon seasons in 2014 and 2015. The time-latitude figure (main panel) summarizes the IMERG precipitation estimates over India from April 2014 through May 2016. The heavy, black, dashed line shows the climatological advance and retreat of India's monsoon. The dates of the climatological advance and retreat are shown also on the two maps on the upper left. The area over which IMERG was averaged is indicated by the blue-gray rectangle stretching across India and the Bay of Bengal; the latitude on the main panel is along the mid-line of the rectangle, and the averages are taken along the perpendiculars to the mid-line.

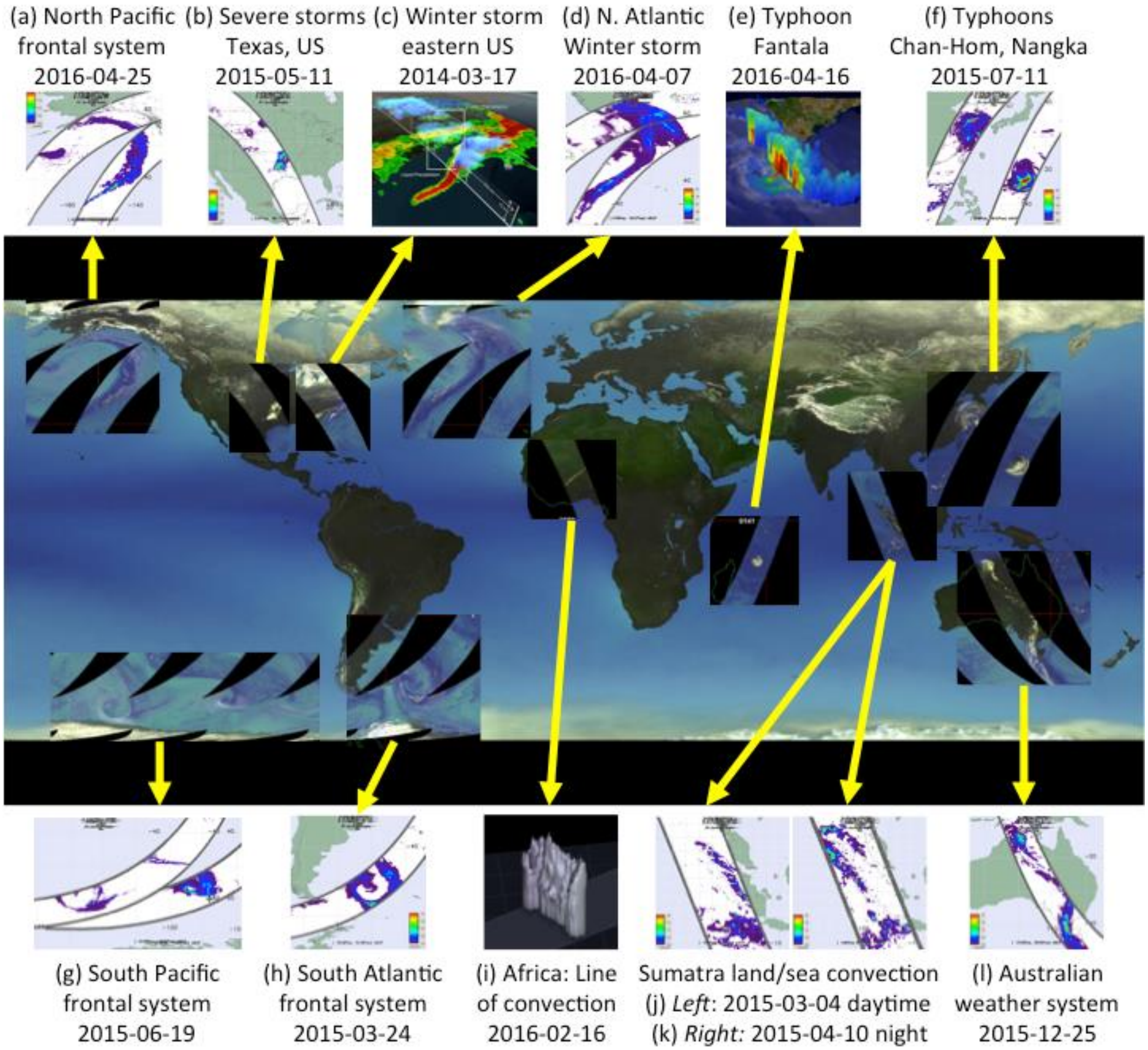


Figure 1: GPM-CO GMI composite brightness temperatures and example precipitation event cases. Center panel: composite 89 GHz brightness temperatures averaged over 24 months showing the latitudinal extent of the GPM-CO measurements. Example precipitation cases (a) A North Pacific frontal system from GMI, (b) Severe storms in Texas from GMI, (c) winter storm over the Eastern U.S. as observed in 3D from