



NASA Sounder Science Team Meeting

MIIDAPS: A Consistent Variational Approach for the Inversion & Assimilation Preprocessing of Microwave and Infrared Satellite Data

Sid Ahmed Boukabara

NOAA NESDIS/STAR

Contributions from:

Eric Maddy, K. Garrett, E. Jones, N. Shahroudi and K. Ide





MIIDAPS

Multi-Instruments Inversion and Data Assimilation Pre-Processing System

Content

1

Introduction, Background, Heritage and Motivation

2

Overview of MIIDAPS Concept

3

Highlight of MIIDAPS Applications

- In Standalone retrieval Mode (for IR and MW)
- In Data Assimilation Pre-processing Mode

4

Summary & Conclusion



Background & Heritage (Science)

- Satellite data are sensitive to many parameters. Scientifically, it is preferable to invert all products simultaneously and use all measurements together.
- Increased information content at radiometric level before inversion, is superior to inversion separately. Advocate for MW+IR synergy. (2-channels > 1 channel)
- Basically no difference between IR and MW sensing phenomena besides different sensitivity, spectroscopy, NeDT, ...handled by variational approach.
- Radiative Transfer Modeling already handles uniformly MW and IR measurements simulations (including Jacobians): ex: CRTM
- Coupled data assimilation is a major focus: will lead to using DA analysis beyond NWP (to ocean, land, cryosphere, hydrometeors, etc). Same drivers should apply to Inversion Approaches (coupled inversion: sounding, surface, cryosphere, hydrometeors,...).
- MiRS algorithm, operational since 07 in NOAA, is proof of concept. Applied to microwave sensors (conical, cross-track, polar, low-orbit,..), generating sounding, hydrometeors, cryosphere, land and ocean products



Trends Foreseen (non-scientific)

- **Budget Pressure to continue, leading to innovations in cost effectiveness**
 - Need for Consolidating Algorithms into Enterprise Approaches
 - Leveraging Existing investments (i.e radiative Transfer, Inversion,...)
- **Multiplication of Satellites: International, Commercial, Small Satellites**
 - Need for Agile Approaches for handling satellite data
 - Need for consistent processing with single algorithms to avoid discrepancies
- **Increased spatial and temporal resolutions will lead to usage of data assimilation for situational awareness (nowcasting and short term forecasting)**
 - Relevance of separate inversions? (different orbits, formats, quality, etc)
 - Blurriness between retrieval and data assimilation/fusion for nowcasting applications
 - Advocate for making retrieval algors '**dual use**': for Inversion and DA pre-processing
 - Radiances vs retrieval Assimilation: Integration of algorithms technology as an alternative



MIIDAPS

Multi-Instruments Inversion and Data Assimilation Pre-Processing System

Content

1

Introduction, Background, Heritage and Motivation

2

Overview of MIIDAPS Concept

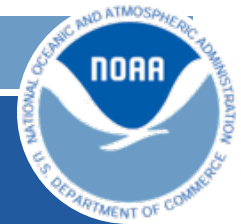
3

Highlight of MIIDAPS Applications

- In Standalone retrieval Mode (for IR and MW)
- In Data Assimilation Pre-processing Mode

4

Summary & Conclusion



MIIDAPS Concept

3 years ago (In 2014), a New generation System started was initiated (as a pilot project)

Goal was to build on MiRS heritage and develop a system with a dual use: Retrieval and Data Assimilation Pre-processing.

Goal is also to generalize it to all sensors including IR and Hyperspectral IR

Goal is to cross-fertilize DA and retrieval applications. Similar in concept to the CRTM (used for inversion, calibration, climate applications and at the same time to data assimilation)

Variational Mathematical Basis

- Cost Function to Minimize (similar between retrieval & assimilation):

$$J(X) = \left[\frac{1}{2} (X - X_0)^T \times B^{-1} \times (X - X_0) \right] + \left[\frac{1}{2} (Y^m - Y(X))^T \times K_n^{-1} \times (Y^m - Y(X)) \right]$$

Jacobians & Radiance Simulation from Forward Operator: CRTM

- To find the optimal solution, solve for: $\frac{\partial J(X)}{\partial X} = J'(X) = 0$

- Ass Same Methodology applied to all parameters:

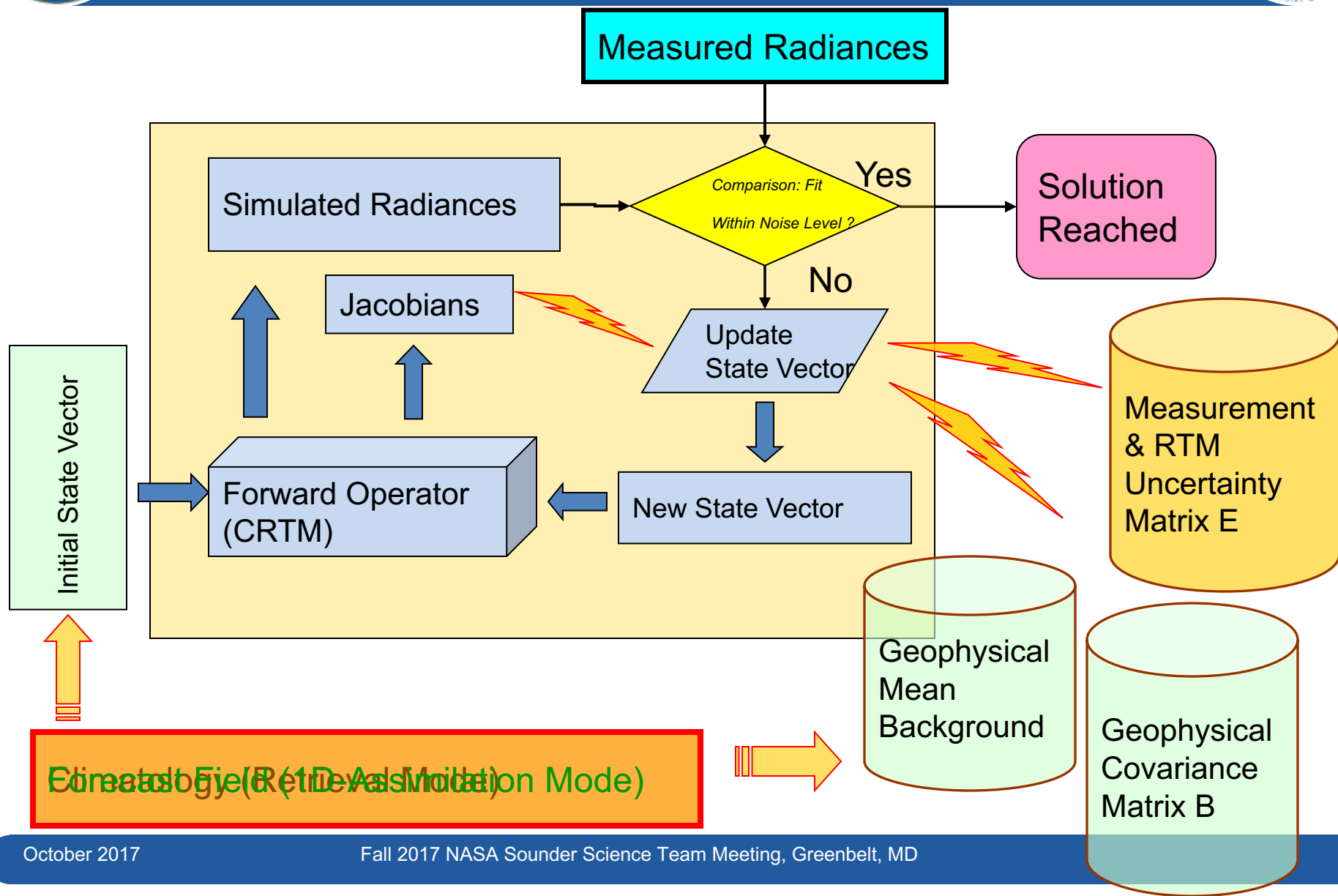
State Vector X contains:

