

Exploring Using Artificial Intelligence (AI) to Exploit Big Satellite Data for NowCasting and NWP

- Focus on JPSS-related Applications/Sensors-

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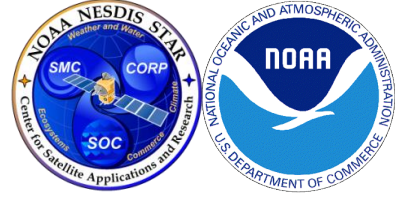
~ University of Miami, Miami, FL, USA

#Aerospace Co, Arlington, VA

^ Atmospheric and Environmental research (AER Inc.)

& University of Maryland, College Park, MD

Agenda



1

Why Artificial Intelligence (AI) ? Background and Motivations

2

What are we exploring using AI for?

3

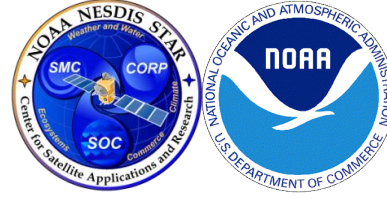
JPSS-related Examples of AI Applications:

- Remote Sensing Algorithms (next-Gen Algorithms)
- Radiative Transfer
- Data Fusion / Nowcasting & Data Assimilation
- Exploiting Satellite data for Supporting Prediction
- Morphing

4

Conclusions

Trends in Global Earth Observation Systems



• GOS Trends:

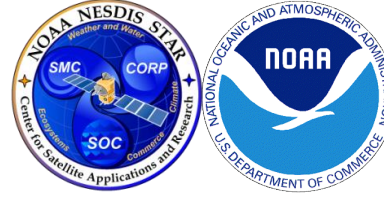
- New Players in GOS (international, commercial, etc)
- New Sensors (higher resolutions, etc)
- New technologies (small sats, etc)
- Emergence of New GOS (IoT, etc)
- **Significant Increase in volume and diversity of data**

• Parallel Trends

- Budget, HPC Constraints
- Higher societal impact and expectations
- Higher users expectations
- Demand for Increase in quantity of data assimilated (5% currently assimilated)



Evolution of Environmental Data Exploitation



This evolution applies to all areas of computing.

It has led several major companies to adapt their business models to take advantage of AI

DIARY OF THE WEATHER at Fort Edwards (Illinois)
for the quarter ending the thirtieth day of September 1843.

MONTH.	TEMPERATURE.			WIND.	WINDS.	REMARKS.
	VEL.	11	IX			
July	A. M.	P. M.	P. M.			
1st						
2d						
3d						
4th						
5th						
6th						
7th						
8th						
9th						
10th						
11th						
12th						
13th						
14th						
15th	76	59	50	S.	Clear	
16th	75	87	77	S.	Cloudy Rain P. M.	
17th	76	87	74	S. E.	Cloudy Rain A. M.	
18th	73	84	77	N. E.	Fair	
19th	74	83	76	N. E.	Fair	

No Thermometer records at post till the 14th Inst.

Tabulating Systems Era (1800-1940s)
Manual Measurement and analysis



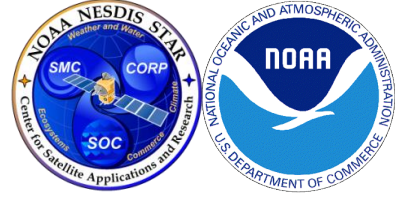
Programmable Intelligence Era (1950-Now)
Transformational efficiency in Applying Human Intelligence



Cognitive Computing –AI Era (2011- Foreseeable Future)
Enhancement/Augmentation of Human Intelligence

Credit: Materials adapted for Environmental Observations specifically, inspired from an IBM presentation by Dr John Kelly, senior vice president, cognitive solutions, to the NOAA Science Advisory Board (SAB) on November 2016