the DPR, (d) North Atlantic winter storm from GMI, (e) Typhoon Fantala as observed in 3D from the DPR, (f) Typhoons Chan-Hom and Nangka in two successive orbits from GMI, (g) a South Pacific frontal system from GMI, (h) a South Atlantic frontal system from GMI, (i) a line of convection in Africa in 3D from the DPR, (j-k) Sumatra land/sea convection day and night from GMI, and (l) an Australian weather system from GMI.

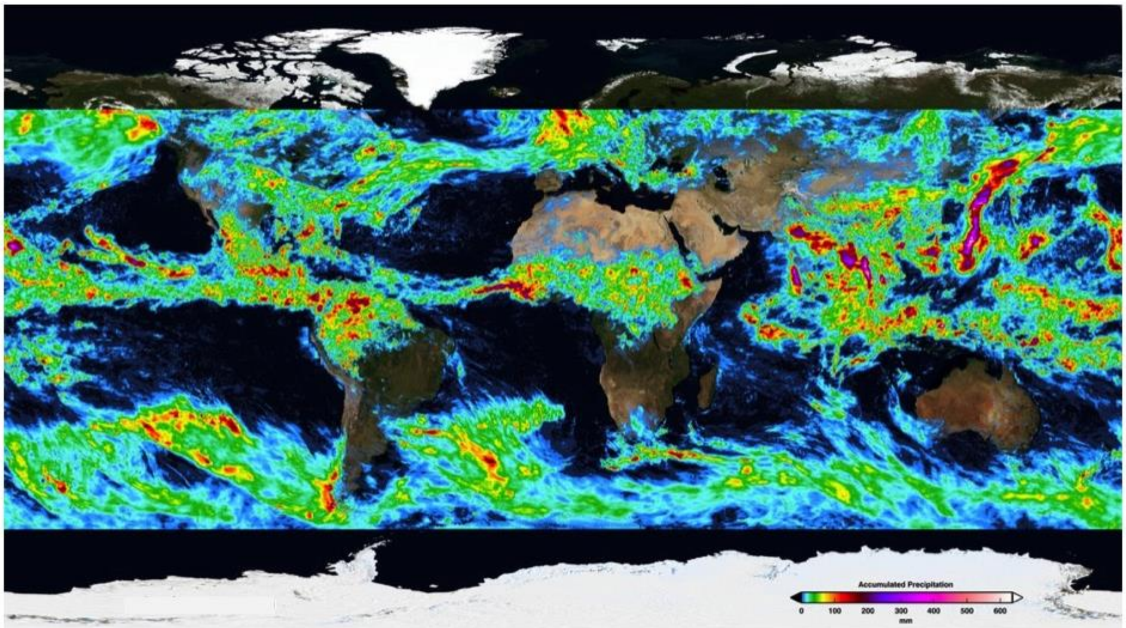


Figure 2: Integrated Multi-satellitE Retrievals for GPM (IMERG) accumulated precipitation totals from 4-11 August 2014. The IMERG retrieval algorithm has not yet been developed for pole-to-pole retrievals. The large accumulation near Japan is Typhoon Halong. The accumulation also shows major storm over the North Sea near Europe, the origins of Hurricane Gonzalo on the western coast of Africa, and a deep tropical depression that produced floods across northern India. IMERG gridded products are produced every 30 minutes with 0.1º x 0.1º grid boxes, currently covering the latitude band 60°N-S.