- T(p), Q(p), T_{skin}
- Emissivity Vector
- Hydrometeors (frozen and liquid water)

- This leads to iterative solution.

$$\Delta X_{n+1} = \left\{ B K_n^T (K_n B K_n^T + E)^{-1} \right\} \left[(Y^m - Y(X_n)) + K_n \Delta X_n \right]$$

MIIDAPS Variational Retrieval/Assimilation

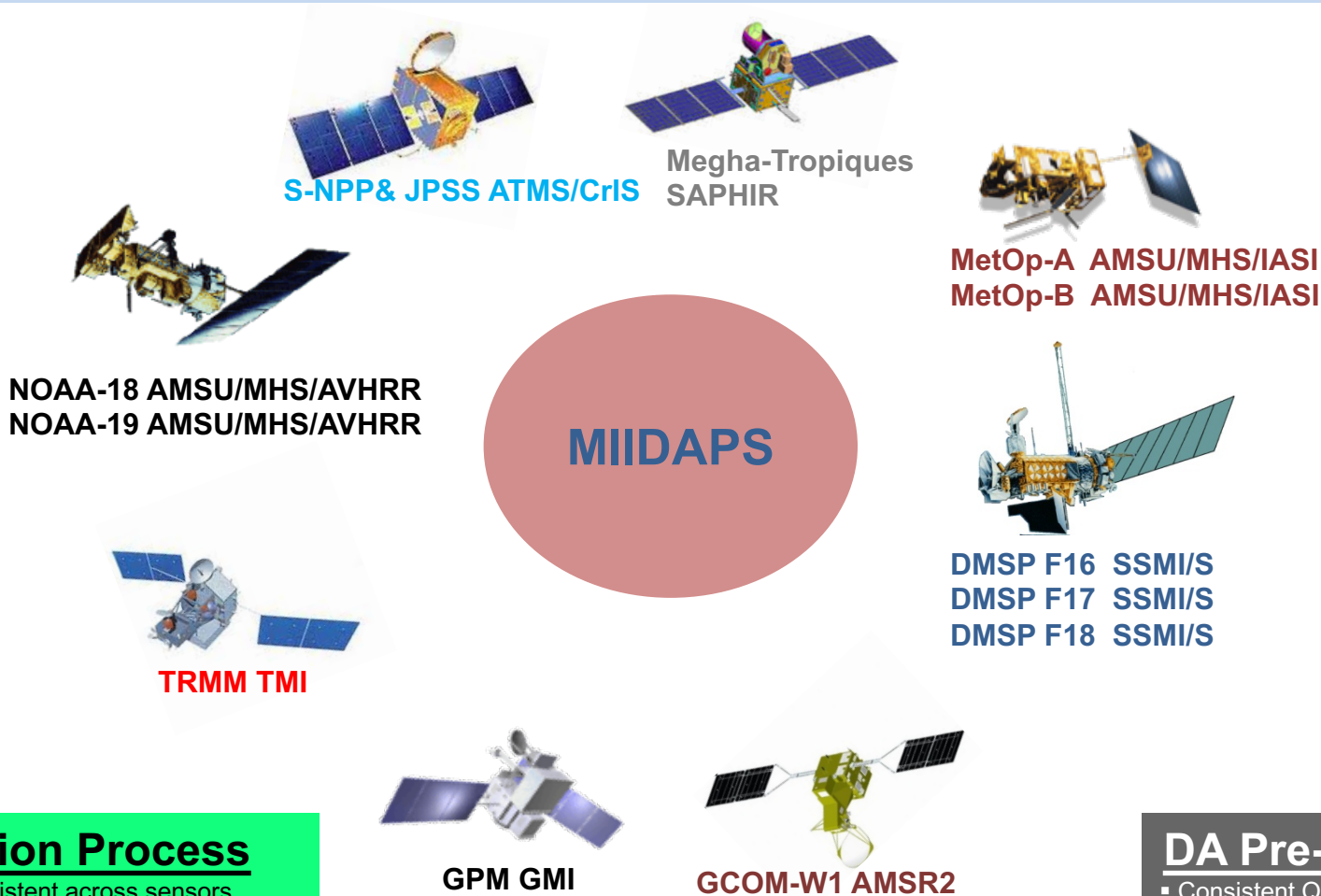




MIIDAPS Algorithm Concept

Multi-Instrument Inversion and Data Assimilation Preprocessing System

Motivation: Universal retrieval and Data Assimilation preprocessor for all satellite observations



Inversion Process

- Inversion consistent across sensors
- All parameters included in state vector
- CRTM for forward/Jacobian operators
- Valid over all surfaces/all-sky conditions
- Use forecast, fast regression or climatology as first guess/background

****MIIDAPS also applicable to GOES-16 Sounder, Meteosat SEVIRI, AHI, ABI, MODIS, AIRS, etc**

DA Pre-Processing

- Consistent Quality Control (rr,ice,...)
- Consistent pre-processing for non-analyzed parameters
- Corrects displacements (fronts,...)
- Modular design, scalable
- Use of MPI for HPC



MIIDAPS

Multi-Instruments Inversion and Data Assimilation Pre-Processing System

Content

1

Introduction, Background, Heritage and Motivation

2

Overview of MIIDAPS Concept

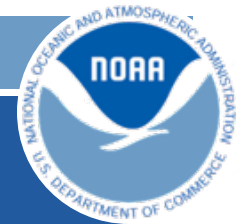
3

Highlight of MIIDAPS Applications

- In Standalone retrieval Mode (for IR and MW)
- In Data Assimilation Pre-processing Mode

4

Summary & Conclusion



MIIDAPS :

AS STANDALONE INVERSION SYSTEM (FOR IR and MW SENSORS)

- Applied to MW and IR sensors
- Coupled Inversion: Sounding, Hydrometers, and Surface Parameters
- Quality of Products inverted depends on Information Content



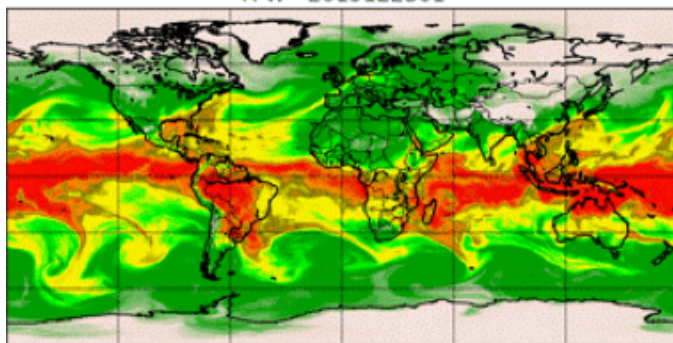
MIIDAPS Capabilities (Sensors & Geophysical)

