



Benefits of Using a Variational Preprocessing Approach for the Assimilation of Satellite Radiances: An Application to Data Assimilation in Environmental Data Fusion

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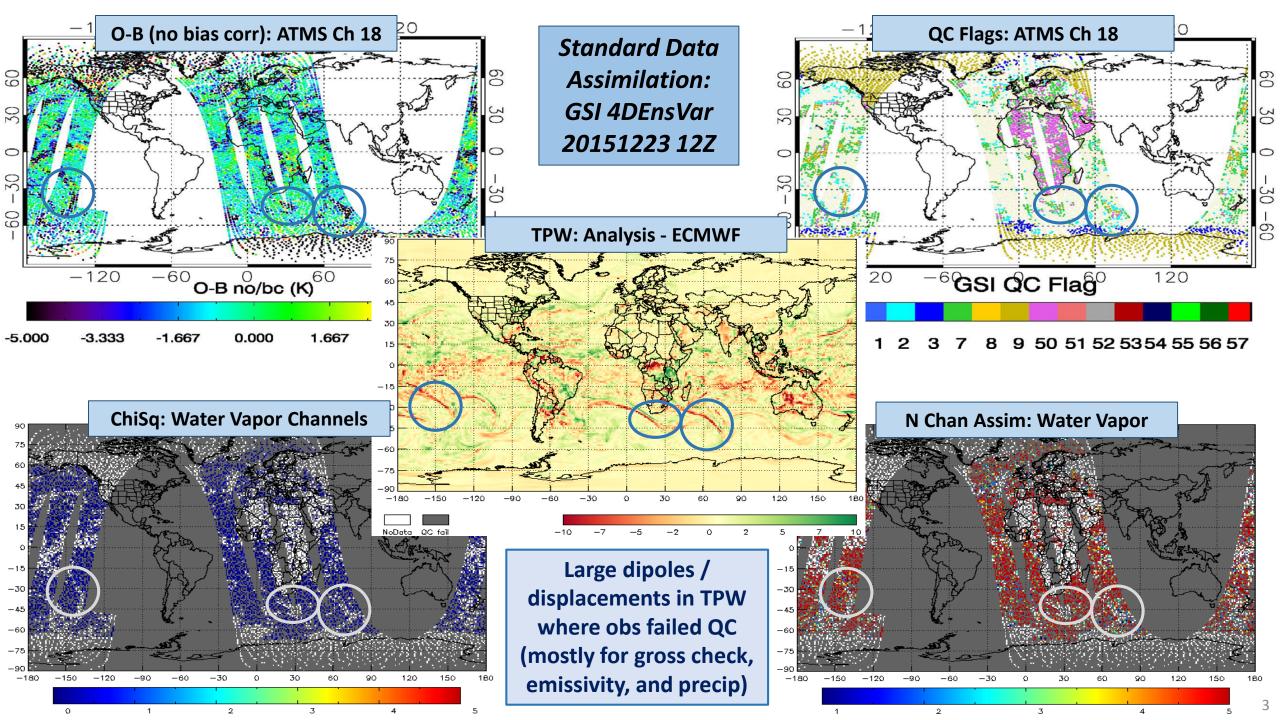
² NOAA NESDIS/STAR

Objective and Motivation

- The problem:
 - Satellite observations, especially those sensitive to moisture variables, tend not to be assimilated very well in data assimilation systems
 - Observations are often QC-ed out where the background differs too much from the observation, and/or in regions where observations tend not to be modeled well (e.g. cloudy or precipitating areas)
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- The goal:
 - Lessen dependence on the background in the analysis:
 - Create an analysis that is more informed by satellite observations, especially for moisture variables
 - Address the problem of displacements in the analysis fields
 - This can be achieved by putting satellite data through a preprocessor prior to assimilation: MIIDAPS (The Multi-Instrument Inversion and Data Assimilation Preprocessing System)





MIIDAPS Background

• MIIDAPS: The Multi Instrument Inversion and Data Assimilation Pre-Preprocessing System

 $y(x) = y(x_0) + K | x - x_0 |$

• 1DVar preprocessor based on the MiRS (Microwave Integrated Retrieval System) algorithm, which has been operational at NOAA since 2007:

