

living planet symposium BONN 23-27 May 2022

TAKING THE PULSE OF OUR PLANET FROM SPACE

EUMETSAT CECMWF



The ESA RainCast Study For Global Snowfall Monitoring: New Concepts And Perspectives in view of the Arctic Weather Satellite mission

> D. Casella¹ (Presenter) G. Panegrossi¹ P. Sanò¹ A. Camplani² A. Battaglia³ 1 National Research Council, Institute of Atmospheric and Climate Sciences (CNR-ISAC), Rome, Italy 2 Department of Civil, Constructional and Environmental Engineering (DICEA), Sapienza University of Rome, Italy 3 Department of Environment, Land and Infrastructure Engineering, Politecnico of Torino, Torino, Italy

ESA UNCLASSIFIED – For ESA Official Use Only

25 / 05 / 2022

PMW Remote Sensing of Snowfall: Challenges and synergistic approach



Detection and quantification of snowfall by passive microwave observations remains among the most challenging tasks in global precipitation retrieval

Challenges:

The *PMW spectral signature in presence of snowfall* is highly dependent on the complex scattering properties of snowfall

Snowfall scattering signal is weak and tends to be masked by the water vapor and supercooled cloud liquid water (SCLW) emission

Sea ice and snow-covered land surface emissivity is extremely variable due to rapid changes of sea ice properties of snow cover extent, snow accumulation on the ground, and snowpack and sea ice physical properties.

GMI 10 GHz H-Pol

200

65 N

63° N

57

65[°]

63° I

61[°] N

Frontal snowfall over Siberia 30 April 2014 GMI/CloudSat-Calipso co-located observations



Mostly affected by bakground sfc (and liquid precipitation emission over sea)

Panegrossi et al. 2017 Rem. Sens.

RainCast – Scientific Evaluation of Future Atmospheric Mission Concepts to Monitor Precipitation



(in response to ESA ITT TT 1-9324/18/NL/NA)

The study aims at identifying and consolidating the science requirements for a **European precipitation satellite mission** that could complement the existing space-based precipitation observing system

(fits the purposes of Earth Observation Science for Society https://eo4society.esa.int)

Phase 1 Closed on May 2021

GOAL (WP3200) Assessment of snowfall observation capability of spaceborne active and passive sensors

We focused on

CloudSat CPR and the two most advanced microwave radiometers currently available **GMI** (conically scanning) and **ATMS** (cross-track scanning) (in view of MWI and MWS for EPS-SG mission)

SLALOM (GMI) SLALOM-CT (ATMS)

- Algorithms for snowfall retrieval from PMW
- Based on ML
- Based on Observational data (Coincident CPR –Radiometer OBS.)





SLALOM-CT Snow retrievaL ALgorithm fOr gpM – Cross Track





SLALOM-CT: Coincidence dataset



Period	16/01/2014 – 31/08/2016	
Geographical area	82°S–82°N, 180°W–180°E	
Number of database points	6.5 M	
Number of database points with snowfall	1.1 M	
Horizontal resolution (Km)	15.8 x 15.8 (nadir) 30 x 68.4 (scan edge)	
INPUT Variables	Data source	
ATMS BTs		
ATMS Scan angle	NOAA	
Temperature @ 2m Total column integrated water vapor Freezing level Height Temperature profile Relative humidity profile Absolute humidity profile	ECMWF	

Supercooled Droplet

Snowfall Rate

Snow Water Path

DARDAR (raDAR/liDAR) LATMOS-

Reading Univ.

2C-SNOW-PROFILE (CPR product)

2C-SNOW-PROFILE (CPR product)



SLALOM-CT: Cold Surfaces Characterization



Passive Microwave Empirical Cold Surfaces Classification Algorithm (PESCA)

Characterization of surface emissivity is supposed to be very important for snowfall estimate

variable emissivity and light signals

PESCA algorithm identifies 4 classes of Snow over Land and 3 classes of Sea Ice

- at the time of the MW overpass
- using low frequency channels (almost unsensitive to ice clouds)
- Applicable to both Cross Track (ATMS) and Conically Scanning (GMI)



SLALOM-CT: Comparison with GPROF

- Almost insensitive to surface
- Detection capabilities largely better than GPROF
- Very accurate Estimates (compared to GPROF, using CPR as reference)