MIIDAPS Satellite Products											
	T(p)	Q(p)	SST/LST	TPW	Cloud & Ice Amt	Cld Type/Top	Precip	Sfc Emiss	SIC/SWE	Trace Gas	QC (ChiSq)
NOAA-18 AMSU/MHS											
NOAA-18 AVHRR											
NOAA-19 AMSU/MHS											
NOAA-19 AVHRR											
Metop-A AMSU/MHS											
Metop-A IASI											
Metop-B AMSU/MHS											
Metop-B IASI											
SNPP ATMS											
SNPP CrIS											
DMSP SSMI/S											
Aqua AMSU											
Aqua AIRS											
Megha-T SAPHIR											
GPM GMI											
GCOM-W1 AMSR2											
GOES-15 Sndr/Imgr											
Meteosat SEVIRI											
Himawari-8 AHI											
GOES-16 ABI											

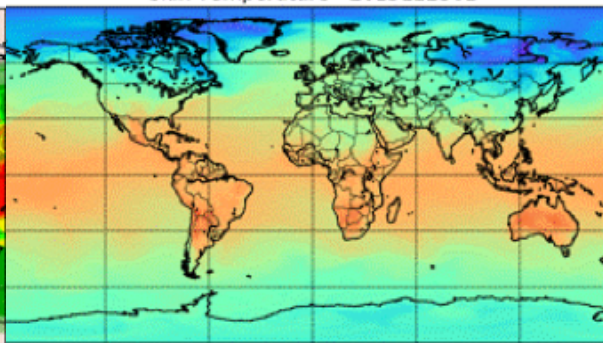
Illustration of the MIIDAPS Outputs

(geophysical parameters)

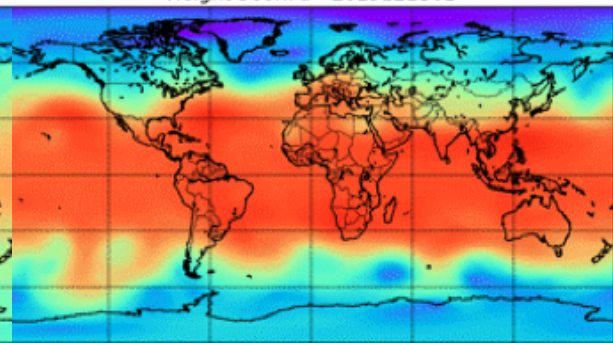
TPW - 2015122301



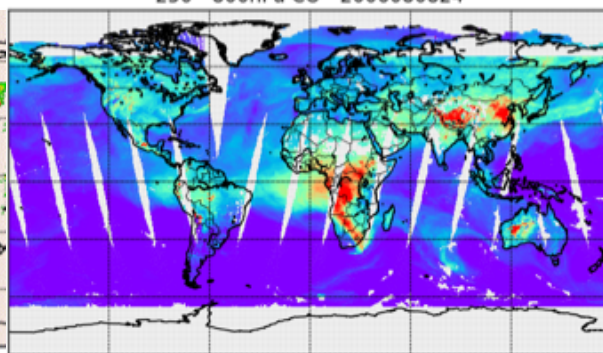
Skin Temperature - 2015122301



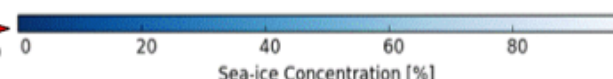
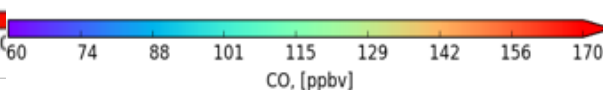
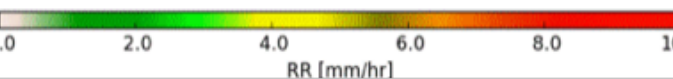
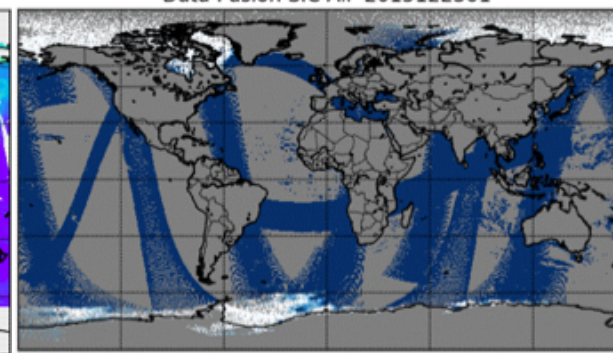
Height 500hPa - 2015122301