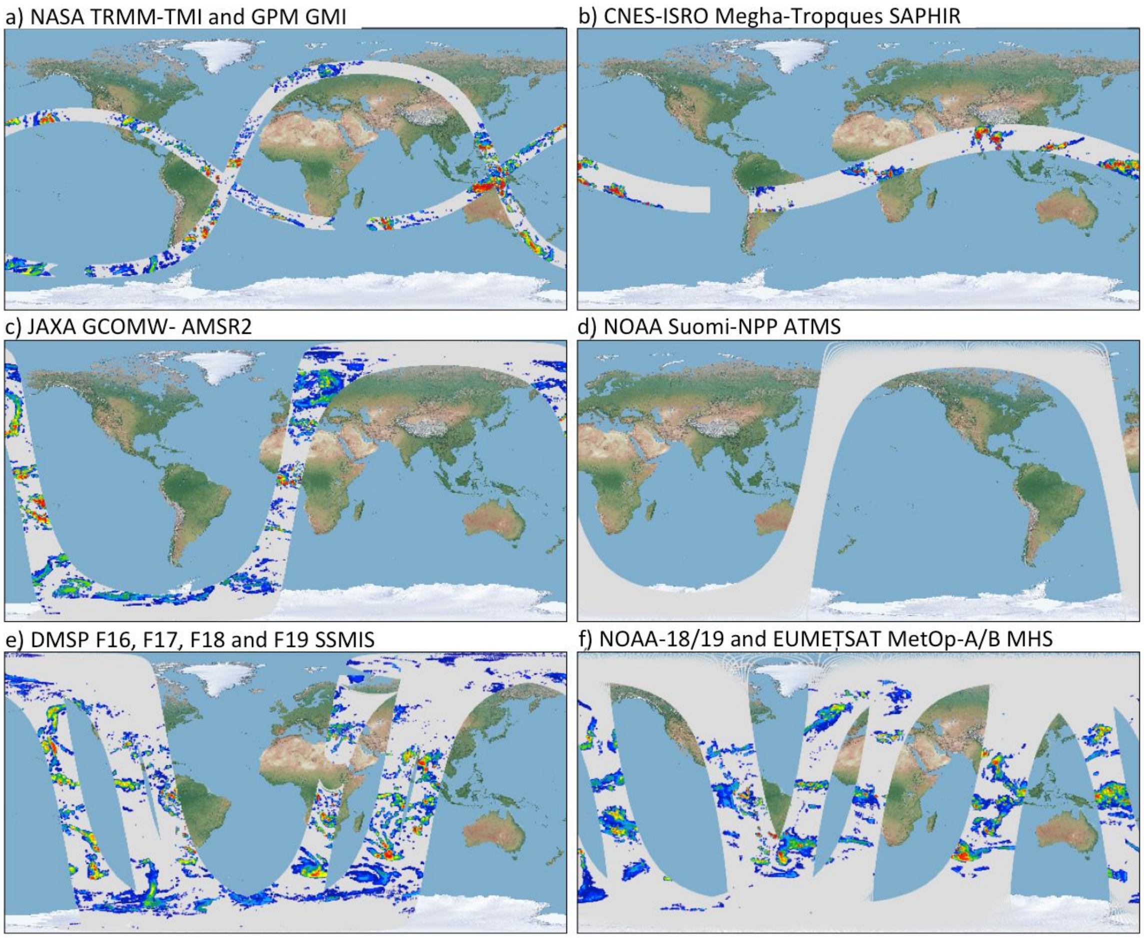


Figure 3: Precipitation estimates are shown for a single orbit of each of the GPM constellation radiometer types for January 1, 2015. The conically-scanning window-channel radiometers are shown on the left and the cross-track scanning water vapor sounding radiometers are shown on the right. The constellation radiometers include a) TMI and GMI on board the NASA TRMM and GPM satellites, b) ATMS on board NOAA's Suomi NPP satellite, c) AMSR2 on board JAXA's GCOM-W1 satellite, d) SAPHIR on board the CNES-ISRO Megha-Tropiques satellite, e) SSMIS on board the DMSP F16, F17, F18 and F19 satellites, and f) MHS on board the NOAA-18, NOAA-19, and EUMETSAT MetOp-A and Metop-B satellites.