Surface	SNOW		Ocean		Land	
	GPROF	SLALOM- CT	GPROF	SLALOM- CT	GPRO F	SLALOM- CT
RMSE [mm/h]	0.18	0.10	0.24	0.09	0.24	0.10
ME [mm/h]	0.006	0.002	-0.006	-0.002	0.090	-0.01
Corr	0.55	0.80	0.04	0.84	0.47	0.79
POD	0.20	0.76	0.28	0.86	0.23	0.80
FAR	0.55	0.22	0.44	0.15	0.39	0.22
HSS	0.05	0.63	0.03	0.68	0.21	0.71



■ _ ■ ■ = = + ■ + ■ != = ■ || || || = # = = ■ || || || || ||

RainCast Phase 1 Final Recommendations



- > PMW products benefit from calibration/training by spaceborne radars:
 - Demonstrated benefits of using CloudSat/Calipso-based Machine Learning SLALOM approach (mostly for higher latitudes) -> Key role of EarthCare mission and NASA ACCP Radar
 - **G-band** radar can assist both in water vapour sounding and in cloud microphysics characterization.
- Frequency range: 19 to 190 GHz available in most advanced PMW radiometers (e.g., EPS-SG):
 - 150-166 GHz highly recommended-> Key role for shallow snowfall
 - The full spectrum of T and WV sounding is needed -> **Definition of environmental conditions (T, WV)**
 - Lower frequency channels -> characterization of the frozen background surface
 - 85-90 GHz channels -> key role in detection of supercooled liquid cloud droplets
- Conically scanning is preferable (dual pol., high resolution) but only EPS-SG MWI will be available -> Need to advance research on cross-track scanning radiometer capabilities (ATMS, MWS, AWS)





Offcially Started April 2022 !!

Objectives:

Arctic Weather Satellite (AWS) snowfall retrieval at high latitudes

- W-G band Doppler radar for ice/snow retrievals
- synergy between G-band radar and high-frequency radiometers.







Objectives:

Arctic Weather Satellite (AWS) snowfall retrieval at high latitudes

 impact of the lack of low frequency window channels (< 60 GHz) for snowfall detection capabilities

W-G band Doppler radar for ice/snow retrievals synergy between G-band radar and high-frequency radiometers.

💳 🔜 📲 🚍 💳 📥 📲 🔚 🔚 🔚 📰 👬 🔚 🔤 🛶 🔯 🛌 📲 🗮 🗮 📾 🖬 👘 🛨

ATMS vs AWS channels



	AWS			ATMS	
No.	Central frequency	Polarisation	No.	Central frequency (GHz)	Polarisation
			4	23.8	QV
			2	31.4	QV
1	50.3 GHz	QV	3	50.3	QH
			4	51.76	QH
2	52.8 GHz	QV	5	52.8	QH
3	53.246 GHz	QV	6	53.596 ± 0.115	QH
4	53.596 GHz	QV			
5	54.4 GHz	QV	7	54.4	QH
6	54.94 GHz	QV	8	54.94	QH
7	55.5 GHz	QV	9	55.5	QH
8	57.29 GHz	QV	10	f0 = 57.29	QH
			11	f0 ± 0.217	QH
			12	f 0 ± 0.3222 ± 0.048	QH
			13	f 0 ± 0.3222 ± 0.022	QH
			-14	f 0 ± 0.3222 ± 0.010	QH
			15	f0 ± 0.3222 ± 0.0045	QH
9	89 GHz	QV	16	89.5	QV
10	165.5 GHz	QV	17	165.5	QH
11	176.311 GHz	QV	18	183.31 ± 7.0	QH
12	178.311 GHz	QV	19	183.31 ± 4.5	QH
13	180.311 GHz	QV	20	183.31 ± 3.0	QH
14	181.311 GHz	QV	21	183.31 ± 1.8	QH
15	182.311 GHz	QV	22	183.31 ± 1.0	QH
16	325.15±1.2 GHz	QV			
47	325.15±2.4 GHz	QV			
18	325.15±4.1 GHz	QV			
19	325.15±6.6 GHz	QV			

Emissivity dependence on scan angle



Cross track scanners mixed polarization with scan angle Expected impact of QV-QH change should be further investigated $QV = ev \cdot cos^{2}(\theta) + eh \cdot sin^{2}(\theta)$ $QH = eh \cdot cos^{2}(\theta) + ev \cdot sin^{2}(\theta)$