250 - 800hPa CO - 2006080824



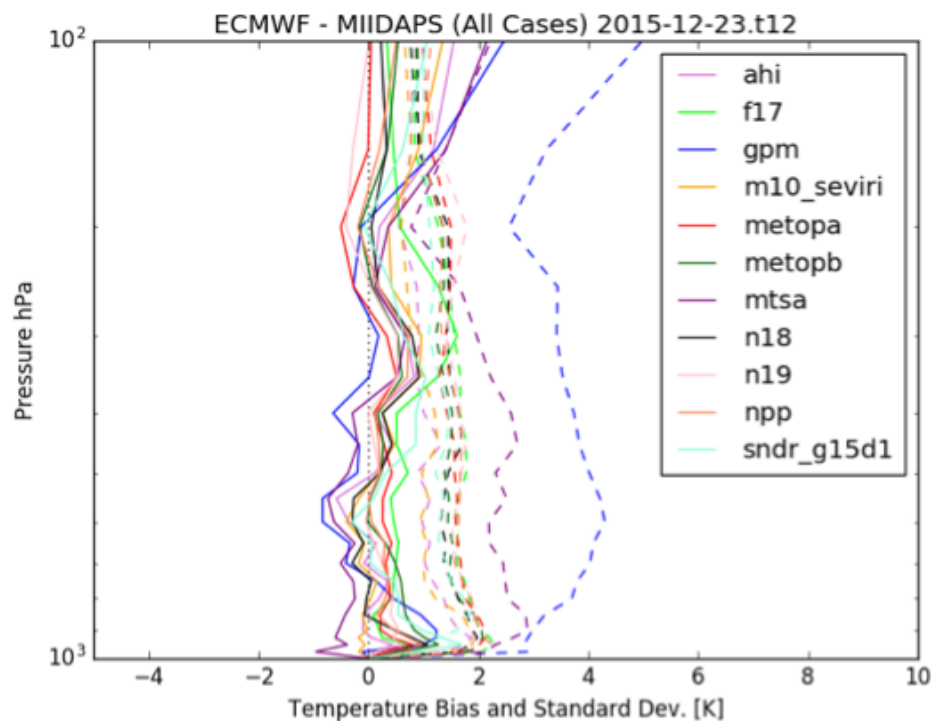
Data Fusion SIC All 2015122301



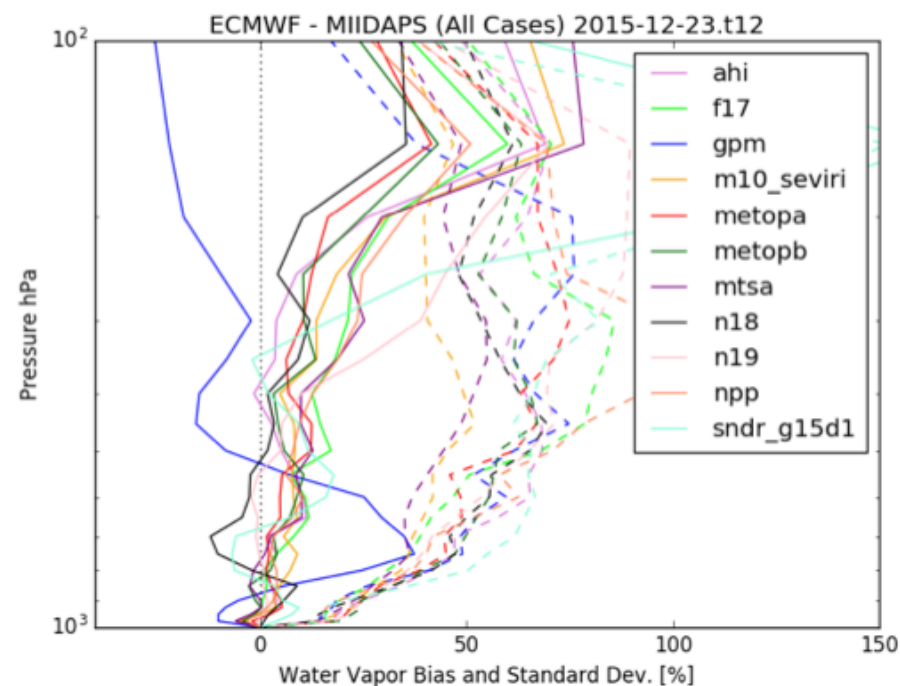
MIIDAPS Processes Microwave and Infrared Satellite Observations to Invert Atmospheric, Surface, Hydrometeors and Cryospheric products. It has Dual Use: It can be used as standalone for inversion applications and as a pre-processing system for data assimilation applications.



MIIDAPS Sounding Performances Assessment on 11 Sensors (Use ECMWF as a reference)



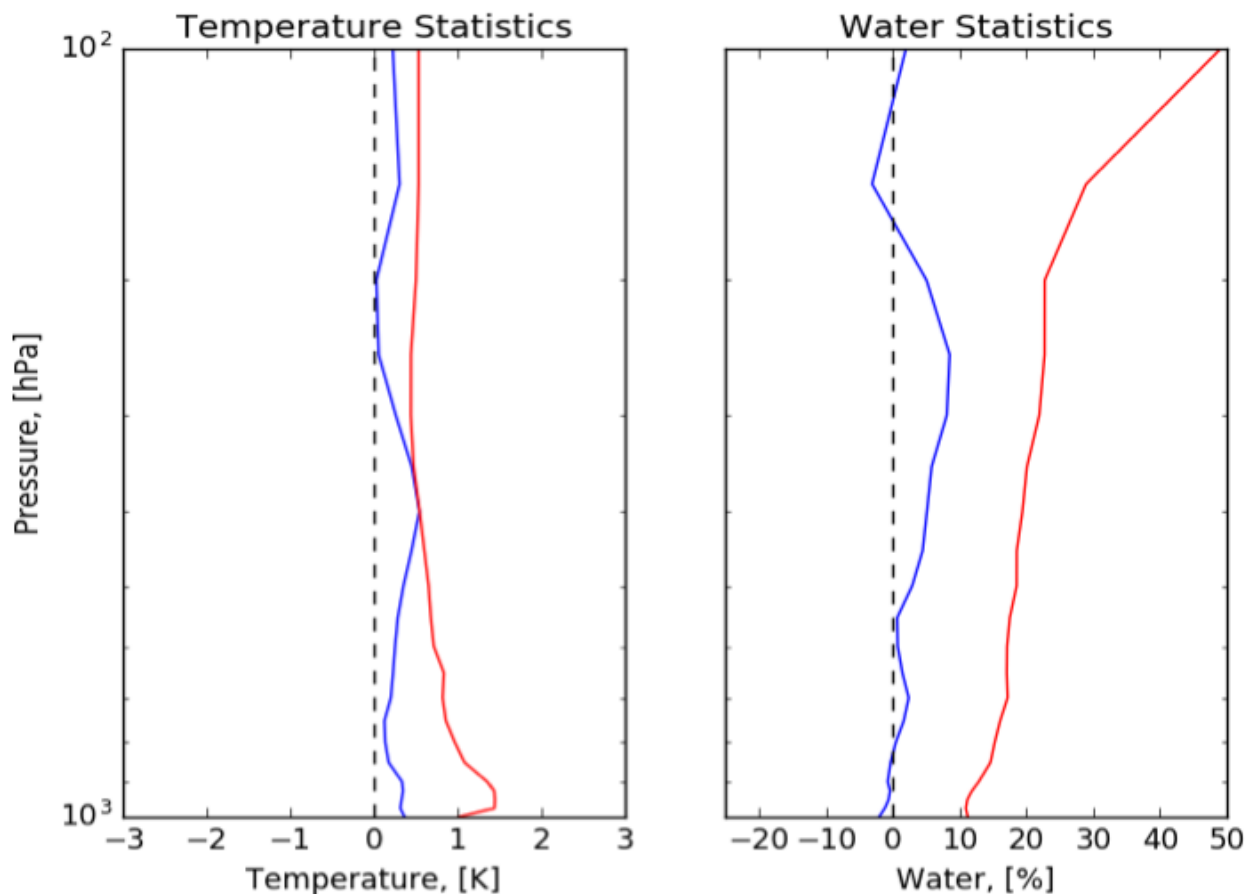
This is a rough assessment of the performances. This needs closer attention (margin for improvements). Examples: Assumed Obs Errors not tuned, all cases included (ocean, land, rainy, cloudy, clear-sky, etc). IR retrievals have clouds filtered out.



MIIDAPS Performances

Temperature and Water Vapor Profiles

Global Bias and Standard Deviation



- Clear Sky Data
- Land and Ocean
- No bias correction applied
- CrIS configuration
- Out-of-the-box testing (no optimization)
- Full spectrum used
- ECMWF Used as reference



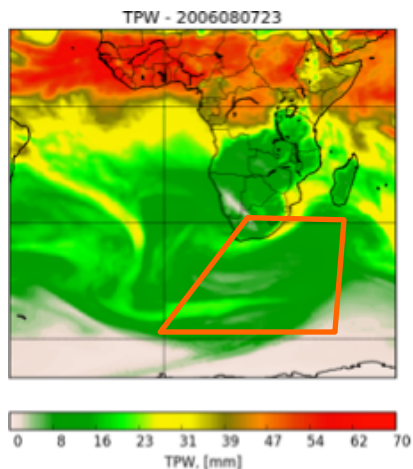
MIIDAPS :

AS PRE-PROCESSING FOR SATELLITE DATA ASSIMILATION (NWP)