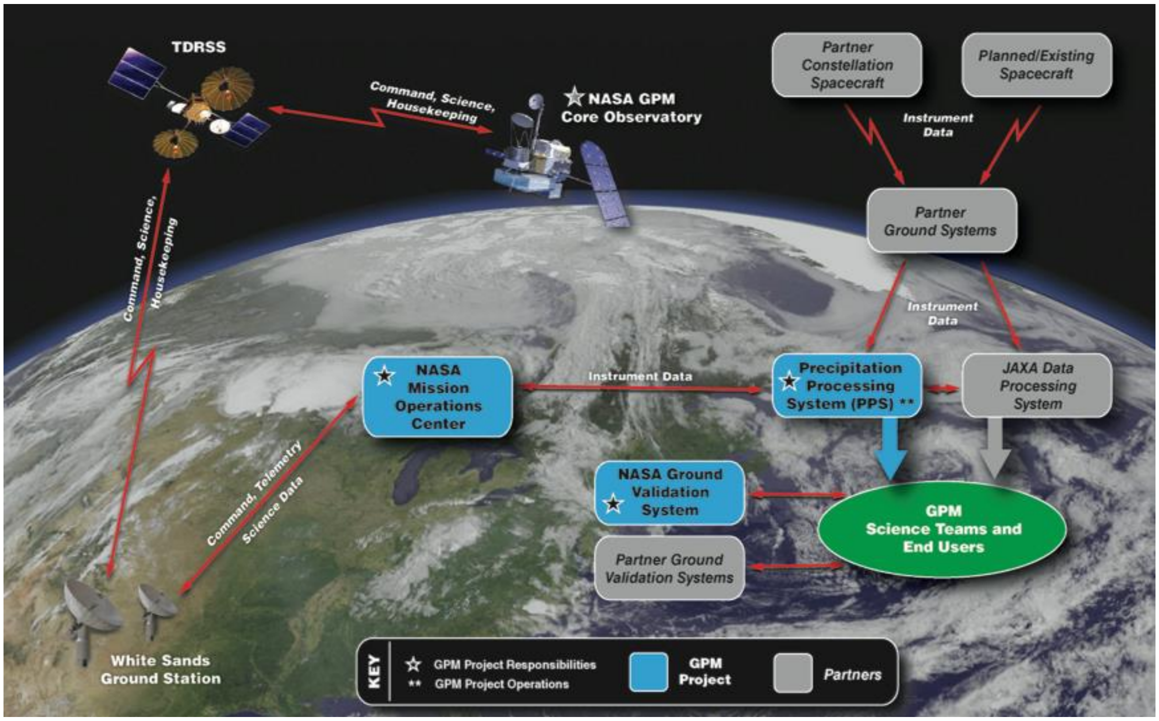


Figure 4: GPM mission operations data and communication system. GPM-CO satellite data are downlinked in near-real-time via the NASA Tracking and Data Relay Satellite System (TDRSS) to White Sands, New Mexico, where the GPM Mission Operations Center retrieves it, ensures its integrity and passes it to PPS. Partner data, ancillary information and validation measurements are also processed by mission operations.

