



## The NOAA Microwave Integrated Retrieval System (MiRS) Validation Update and Applications

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- Introduction of MiRS
- Evaluation of MiRS algorithm performance (NOAA-20 ATMS)
  - : T and WV
- MiRS Users and applications
- Improvements in MiRS
- Future plan
- Path Forward

## Algorithm Overview



- MW Only, Variational Approach: Find the "most likely" atm/sfc state that: (1) best matches the satellite measurements, and (2) is still close to an a priori estimate of the atm/sfc conditions.
- "Enterprise" Algorithm: Same core software runs on all satellites/sensors; facilitates science improvements and extension to new sensors.
- Initial capability delivered in 2007. Running v11.2 since Jan 2017 on SNPP/ATMS, N18, N19, MetopA, MetopB, F17, F18, GPM/GMI, Megha-Tropiques/SAPHIR, (eventually MetopC...)
- Transition of v11.3 (extended to NOAA-20/ATMS) to operations in March 2019.
- V11.4 to operations in July 2019.
- External Users/Applications: TC Analysis/Forecasting at NHC, Blended Total/Layer PW Animations at NHC and WPC Animations (CSU/CIRA, U. Wisconsin/CIMSS), CSPP Direct Broadcast (U. Wisconsin), NFLUX model (NRL, Stennis), Global blended precipitation analysis at NOAA/CPC (CMORPH).

## Dropsonde, GFS 6h Forecast, and MiRS Retrieval



Within 30 minute ranges between Dropsonde, GFS, and Mirs retrieval



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#### Temperature Profile Validation based on ECMWF data



Same instrument: ATMS **Only 50 minute apart** 

180

180

10

#### Temperature Profile Validation based on sonde data



#### Water Vapor Profile (g/kg) Validation based on ECMWF data



#### Water Vapor Profile (%) Validation based on sonde data

All conditions and surface types



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# MiRS Users

- Many users use MiRS products for operational applications and research.
- Operational users, who have subscriptions on PDA, are:
  - NWS NHC, NESDIS\_VIZLAB, CLASS, STAR, NASA GPM, NAVO, FNMOC, 557TH, NASA JPL, GNC-A, NWC, CHINA CMA, NWS AWIPS, NWS NCO, CIRA, NASA SPORT, SSEC/CIMSS, NESDIS operational applications: eTRaP, blended TPW, blended RR, IMS and TC,
  - NHC TC intensity (POC: Galina Chirokova (CIRA))
  - International Arctic Research Center, Alaska (POC: Carl Dierking, NOAA Liaison)
  - NOAA Climate Prediction Center (POC: Pingping Xie)
  - National Ice Center
  - NWS Weather Forecast Offices
- CSPP generates MiRS products for direct broadcasting
- Users also download MiRS from CSPP and generate MiRS products at their own machines

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- Motivation:
  - MiRS data currently used in the operational TC Intensity Algorithm (HISA, developed at CIRA). Utilizes T and WV sounding to estimate warm core structure combined with statistical/dynamic model to predict future intensification.
  - Challenge: (1) retrieval of warm core structure complicated due to presence of hydrometeors; scattering signal in TBs can interfere with retrievals (2) hurricane warm core structure is anomalous relative to "global climatology" currently used as a priori constraint in MIRS.
- Development of a MiRS-TC, a version of MiRS optimized for tropical cyclones, adapting several components:
  - Modify use of higher frequency channels in scenes likely to have large amounts of scattering
  - Test varying sources of First Guess/Background constraints (developing updated background mean/covariance based on observed TCs
  - Vary number of EOF basis functions for T and WV profiles:
- Testing ongoing with multiple TC cases
  - Dorian (2019), Florence (2018), Joaquin (2016), Matthew (2015), Edouard (2014)
  - Planned collaboration with CIRA to determine impact on TC intensity estimates

## Rain rate and Temperature Anomaly Cross-sections

Florence: 2018-09-12

#### **NPP Des**







#### TC Florence and Dorian NPP and N20 Temperature Bias: MiRS-ECMWF (Rainy, dist <= 100km)



### **Development of a Neural Network for MiRS Radiometric Bias Correction**

- Why bias correction?
  - To remove potential systematic bias between the measurements and the CRTM simulations.
- Current bias correction (static):
  - statistics of bias over oceanic and clear scenes. The bias is channel and scan position dependent. Local variations are not accounted for.
- A machine learning-based bias correction: Neural Network (NN)
  - NN was trained to learn the bias structure. Inputs are:
    - > Brightness temperature (TB) of the measurements,
    - Satellite viewing angle
    - Iatitude,
    - other geophysical parameters like cloud liquid water (CLW), total precipitable water (TPW), Tskin, Psfc, emissivity, etc.







Neural Network Schematic





## NN TB\_bias applied in MiRS



Yan Zhou et al. "Development of a Machine Learning-Based Radiometric Bias Correction for NOAA's Microwave Integrated Retrieval System (MiRS)" in poster session

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### JPSS/MetOp-SG sensor intercomparison (ATMS vs. MWS): Channel Selection

- MWS has similar channel set to ATMS, but with different polarizations for some frequencies.
- Will require updating several MiRS ancillary files (e.g. emissivity background means/covariances, ice/snow emissivity catalogs).
- Additional channel (24) at 229 GHz should help with ice detection and removal of impact on WV channels 19-23.
- Will need updated CRTM coefficients for MWS. UPDATE: coefficients now generated.