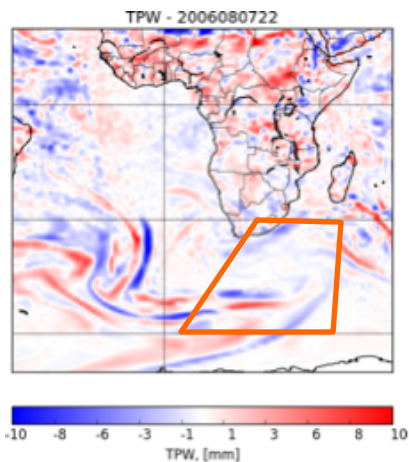
- Universal Quality Control Tool (Non-Convergence, Rain, Ice detection, etc)
- Pre-Processing system to provided non-analyzed data (add to the NWP analyses)
- Corrects forecast background displacement(s): cold/warm fronts, Cloud/storm position, etc

Assessment of Displacement Correction Using OSSE Testbed & Associated Nature Run (NR)

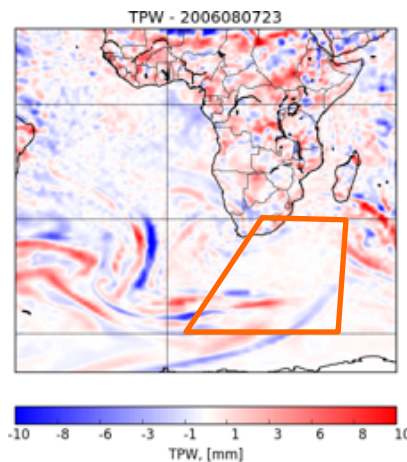
G5NR TPW



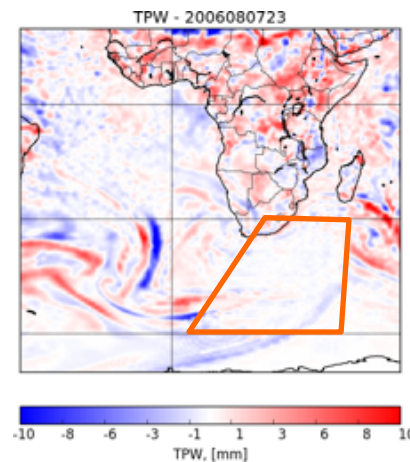
a) Background – G5NR



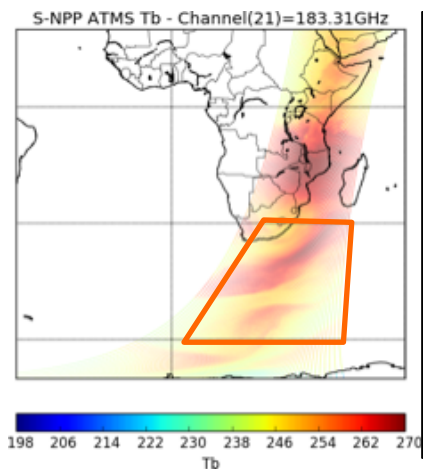
b) GSI analysis – G5NR



c) w MIIDAPS pre-processing



Location of ATMS obs.



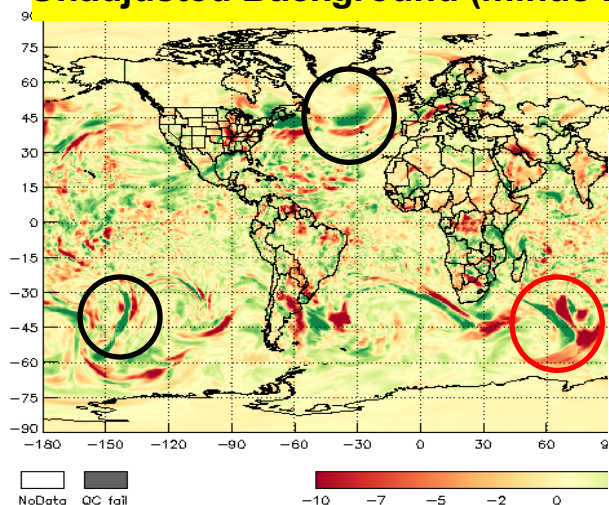
Example Analysis Cycle at 2006-08-07 23Z

- a) Background-G5NR shows large displacements (dipoles) in TPW field
- b) GSI analysis-G5NR reduces magnitude of dipoles slightly where SNPP ATMS data exists (red trapezoid)
- c) Data Fusion analysis through MIIDAPS-based background adjustment removes most of dipole feature and reduces TPW differences where SNPP ATMS data exists

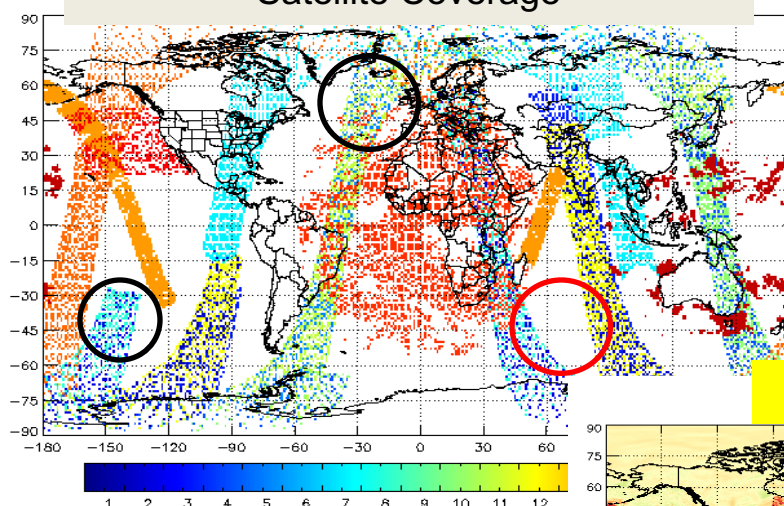
Illustrative Example with real Data: (of Displacement Correction)

Background Adjustment Example, All Satellite Data: December 23, 2015 00Z

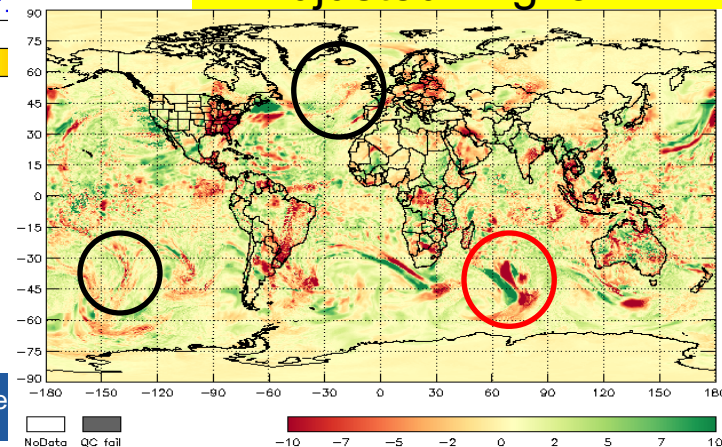
Unadjusted Background (minus ECMWF)