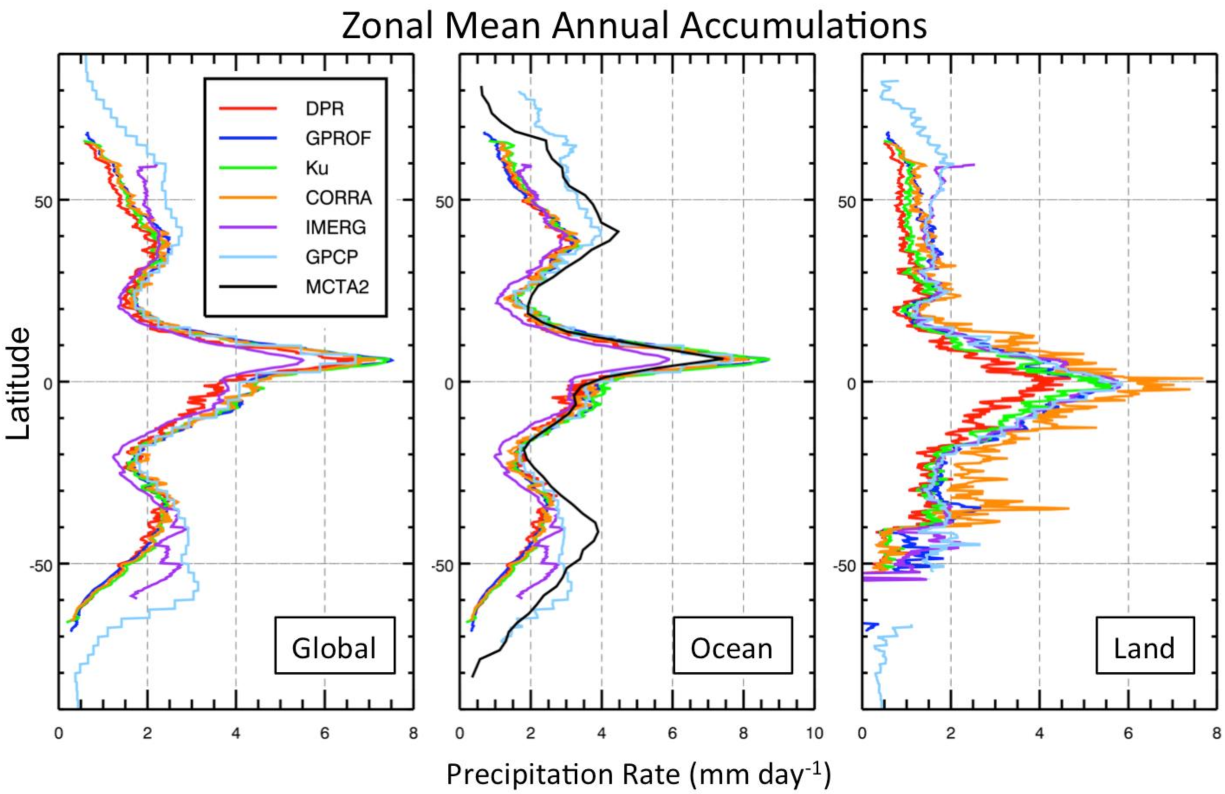
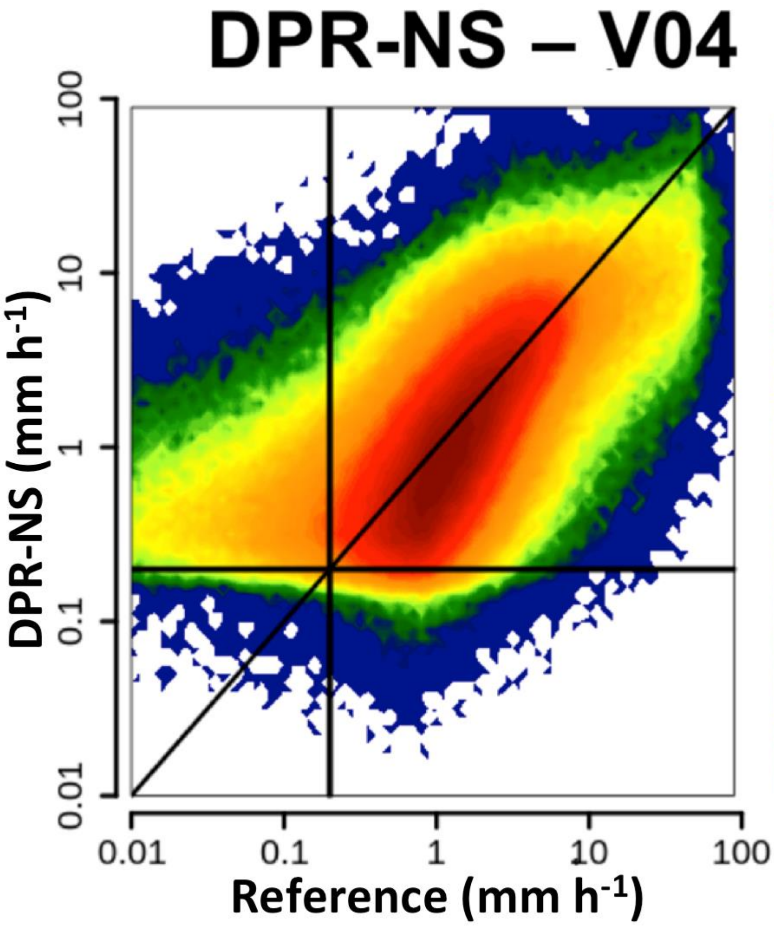


Figure 5: Zonal precipitation averages (in mm day^{-1}) for the full annual cycle in 2015. The five estimates are: GPM DPR (dual-frequency radar in red), GPM GPROF (GMI passive radiometer in blue), GPM Ku (single-frequency radar in green), GPM CORRA (DPR+GMI in orange), IMERG (GPM merged with constellation estimates in purple), GPCP global estimates (in light blue), and MCTA2 estimates over ocean (in black, covering the years 2007-2010). The GPCP is Version 2.3, MCTA is Version 2, IMERG is Version 03, and the other GPM products are Version 04.