SLALOM-CT Snowfall Detection: AWS and ATMS:





13

SLALOM-CT SWP Estimate: AWS and ATMS:



Comparison of SWP SLALOM-CT (ATMS and AWS) y axis and CPR x axis In the test dataset





SWP	ATMS	AWS
RMSE [kg/m ²]	0.050	0.061
R ²	0.86	0.80
ME[kg/m ²]	-1.6x10 ⁻⁵	1.3x10 ⁻⁴
CC	0.93	0.90





.[4

Case Study: Greenland 24 April 2016



Almost a Worst Case Scenario No sensible impact of low frequency channels





15



Objectives:

Arctic Weather Satellite (AWS) snowfall retrieval at high latitudes

- impact of the lack of low frequency window channels (< 60 GHz) for snowfall detection capabilities
- Analysis on the combined use of the 325 GHz -183.31 GHz channels to improve very light snowfall retrieval at high latitudes.

W-G band Doppler radar for ice/snow retrievals

synergy between G-band radar and high-frequency radiometers.

Simulation Strategy: AWS and G-Band Radar



→ THE EUROPEAN SPACE AGENCY

17



Objectives:

Arctic Weather Satellite (AWS) snowfall retrieval at high latitudes

- Sinergy with other radiometers for background surface characterization (CIMR, MWI)
- Exploitation of the full spectrum of the AWS T and WV sounding channels to reduce the dependence on model-derived auxiliary Variables
- Analyze the combined use of the AWS channels to improve supercooled water droplets detection.

W-G band Doppler radar for ice/snow retrievals synergy between G-band radar and high-frequency radiometers.

Some References



Thank you

- Sanò P., Casella D., Camplani A., D'Adderio L.P, Panegrossi G.(2021). Intercomparison of Machine Learning Techniques within the SLALOM-CT algorithm for snowfall retrieval from ATMS microwave measurements. *Remote Sensing*, 2022, 14.6: 1467.
- Camplani, A., Casella, D., Sanò, P., & Panegrossi, G. (2021). The Passive microwave Empirical cold Surface Classification Algorithm (PESCA): Application to GMI and ATMS. *Journal of Hydrometeorology*, 22(7), 1727-1744.
- Panegrossi, G., Casella, D., Sanò, P., Camplani, A., & Battaglia, A. (2022). Recent advances and challenges in satellite-based snowfall detection and estimation. In *Precipitation Science* (pp. 333-376). Elsevier.
- Turk, F. J., Ringerud, S. E., You, Y., Camplani, A., Casella, D., Panegrossi, G., ... & Peters-Lidard, C. (2021). Adapting passive microwave-based precipitation algorithms to variable microwave land surface emissivity to improve precipitation estimation from the GPM constellation. *Journal of Hydrometeorology*, 22(7), 1755-1781.
- Mroz, K., Montopoli, M., Battaglia, A., Panegrossi, G., Kirstetter, P., & Baldini, L. (2021). Cross validation of active and passive microwave snowfall products over the continental United States. *Journal of Hydrometeorology*, 22(5), 1297-1315.
- Milani, L., Kulie, M. S., Casella, D., Kirstetter, P. E., Panegrossi, G., Petkovic, V., ... & Skofronick-Jackson, G. (2021). Extreme lake-effect snow from a GPM microwave imager perspective: Observational analysis and precipitation retrieval evaluation. *Journal of Atmospheric and Oceanic Technology*, *38*(2), 293-311.
- Turk, F. J., Ringerud, S. E., Camplani, A., Casella, D., Chase, R. J., Ebtehaj, A., ... & Wood, N. B. (2021). Applications of a CloudSat-TRMM and CloudSat-GPM satellite coincidence dataset. *Remote Sensing*, *13*(12), 2264.
- Battaglia, A., & Panegrossi, G. (2020). What can we learn from the CloudSat radiometric mode observations of snowfall over the ice-free ocean?. Remote Sensing, 12(20), 3285.
- Rysman, J. F., Panegrossi, G., Sano, P., Marra, A. C., Dietrich, S., Milani, L., ... & Edel, L. (2019). Retrieving surface snowfall with the GPM Microwave Imager: A new module for the SLALOM algorithm. *Geophysical Research Letters*, *46*(22), 13593-13601.