Satellite Coverage



Adjusted Bkg for DA



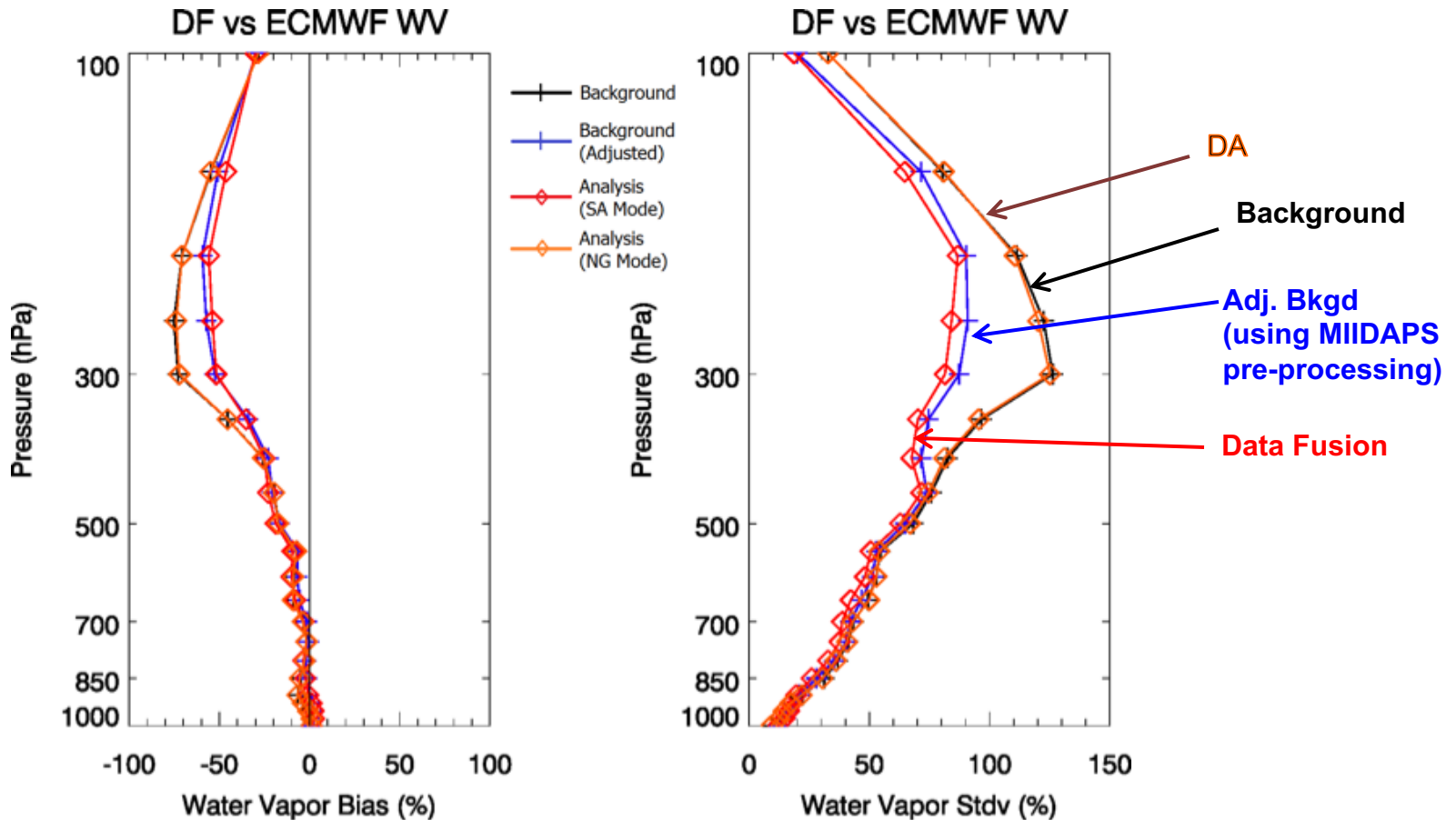
Main advantages of Using MIIDAPS in NWP:

- Complete analyses with more parameters including cryosphere, hydrometeors, land, ocean, trace gases, etc)
- Fit to observations (displacement correction)
- Flexible to extend to other sources of Obs.
- Provide QC and ice/rain detection to users to assimilation step

Performance Assessment

Example of Data Fusion Analyses for 2015-12-23 12Z Cycle

Performance at observation locations



- Background Adjustment provides displacement correction in analysis.



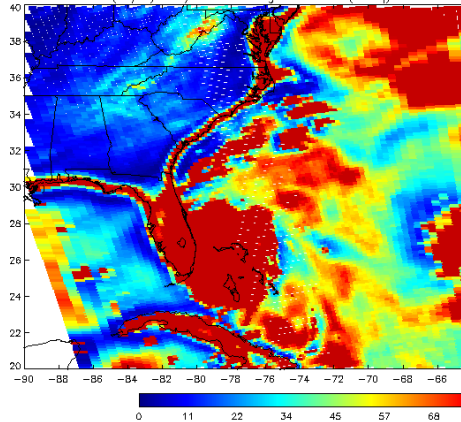
Generating & Displacement Correction of Hydrometeors in Analysis (Case Hurricane Matthew, Oct 2016, SNPP/ATMS)

0:Background:Fcst**#1**

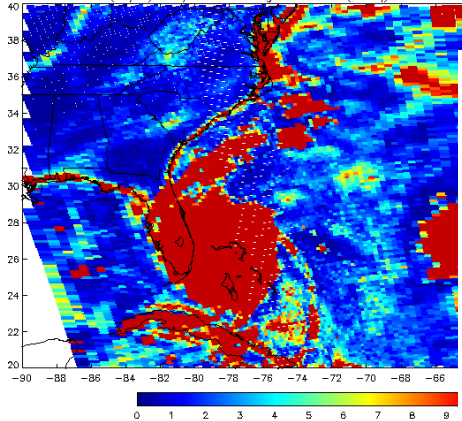
Iterations

#8**#14**

MIIDAPS-ECMWF (a1/10) NPP/ATMS Convergence Metric (ChiSq) 2016-10-06 Asc (r9999)

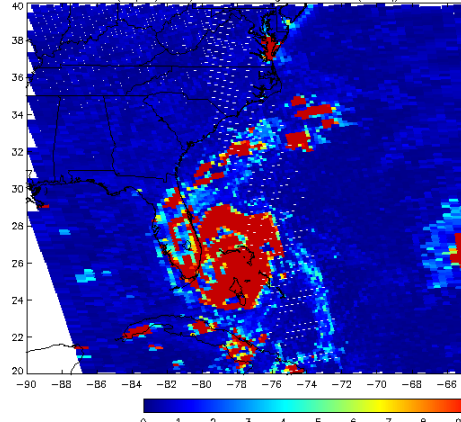


MIIDAPS-ECMWF (a1/11) NPP/ATMS Convergence Metric (ChiSq) 2016-10-06 Asc (r9999)

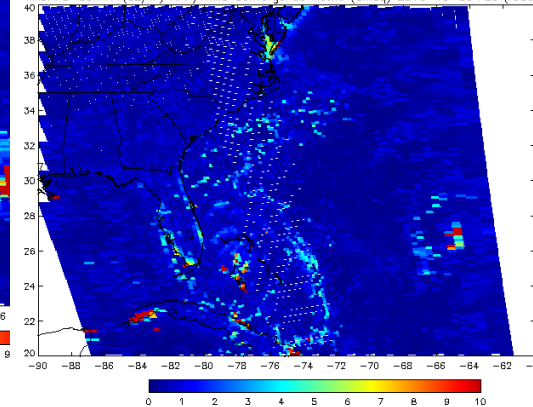


Convergence: Fitting Observations when using MIIDAPS

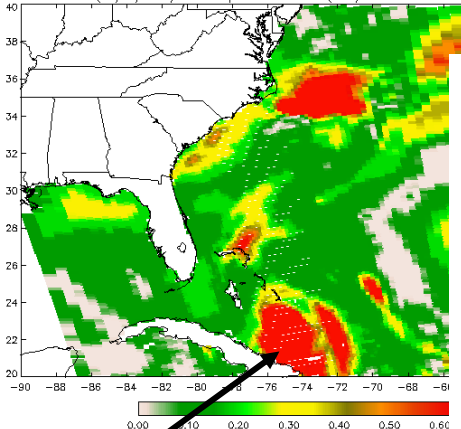
MIIDAPS-ECMWF (a2/11) NPP/ATMS Convergence Metric (ChiSq) 2016-10-06 Asc (r9999)