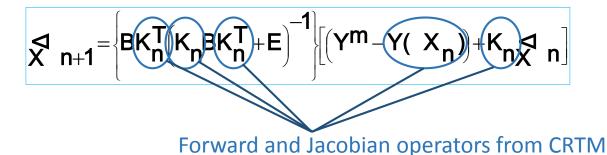
 $Minimize \ the \ cost \ function: \qquad J(X) = \left[\frac{1}{2}(X - X_0)^T \times B^{-1} \times (X - X_0)\right] + \left[\frac{1}{2}(Y^m - Y(X))^T \times E^{-1} \times (Y^m - Y(X))\right]$

Solve for:
$$\frac{\partial J(X)}{\partial X} = J(X) = 0$$

This methodology can be applied to all parameters, including hydrometeors

Iterative solution:

Assume linearity:

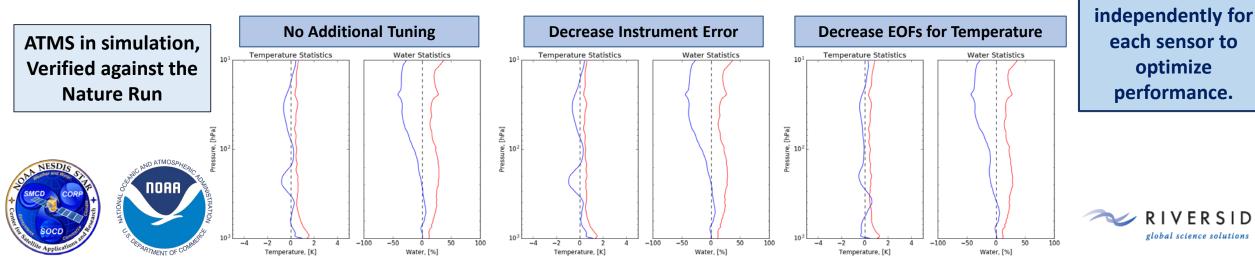






MIIDAPS Background

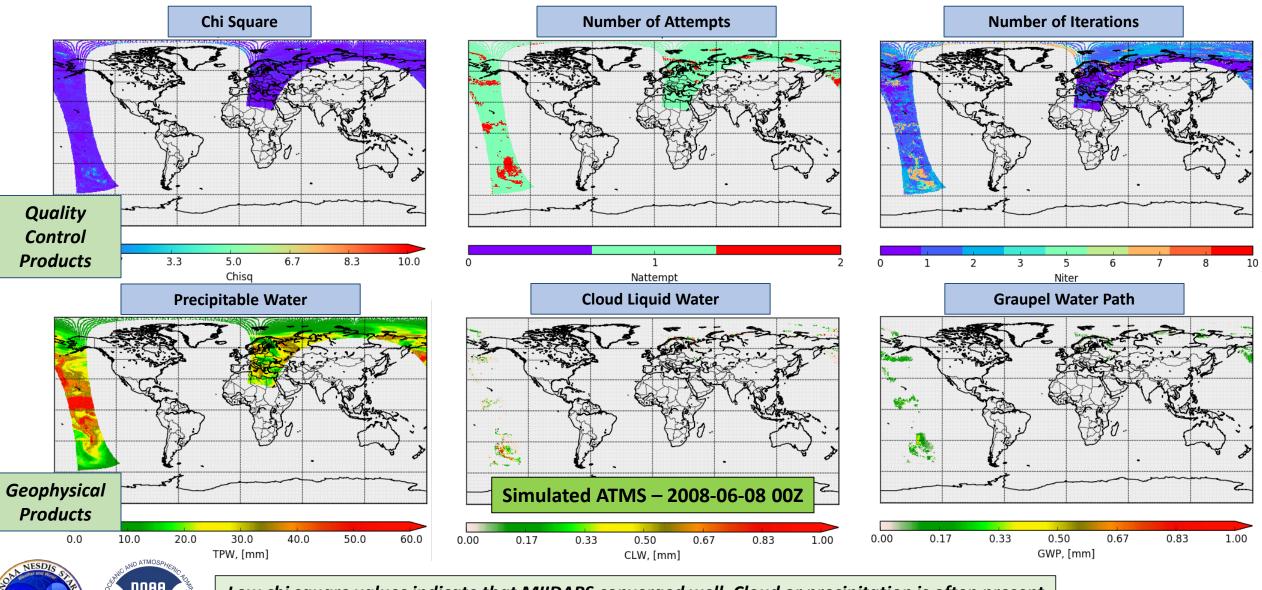
- MIIDAPS: The Multi Instrument Inversion and Data Assimilation Pre-Preprocessing System
 - Methodology doesn't vary based on platform; approach could be extended to any sensor (microwave or infrared)
 - Approach is valid over all surfaces and in all-sky conditions
 - Can use climatology or an NWP field as a first guess/background to the 1DVar
 - Has several benefits, including:
 - Highly tunable retrievals
 - The potential to provide consistent quality control for a DA system
 - The ability to be run in a parallel in an HPC environment



