

Approach(es)

Construct deterministic (e.g., S(Z), S (KDP,Z), etc.) and probabilistic (PQPE) radar-based SWER with minimum bias (random error accepted for large sample). (1) Use multi-regime "reference" networks to obtain "best" footprint area estimators; (2) Assess/correct radar SWER biases relative to reference network and physical character of snowfall (3) Compare reference-based radar-diagnosed SWER to GPM satellite DPR and GMI estimates.

Hyytiälä, Finland. Case-specific Z-S constructed based on snow physical properties per von Lerber et al. 2017, 2018 (JAMC), applied to Ikaalinen (IKA) C-band radar, compared to GPM, winters 2014-18

Grid and average GV data within DPR / GMI pixel

Gridded GV height: 0.5 km Horiz Res: 1.0 km;Vertical Res: 0.25 km DPR/CMB: 5 x 5 km²; GMI: 25 x 25 km²

Generate precipitation rate data "Pairs" Snapshot data matched in time/space Multiple overpass dates combined.

GV scan and GPM overpass coincidence within 6 minutes (adjustable).

Generate plots (Bias/NMAE; Scatter; Density) Conditional Analysis: GV mean rate and Beam-Filling at 50% or 90%: GV data must filling at 50% or 90%: GV data must filling at 50% or 90% or 90% of the second s DPR/GMÍ pixel > 0 mm/hr DPR/GMI pixel at required %

2. Marguette (MQT) Reference Pluvio Network (RPN) and NWS measurement site. Ten GPM-GV Pluvio weighing gauges (single Alter fence) in 15-20 km footprint within 20 km of WSR-88D (KMQT) radar. Micro Rain Radar-2 (MRR) and Precipitation Imaging Package (PIP) installed at the NWS Forecast Office MQT. The RPN provides SWER "reference", MRR and PIP provide physics, KMQT provides "regional scale" distributed pixel measurements of SWER for comparison to GPM swath data.



3. Probabilistic QPE (PQPE) approach (Kirstetter et al, 2015, Water Resources Research). Capture range of Z-S behavior that minimizes bias while providing an estimate of error. Compare PQPE range of Z-S (e.g., 25th, Expected Value, 50th, 75th, percentiles) against those diagnosed in other regimes.



 Left: PQPE Z vs. SWER relationships (inset) for the Expected (S-EV, red, bold), 25th, and 75th percentile (bold black) plotted against CARE, Hyytiälä, and WFO-MQT local KMQT Z-S relationships. • Note all Z-S fit broadly within PQPE envelope. KMQT WFO relationship most similar to PQPE 25th% behavior.

 When used with KMQT radar and verified against the MQT RPN how well do the Z-S relationships perform?

4. Continental-scale comparisons via application of SWER estimates in the GPM GV radar Validation Network (VN). Assess PQPE, KMQT Z-S and polarimetric SWER(KDP,Z) SWER estimators against GPM satellite estimates.



Radars in VN network. Colors indicate percentage of GPM overpasses of VN radars with precipitation (solid or liquid).

Translating the Physics of Snowfall to Radar-Based Validation of GPM

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Objective: Validate GPM estimates of Snow Water Equivalent Rate (SWER) • Original GPM requirement is for "detection" of falling snow rates, but desire is to reliably estimate SWER

• Reliable satellite-based remote sensing of SWER is hard- but also the case for ground-based instrumentation (a "validation" source). • How do we assess agreement between distributed (i.e., multiple footprints) GV-radar and near instantaneous satellite-based SWER estimates?

IKA Ground Radar Reflectivity \rightarrow Derive mapped SWER



KA Snow Event Determine event-specific and snow density-tuned Z-S derived using Precipitation Imaging Package (PIP) and Pluvio Generate SWER field in polar

coordinates from calibration idjusted radar reflectivity



MQT RPN: Pluvio gauges (green dots) SWER "reference" sampling of KMQT estimators; Physics instruments located at MQT,

Dataset collection for winters 2017/18 and 2018/19 (2017/18 analyzed so far)





• We have developed an ensemble of tools/approaches designed to provide physically-tuned/consistent and/or statistically optimal radar-based SWER estimates to validate instantaneous GPM estimates of SWER from the pixel to swath scale. Comparison of the GPM products to reference radar-based SWER estimates over Finland and the U.S. suggest that GPM GMI-GPROF and DPR SWER estimates may be biased low.

data analysis support.

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Summary

• Ongoing and future analysis work will include expansion of the case database in Finland and Canada, continued evaluation of snow physical properties defining regime Z-S behavior, examination of footprint to sub-footprint scale SWER variability. Acknowledgements: NASA PMM (Dr. G. S.-Jackson) and GPM (Dr. S. Braun) funding. NOAA NWS MQT for GV instrument network support, Ms. M. Mateling, U. Wisconsin and Mr. C. Pabla (SSAI/WFF) for





Compared to case-specific IKA radar Z-S (tuned for snow density)

- GMI-GPROF biased low ~60%
- Radar-based products (dual or single frequency) also biased low ~55%
- Daily accumulated KMQT Radar SWER using S25% and SEV Z-S plotted against RPN (lowest 3tilts of KMQT volume scan-heights 100m - 700m
- PQPE SEV (similar to MRMS): +85%,
- PQPE 75% and polarimetric estimator (S[Z,
- KDP], not shown) are biased ~100% high. Local KMQT (not shown) Z-S, SMQT: +4%

• SEV PQPE Z-S is much higher for common events. Note that default MRMS Z-S

 DPR matched swath (MS) product: only slightly lower than "best" relationships (PQPE 25% and KMQT Z-S) in the KMQT radar domain. [Combined algorithm, similar, not shown].

- DPR Markedly lower for PQPE SEV and dual-pol KMQT (as expected based on RPN comparisons).
- Left: PQPE 25th% results (KMQT Z-S similar) for the Combined (left) and DPR-MS algorithms (right). Remainder of PQPE and dual-pol Z-S are not shown, but exhibit much higher values compared to DPR and GPM Combined.
- For CONUS VN radars DPR estimates of SWER are very similar to the 25th% PQPE Z-S, and KMQT Z-S (though noisy, both of these Z-S' exhibited the least bias compared to the RPN).