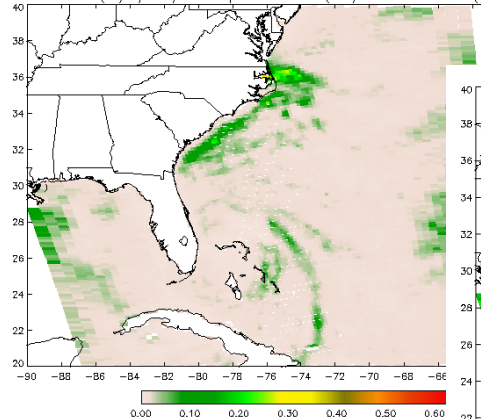
MIIDAPS-ECMWF (a2/17) NPP/ATMS Convergence Metric (ChiSq) 2016-10-06 Asc (r9999)



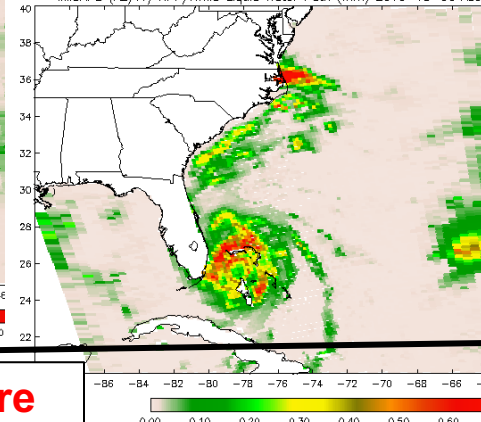
MIIDAPS (A1/10) NPP/ATMS Liquid Water Path (mm) 2016-10-06 Asc (r9999)



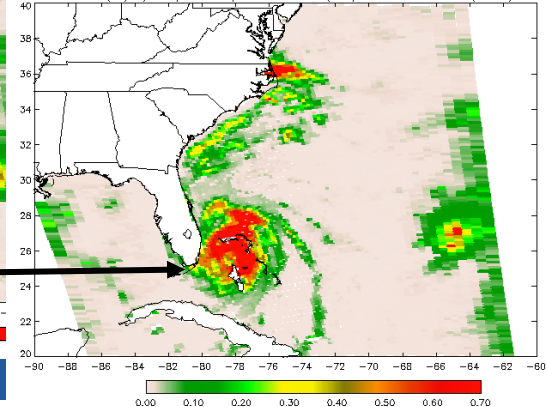
MIIDAPS (A1/11) NPP/ATMS Liquid Water Path (mm) 2016-10-06 Asc (r9999)



MIIDAPS (A2/11) NPP/ATMS Liquid Water Path (mm) 2016-10-06 Asc (r9999)



MIIDAPS (A2/17) NPP/ATMS Liquid Water Path (mm) 2016-10-06 Asc (r9999)



Cloud field: Displacement Correction & More Accurate QC Tool for DA when using MIIDAPS

Temperature Vertical and Horizontal Displacements Correction

0:Background:Fcst

#1

Iterations

#8

#14

54 GHz Residual Convergence: Fitting Observations

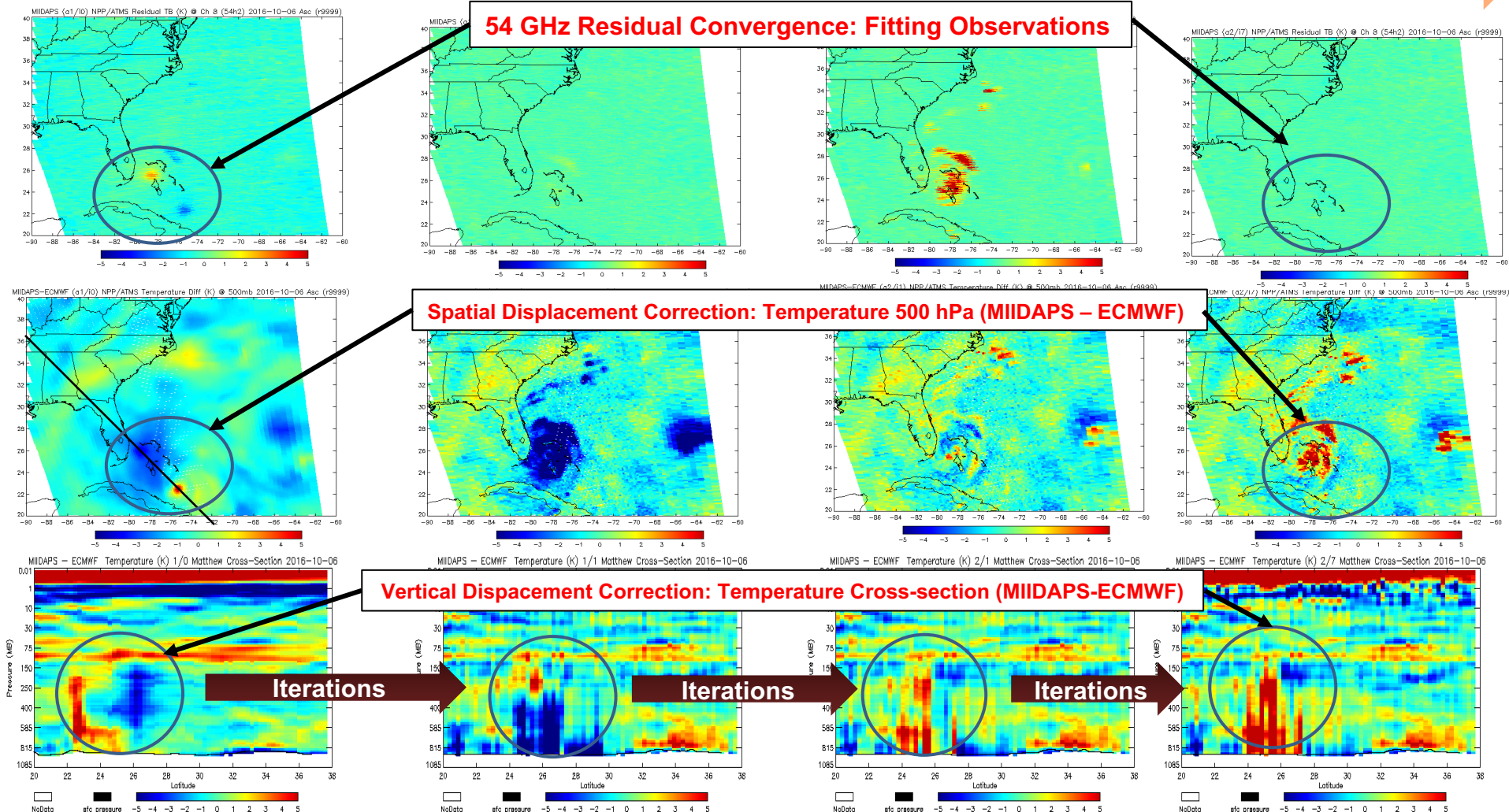
Spatial Displacement Correction: Temperature 500 hPa (MIIDAPS - ECMWF)