MIIDAPS retrievals

can be tuned

MIIDAPS Products and Performance



Low chi square values indicate that MIIDAPS converged well. Cloud or precipitation is often present where chi square values are high; information that can be useful in informing QC.

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MIIDAPS Products and Performance

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											

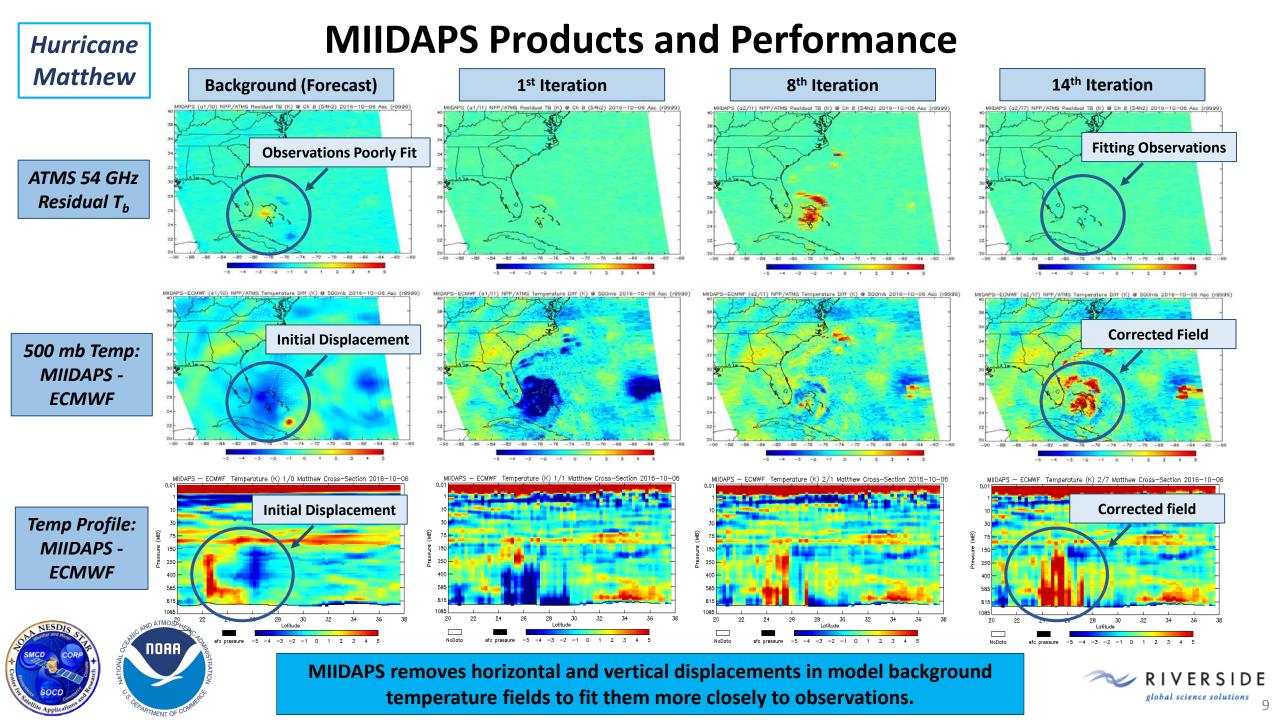


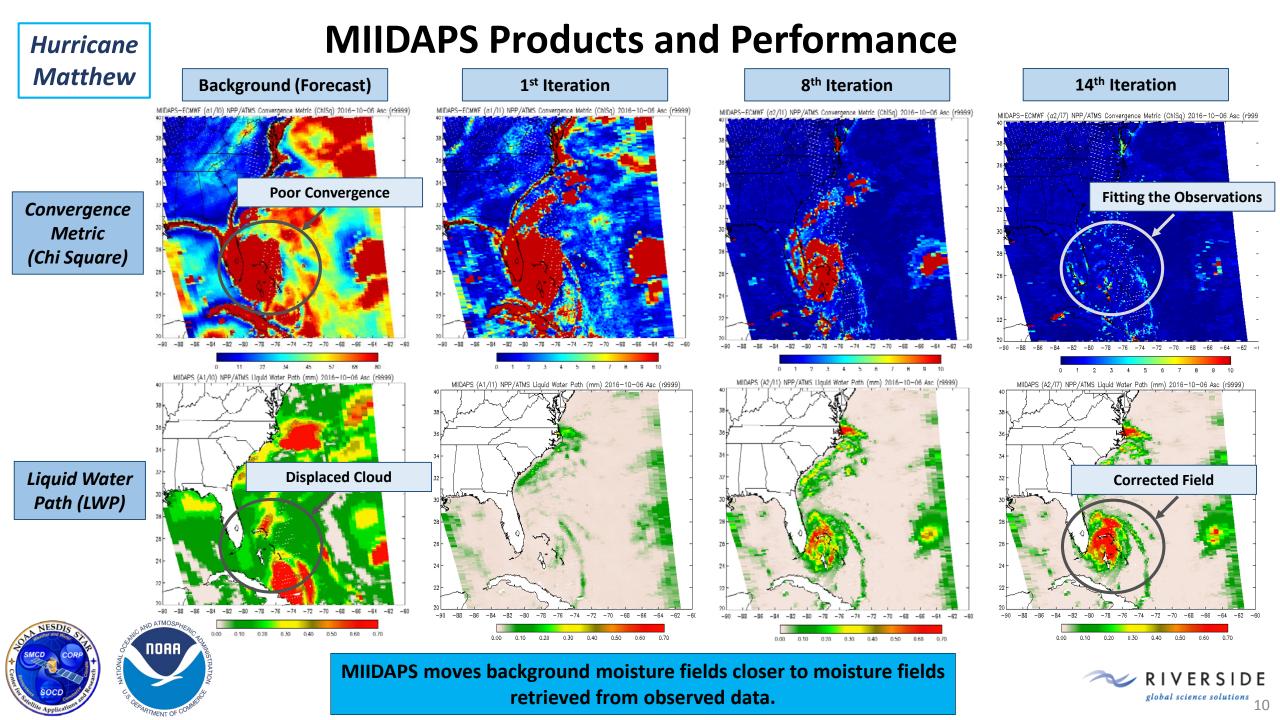
Moderate Information Content / Confidence

