#### Investigating the impact of assimilating all-sky window-channel infrared radiances from GOES-ABI and Himawari-AHI on cloud analyses and forecasts

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## All-sky satellite radiance DA

- Advances in the assimilation of satellite radiances allowed us to **directly** assimilate cloud-**free** and cloud-**affected** observations
- □ Significant improvements in analyses and weather forecasts has been found especially using **cloud**-affected microwave radiances (e.g., Geer et al., 2011, 2017, 2018; Zhu et al., 2016, 2019)
- □ Efforts towards the assimilation of **cloud-affected infrared radiances** (e.g., Okamoto et al., 2014; Geer et al., 2019; Zhu et al., 2022; Degelia et al., 2023), mostly using **water vapor bands** 
  - □ exhibit more Gaussian characteristics than window channels (Okamoto, 2017)
- Cloud-affected infrared radiances from **window channels remains a challenge**:
  - Observation error distribution
  - ❑ Non-linear observation operators
  - Variational bias correction
  - Quality control procedures
  - Cloud representation in forecast models



# All-sky satellite radiance DA in MPAS-JEDI

- → Leverage several quality control filters, variational bias correction, and observations operators in UFO:
  - Community Radiative Transfer Model (CRTM) interfaced in JEDI-UFO:
    - All sensors and conditions (Liu and Collard, 2019)
    - Hydrometeor types: water, ice, rain, snow and hail and graupel
- → 3DVar, 3D/4DEnVar, hybrid data assimilation, EDA, LETKF (recent, under testing) capabilities
- → Analysis variables: temperature, specific humidity, zonal and meridional components of horizontal velocity, surface pressure, mixing ratios of cloud liquid water, cloud ice, rain, snow, and graupel
- → Assimilate aircraft, sondes, surface (pressure), derived atmospheric motion winds (AMVs), GNSS radio occultation, radiances (AMSUA, MHS, ATMS, ABI, AHI)
  - AMSU-A window channels in all-sky scenes (Liu et al., 2022)
  - ATMS water vapor and window channels in all-sky scenes (Ban et al., 2023)
- → Cycling experiments, HofX



# Objective

Examine the assimilation of infrared window channel 13 from GOES-ABI and Himawari-AHI sensors, using the Model for Prediction Across Scales – Atmosphere (MPAS-A) coupled with the Joint Effort for Data assimilation Integration (JEDI)

- Estimate observation errors for ABI and AHI channel 13
- Estimate observation bias for bias correction
- Run cycling experiments
- Preliminary verification of cloud analysis and forecasts



## **Observation error model**

Symmetric cloud impact (SCI) by Okamoto et al (2014):

$$CI_{O} = BT_{y,BC} - BT_{clr}$$
$$CI_{M} = BT_{x} - BT_{clr}$$
$$SCI = \frac{1}{2} (|CI_{O}| + |CI_{M}|)$$

**BT<sub>v</sub>**, **BT<sub>x</sub>**: observed and simulated brightness temperature

**BT**<sub>clr</sub>: clear-sky background brightness temperature without considering cloud-scattering

Cl<sub>M</sub>, Cl<sub>o</sub>: cloud impact on model and observation

OmB calculated using HofX:

- 30-km forecast from hybrid 3DEnVar cycling experiment
- Observation without bias correction
- Only over water
- Maximum sensor zenith angle: 65.0
- 00Z 20 April 18Z 14 May 2018



#### **Observation error model**



Parameterized ABI and AHI observation errors using a piece-wise linear function of UFO (ObsErrorModelRamp): https://jointcenterforsatellitedataassimilation-jedi-docs.readthedocs-hosted.com/en/stable/inside/jedicomponents/ufo/qcfliters/obsfunctions/ObsErrorModelRamp.html#obserrormodelramp