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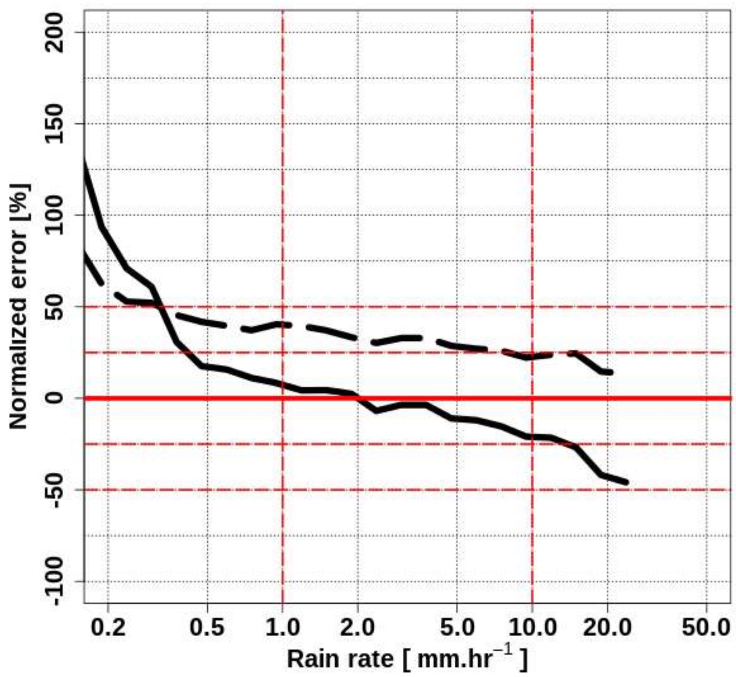

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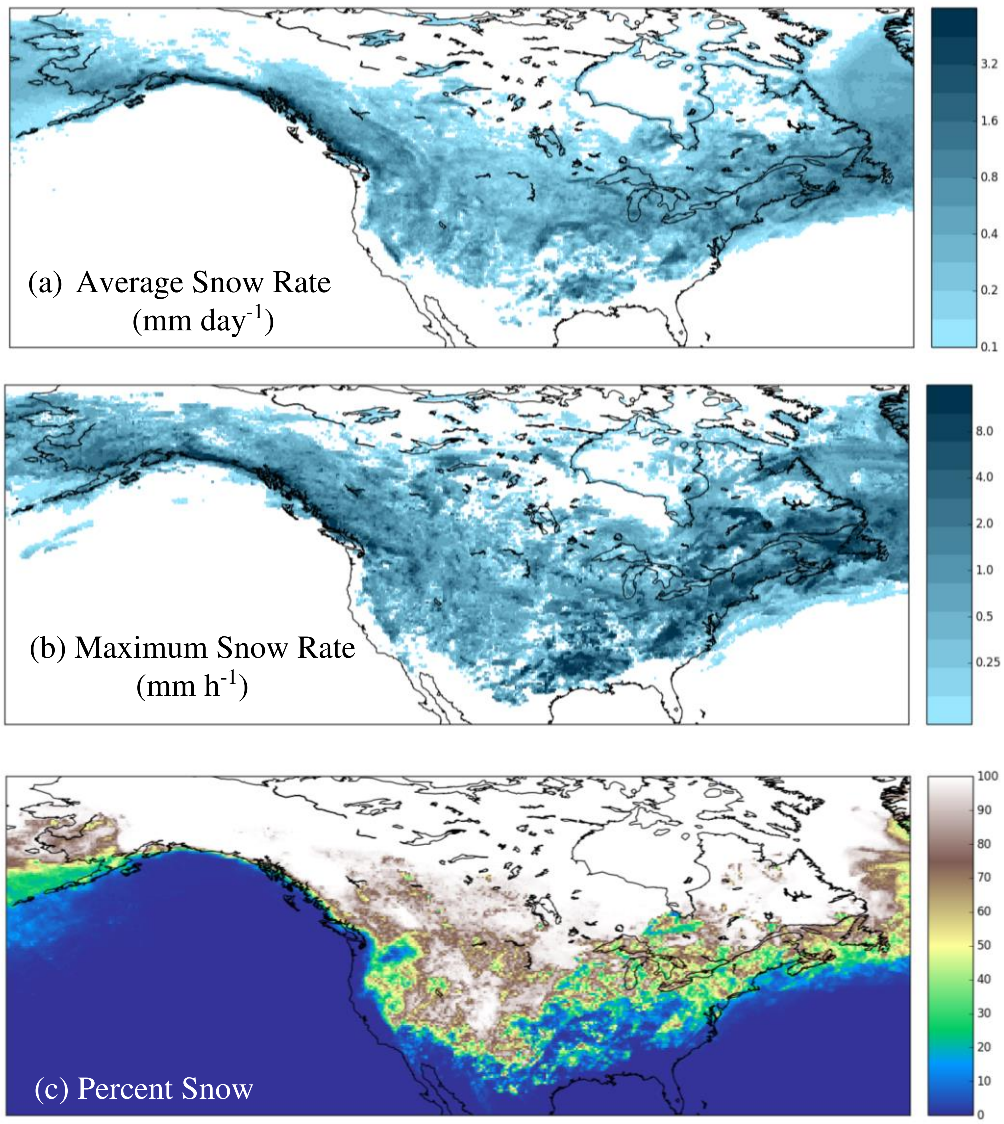