Good Information Content / Confidence

Not Enough Information Content for Retrieval

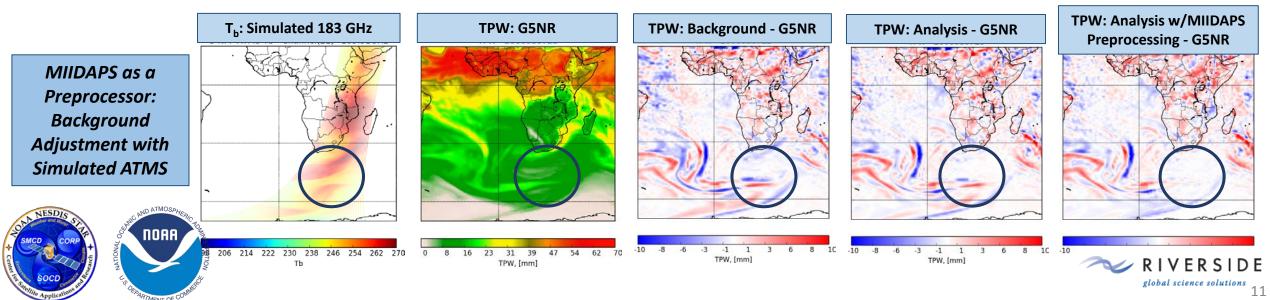




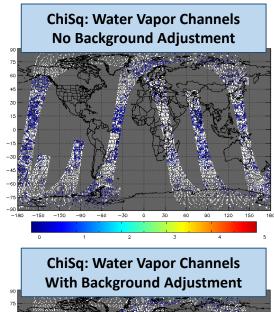


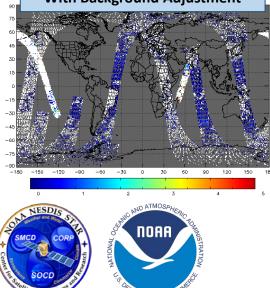
MIIDAPS Applications

- MIIDAPS as a preprocessor for satellite data assimilation:
 - Retrieved products (e.g. temperature, moisture profiles) can be used to adjust the model background prior to assimilation
 - Tunable, depending on the sensor
 - Retrieved products can be used to inform/constrain unanalyzed variables (not part of the state vector, but impact radiance simulation) in the assimilation system
 - Convergence metrics and other products can be used for universal quality control in the assimilation system
 - More accurate quality control and more observations assimilated



MIIDAPS as a Preprocessor in Data Fusion

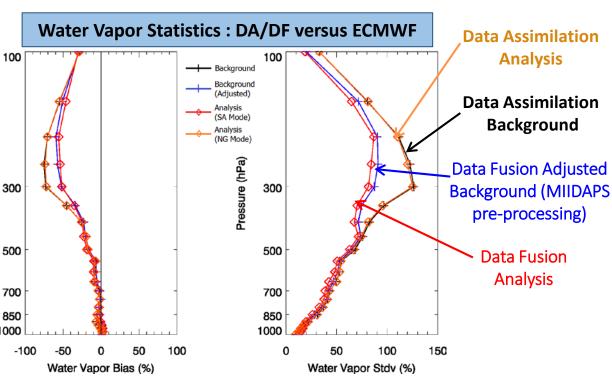


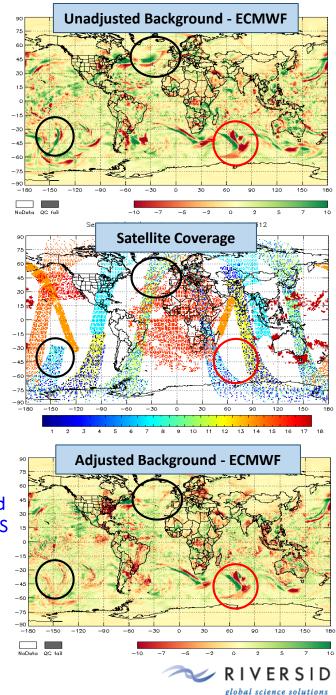


MIIDAPS Applications

• Data Fusion:

- Can use MIIDAPS as a preprocessor for data assimilation, with options to use background adjustment, MIIDAPS QC, and variable constraints
- High resolution (25km) global hourly analysis informed by satellite data over the atmospheric column; useful for situational awareness





Conclusions and Future Work

- MIIDAPS is a sensor-agnostic tool:
 - Capable of retrieving geophysical parameters from satellite data; retrievals are highly tunable
 - Can serve as a preprocessor in data assimilation
 - Valid for use with both microwave and IR sensors
- MIIDAPS results are valid over all surface types, and in all-sky conditions
- MIIDAPS has been shown to correct displacements in modeled fields
 - Able to adjust a data assimilation background to more closely fit the observations
- Work is ongoing to validate MIIDAPS as a preprocessor in the Data Fusion system
- Work is planned to test analyses produced using MIIDAPS as initial background for NWP