ABI x0 x1 err0 err1 clrbias 13 0.40 34.20 0.72 31.46 -1.13

#### AHI x0 x1 err0 err1 clrbias 13 0.20 33.40 0.57 28.98 -0.56

clrbias is used for constant bias correction for each sensor



# **Experiments design**

- → MPAS-JEDI v2.0:
  - 00Z 15 April 2018 00Z 20 April 2018
  - ◆ Hybrid 3DEnVar
  - 30km-60km dual resolution, 80-member EDA forecast
  - ◆ 2 outer loops with 60 iteration each
  - CRTM observation operator
    - with cloud scattering effect for ABI and AHI channel 13 with no humidity sensitivity
  - AHI and ABI channel 13:
    - thinned on a 145-km mesh
    - only over water
- → MPAS-A:
  - non-hydrostatic dynamical core
  - unstructured mesh
  - height-based terrain-following vertical coordinate
  - 55 levels, 30km top
  - quase-uniform 30 km grid
  - "mesoscale reference" physical parameterizations

Exp. name	Observations			
benchmark	Aircraft, AMVs wind, surface (pressure), sondes, GNSS RO bending angle	clear-sky AMSU- A (MetOp A, MetOp B, NOAA- 15, NOAA-18, NOAA-19) - VarBC*	clear-sky MHS (MetOp A, MetOp B, NOAA-18, NOAA-19) - VarBC*	
ch13_raw	$\checkmark$	$\checkmark$	$\checkmark$	ABI GOES-16 channel 13 AHI Himawari-8 channel 13
ch13_constBC	$\checkmark$	$\checkmark$	$\checkmark$	ABI GOES-16 channel 13 AHI Himawari-8 channel 13 - constant offset bias correction

\*Predictors: constant offset, lapse rate, emissivity, scan angle 2, scan angle 3 and scan angle 4

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### **Observation space verification**



RMS (OmB) > RMS(OmA)



### **Observation space verification**



RMS (OmB) > RMS(OmA)



# **Observation space verification: 6-h forecast**

ABI

AHI



→ Experiment with channel 13 leads to improvements in window and water vapor channels for ABI and AHI

RMS < 0 improvement > 0 degradation



#### MPAS 6-h fcst vs GFS analysis STD <0 improvement >0 degradation



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(f) ch13 constVarBC (i) ch13 constVarBC (I) ch13\_constVarBC

(c) ch13\_constVarBC

→ More improvements are found in experiment with constant bias correction

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#### **MPAS 6-h fcst vs GFS analysis**

#### < 0 improvement **STD** > 0 degradation

#### **ABI region**



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![](_page_11_Figure_6.jpeg)

18

![](_page_11_Figure_7.jpeg)

2018-Apr

![](_page_11_Figure_8.jpeg)

![](_page_11_Figure_9.jpeg)

![](_page_11_Figure_10.jpeg)

16 17 18

![](_page_11_Figure_12.jpeg)

#### **Observations vs. Day-3 forecast**

![](_page_12_Figure_1.jpeg)

AHI channel 13 BTs (degree C) valid at 00 UTC 21 April 2018

![](_page_12_Picture_3.jpeg)

#### **Observations vs. Day-2 forecast**

![](_page_13_Figure_1.jpeg)

AHI channel 13 BTs (degree C) valid at 00 UTC 21 April 2018

![](_page_13_Picture_3.jpeg)

#### **Observations vs. Day-1 forecast**

![](_page_14_Figure_1.jpeg)

AHI channel 13 BTs (degree C) valid at 00 UTC 21 April 2018

![](_page_14_Picture_3.jpeg)

#### **Takeaways**

- Demonstrated our capability to successfully do all-sky infrared DA with MPAS-JEDI
- Calculated observation errors and constant bias for ABI and AHI channel 13
- Assimilated ABI and AHI channel 13 with and without constant bias correction in a cycling experiment for 5 days
- More positive impacts are found in the MPAS 6-hr forecast when assimilating channel 13 with bias correction with promising results for longer forecast
- More work is needed in terms of the variational bias correction taking into account more adequate predictors (Okamoto et al., 2023 uses SCI as a predictor) and the observations operator

![](_page_15_Picture_6.jpeg)

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![](_page_16_Picture_15.jpeg)

#### Thank you! Comments/questions?

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![](_page_17_Picture_3.jpeg)

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![](_page_17_Picture_5.jpeg)

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