Vertical Displacement Correction: Temperature Cross-section (MIIDAPS-ECMWF)

Iterations

Iterations

Iterations





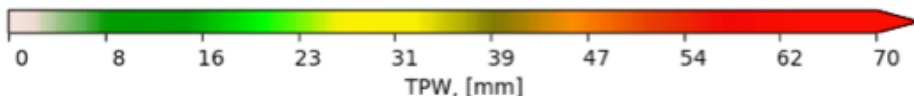
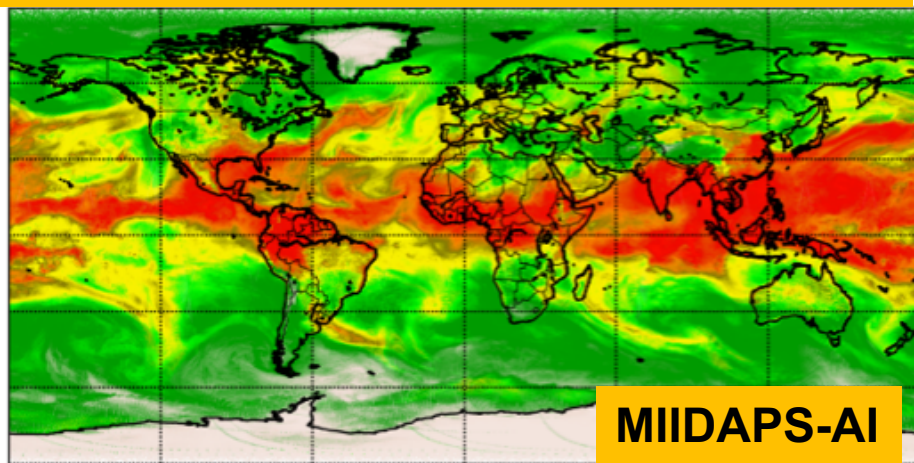
Findings

- MIIDAPS, when applied as standalone retrieval algorithm for both IR and MW sensors, retrieves sounding along with surface and hydrometeors data
- MIIDAPS has been applied (but not optimally tuned) to a large number of sensors.
- Quality of these retrievals depends on the sensors and the information content
- MIIDAPS (like MiRS) adopts a coupled inversion system, where measurements are inverted simultaneously, to provide geophysical data
- MIIDAPS, when applied as a Data Assimilation Preprocessing system, offers a way to correct spatial and vertical displacements in NWP analyses, for (1) cloud, (2) moisture, and (3) temperature, making them useful for Nowcasting
- Convergence Metric offers a powerful QC tool: In DA preprocessing mode, MIIDAPS also offers a universal QC tool to detect rain, ice, cloud, etc
- MIIDAPS major challenge: time cost. CRTM is significantly slow, especially in cloudy/rainy conditions (effort is on going to speed it up)

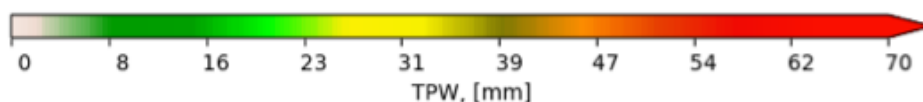
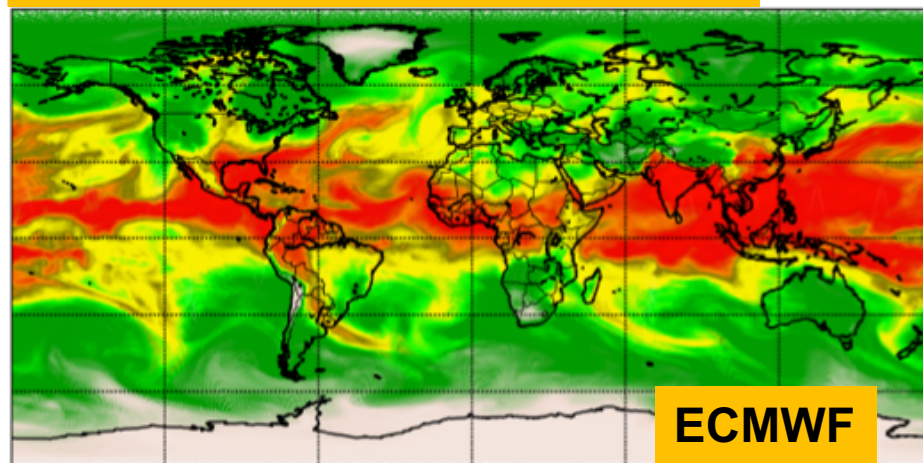
MIIDAPS-AI:

Exploring Artificial Intelligence for Remote Sensing/Data Assimilation/Fusion Applications

MIIDAPS-AI outputs (TPW) Using SNPP/ATMS Real Data



Reference source of TPW: ECMWF Analysis



- Comparison of AI-Based (left) and Reference field (right)
- Uses Google tensorflow tool.
- AI algorithms uses real SNPP/ATMS data as inputs
- 40,000 points used train model using 2 epochs, each with 6000 iterations.
- Using the NR dataset (indep. from ECMWF) for training

	AI-Based MIIDAPS	MIIDAPS (similar to oper. System)
Processing Time (Approx) for a full day data. Excluding I/O	~5 seconds	~ 2 hours

Several approaches being/will be tested:

- (1) Direct Inversion of data Using an Deep-Learning AI algorithm
- (2) Retain Variational Inversion but Use an AI-based RT/Jacobian Forward Operator



MIIDAPS

Multi-Instruments Inversion and Data Assimilation Pre-Processing System

Content

1

Introduction, Background, Heritage and Motivation

2

Overview of MIIDAPS Concept

3

Highlight of MIIDAPS Applications

- In Standalone retrieval Mode (for IR and MW)
- In Data Assimilation Pre-processing Mode

4

Summary & Conclusion



Summary & Conclusion

- MIIDAPS: community software (freely available) with dual use: (1) standalone inversion and/or (2) DA pre-processing system
- It leverages CRTM for radiance and Jacobians computations
- Applicability to MW and IR sensors as long as they are handled by CRTM: sounders/imagers, x-track/conical, Polar/Geo
- Main advantage: Consistency+cost effectiveness+agility, Drawback: Time cost
- Handling cloud/rainy data in MIIDAPS by varying hydromet. in state vector.
- Handling surface-sensitive channels by varying emissivity
- Products Assessment suggests approach provides very reasonable results.
Massive amount of validation is needed to finalize this (tens of sensors, tens of products for each)
- Displacement corrections & Universal QC major benefit of MIIDAPS in NWP
- MIIDAPS as entry point to leveraging Remote Sensing in NWP applications.
- Current Use of MIIDAPS in Data Fusion System for Nowcasting applications