## Data Assimilation Efforts MW Radiances Over Land

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## Motivation

#### Problem:

- Accurate forward modeling of satellite radiances is a key component for successful assimilation of satellite data into Numerical Weather Prediction models.
- To achieve this goal, radiative transfer models require a solid estimate of surface conditions, surface emissivity in particular.
- While this requirement is relatively easy to meet over the ocean, land surfaces have been a challenge.
- Both radiative transfer and the direct estimate of surface emissivity struggle to reach the desired accuracy.

#### Steps taken in approach to improve MW radiance assimilation:

- 1. Provide a better first guess of emissivity to the forward model.
- 2. Allow the emissivity to vary through the assimilation process (i.e. to be a control variable).
- 3. Improve QC for MW over land (ATMS sensor presented here).



#### Emissivity as a control variable in GSI

Goal is to prepare the system for use of any emissivity given as an input, have knowledge and capability to maximize the extraction of available information. Increase the number of assimilated observations over land without degrading the forecast.

## 1. Providing a better first guess of emissivity to the forward model

- GSI currently relies on CRTM's emissivity for MW and IR frequencies.
- We explored alternative sources for information on emissivity over land for MW until CSEM is finalized
- Freely available and widely used Tool to Estimate Land Surface Emissivities at Microwave frequencies (TELSEM2)
- Within GSI crtm\_interface module TELSEM2 atlas is read to provide the "user emissivity" to the CRTM via its optional flag feature.

	CRTM	TELSEM 2
Surface type	All	Land & sea-ice only
Frequency	3 – 190 GHz	10 – 700 GHz
Polarization	H + V	H + V
Spatial Resolution	0.25°	0.25°
Temporal Resolution	Instantaneous	Monthly
Base	"Physical"	Empirical

Note: TELSEM 1 differs from TELSEM 2 by frequency range and the fact it does not include sea-ice.

### First Guess Comparisons

Over land surfaces; ATMS ch. 2 (31.4 GHZ); Bias correction applied; QC applied



0.800	0.833	0.867	0.900	0.933	0.967	1.000

	CRTM	TELSEM 2
Number count	4104	8050
Bias	-0.4	0.05
Std. dev.	2.0	1.9

# 2. Allow the emissivity to vary through the assimilation process - control variable -

- To allow for improved physically-balanced state vector, emissivity is introduced as a control variable into the GSI system. The approach uses "de-attached" stand-alone implementation.
- The optimal emissivity parameter is obtained by minimizing the incremental cost function
- Minimization involves calculation of the gradient with help of the chain rule
- Implementation is applied to all MW sensors

#### Possible improvements to implementation:

- Correlated errors for emissivity (including non-diagonal elements of covariance matrix)
- Exponential Multiplicative parameter (preserve positive definiteness of the emissivity multiplicative parameter)

## Why separating the ATMS impact?

- Advanced Technology Microwave Sounder (ATMS) Suomi NPP ۲
- Land surface MW sensitive frequencies: CH 1-4, 16-22 ٠
- GSI QC improvements were possible ٠
- FSOI shows importance of ATMS sensor in fractional impact summary (shown month: May 2019) ۲



0.3

GMAO Observation Impact Summary (http://ios.jcsda.org/)

## Adding the emissivity to the GSI as a control variable - implementation -

New control variable is a multiplicative factor eta :

$$e = \beta \cdot e_{CRTM}$$

$$\boldsymbol{\beta} = \left( \begin{array}{ccc} \boldsymbol{\beta}_1 & \boldsymbol{\beta}_2 & \cdots & \boldsymbol{\beta}_N \end{array} \right)^T$$

Where:

- e denotes emissivity
- subscript *CRTM* refers to the emissivity calculated in CRTM
- $\beta$  is a vector defined over channels in observation space
- *N* is the total number of channels
- Construct cost function with added parts to account for new control variable  $\beta$  information
- Calculate gradient of the cost function
- Without knowing cross correlations between different parameters the initial choice for parameter error covariance is a diagonal matrix, where  $\sigma$  is the assigned standard deviation of the particular parameter. In case of emissivity, this is determined based on TELSEM2.
- Optimal emissivity parameter is obtained by minimizing the cost function.

#### TELSEM 2 Emissivity Variance for 50.3 GHz



0.0001 0.0011 0.0021 0.0030 0.0040 Emissivity Variance

9

0.0050



#### Model performance sanity check – 15 days



## 3. Improving QC for ATMS over land

• Analytic emissivity used for emissivity sensitivity

$$\varepsilon = \frac{T_b - Tu - Td\Gamma}{(T_s - Td)\Gamma}$$

- ε emissivity
- Tb observed brightness temperature
- Ts skin temperature
- Tu upwelling brightness temperature
- Td downwelling brightness temperature
- atmospheric transmittance
- Screening for rain (information from GPROF)

Regression is trained on a week od data in 2016; shown here are 10 days in July 2018 (for validation).

Observed GPROF rain and regressed ATMS TB89-TB31 rain.

Crosses are relationship between 'TB89-TB31' and 'GPROF rain'. Regression is done as polynomial of 3<sup>rd</sup> order.



CH number Correlation to Ts based on OmB\_nBC

- 1 -0.227570
- 2 -0.181155
- 3 -0.312302
- 4 -0.314278
- 16 -0.658272







#### Results

## Comparison of O-B and O-A before and after bias correction



- The best experiment performance (exp51) compared to the operational run result before (top) and after (middle) bias correction
- Observation counts exceed the operational values by factor of 4.
- Mean O-B and O-A show significant improvements after the bias correction
- Channels 7 through 15 (not shown) preserve the same performance

#### NOTE:

*Exp 51:* TELSEM as a background new QC constructed of precipitation, Ts and emissivity sensitivity screening with emissivity as analysis variable.

*Cntr ModelEm*: CRTM provides emissivity which is not treated as a control variable

### **Comparison of valid observation points**

ATMS over land; both ascending and descending orbits



Valid QC	Exp51 STD	Cntr STD	Counts
Exp51	1.96	7.80	28,780
Cntr	5.36	1.91	7,039
Exp51 & Cntr	1.82	1.87	4,255

Valid QC	Exp51 STD	Cntr STD	Counts
Exp51	0.26	1.28	48,295
Cntr	0.47	0.23	7,039
Exp51 & Cntr	0.19	0.23	6,868
			14

## **Conclusions**

- Using different background in the system makes biggest change to the assimilation
- Using control variable will not increase significantly observation counts as the background will but will make better statistics, meaning the adjustment was in right direction
- QC optimization is necessary with introduction of new or improved background

#### New future directions:

- Merging this work with FV3 is in progress and once done we will evaluate the implementation for forecast impact in longer verification runs
- Optimize Quality Control (QC) and observation errors for land surface sensitive microwave brightness temperatures assimilation
- Optimize surface emissivity background error covariance and background emissivity
- Optimize bias correction implemented for land surface sensitive microwave brightness temperatures

## **Questions?**