Source: MWS L1B ATBD

	AMSU/MHS				ATMS			MWS			
Ch.		GHz	Pol.	Ch.	GHz	Pol		h.	GHz		Pol.
1		23.8	QV	1	23.8	QV	1		23.8		QH
2		31.4	QV	2	31.4	QV	2		31.4		QH
3		50.3	QV	3	50.3	QH	3		50.3		QH
				4	51.76						
4		52.8	QV	5	52.8	QH	4		52.8		QH
							5	;	$53.246 \pm 0.0$	08	QH
5	53.59	05±0.115	QH	6	53.596±0.1	15 QH	6	5	53.596±0.11	15	QH
					•			'	$53.948 \pm 0.0$	81	QH
6		54.4	QH	7	54.4	QH	8	;	54.4		QH
7	4	54.94	QV	8	54.94	QH	9	)	54.94		QH
8	5	55.50	QH	9	55.50	QH	1	0	55.50		QH
9	57.3	290344	QH	10	57.29034	4 QH	1	1	57.290344	ŀ	QH
10	57.290	344±0.217	QH	11	57.290344±0	.217 QH	12	2	57.290344±0	.21	QH
11	57.290344	±0.3222±0.048	QH	12	57.290344 ±0.322	22±0.048 QH	1.	3	57.290344	ŀ	QH
12	57.290344	0.3222±0.022	QH	13	57.290344±0.322	2±0.022 QH	14	4	57.290344±0	.32	QH
13	57.290344	0.3222±0.010	QH	14	57.290344±0.322	2±0.010 QH	1	5	57.290344±0	.32	QH
14	57.290344±	0.3222±0.0045	QH	15	57.290344±0.3222	2±0.0045 QH	1	6	57.290344±0	.32	QH
15	89.0 QV										
16	89.0		QV	16	88.2	QV	1	7	89.0		QV
17	157.0		QV	17	165.5	QH	1	8	164-167		QH
18	183.311±1.0		QH	22	183.311±1	.0 QH	23	3	183.311±1.0		QV
				21	$183.31 \pm 1$	.8 QH	2	2	183.311±1.	8	QV
19	183.	311±3.0	QH	20	183.311±3	.0 QH	2	1	183.311±3.	0	QV
				19	183.311±4	.5 QH	20	0	183.311±4.	5	QV
20	191.31		QV	18	183.311±7	.0 QH	1	9	183.311±7.	0	QV
							24	4	229		QV
M	Iatched	Pol. is Different		Unique Passband Pol. is Dif		Pol. is Differen	it and U	nique	Passband	New	channel

### JPSS/MetOp-SG sensor intercomparison (ATMS vs. MWS): Scan Geometry



20.0 km x 20.0 km C

Channels 3 to 16 34.2 km x 66.9 km



17.0 km x 17.0 km Channels 17 to 24 29.0 km x 56.7 km

- Both MWS and ATMS are cross-track scanning
- Scan positions: MWS=95, ATMS=96
- Scan angle range: MWS = ±49.31, ATMS = ±52.73
- MWS channels 1 and 2 FOVs are significantly smaller than ATMS; this may preclude the need for resampling/footprint matching

Channels (MWS)	Nadir re (ki	solution m)	Edge of scan resolution (km)				
	MWS	ATMS	MWS	ATMS			
1-2	39.6 x 39.6	74.8 x 74.8	67.6 x132.6	141.8 x 323.1			
3-16	20.0 x 20.0	31.6 x 31.6	34.2 x 66.9	60.0 x 136.7			
17-24	17.0 x 17.0	15.8 x 15.8	29.0 x 56.7	30.0 x 68.4			

Source: MWS L1B ATBD

# Path Forward

#### Improvements implemented

- Snowfall rate integration for N20
- Hydrometeors (CLW over land for light rain detection)
- Snow cover/amount (vegetation correction)
- Updated sea ice climatology mask (finer temporal/spatial resolution) → Great Lakes ice detection
- Planned further improvements
  - Air mass-dependent bias corrections (machine learning)
  - Rainy condition sounding (update a priori constraints)
  - Experimenting with MiRS version adapted for tropical cyclones (MiRS-TC)
  - Hydrometeors: precharacterization of precip type, improvements to CRTM i.e. scattering, particle size/shape distribution in CRTM)
  - Applications/user feedback
- Planned/ongoing activities
  - N20/ATMS validation: Continued daily assessments (as posted to MiRS website) including comparisons with radiosondes, and other in situ data (e.g. for rain rate and land surface temperature)
  - Reprocessing of NPP and N20 mission data (Spring/Summer 2020)
  - Plan for JPSS-2 (2020-2021), and EPS-SG (2019-2022)

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# 3-D Visualization with NOAA-20/ATMS: Super Typhoon Yutu, 2018<u>-10-24</u>