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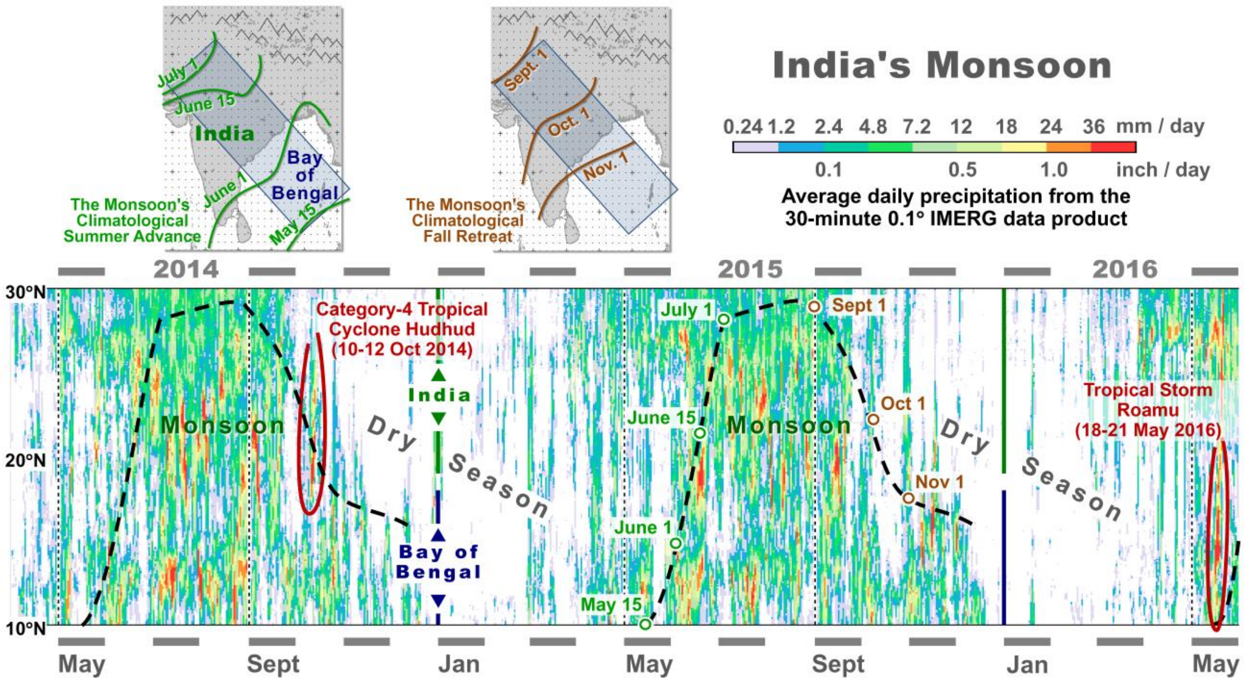

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