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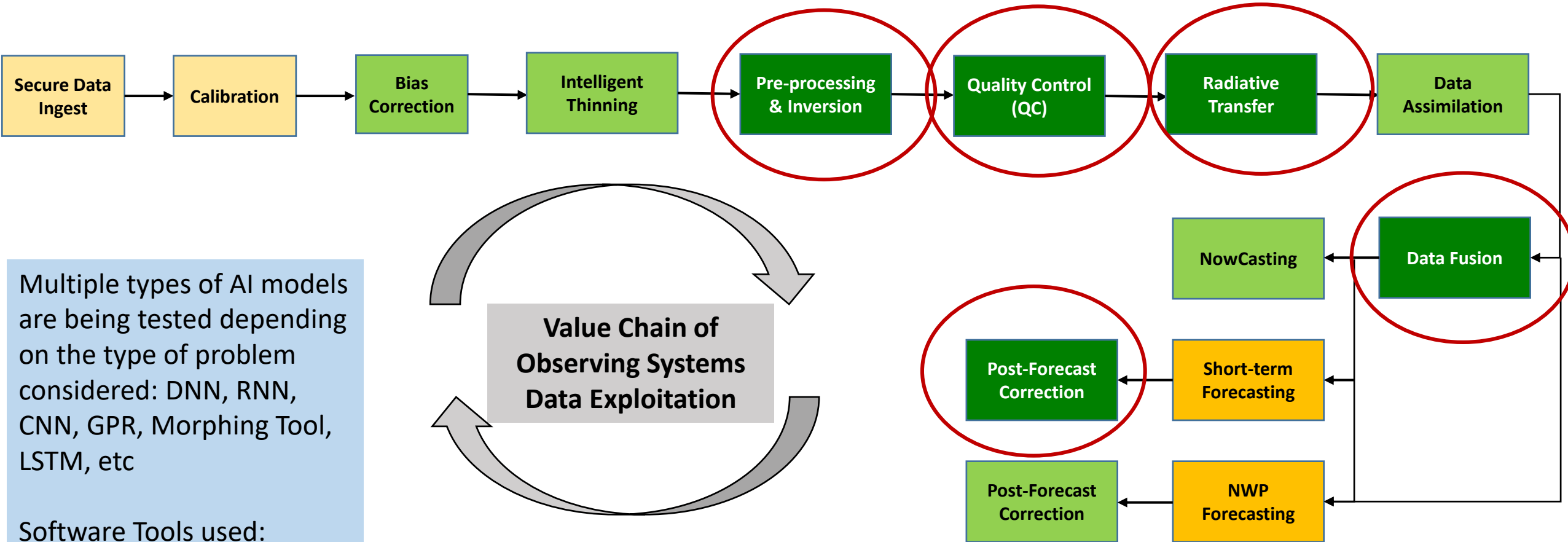
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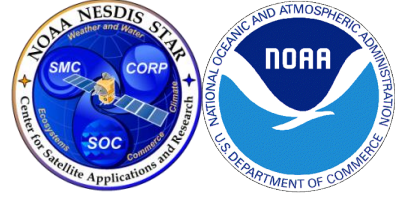
Exploring AI for Remote Sensing, NWP & Nowcasting/Situational Awareness (SA). Status



Multiple types of AI models are being tested depending on the type of problem considered: DNN, RNN, CNN, GPR, Morphing Tool, LSTM, etc

Software Tools used: TensorFlow and KERAS

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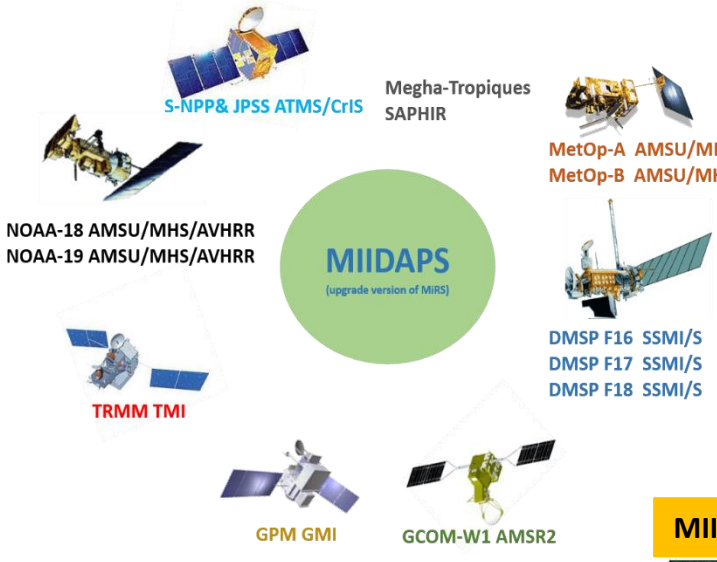
4

Conclusions

Pilot Project: MIIDAPS-AI:

Multi-Instrument Inversion and Data Assimilation Preprocessing System

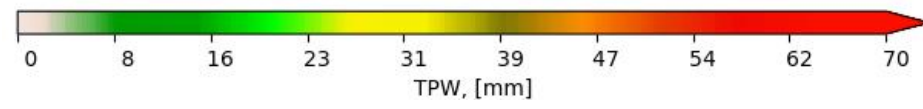
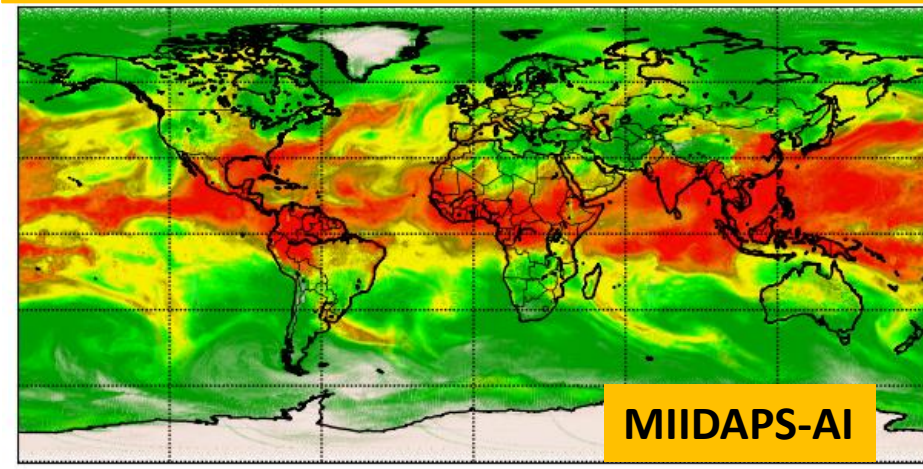
Use of Deep Neural Network (DNN) for Geophysical Retrieval and Quality Control Purposes



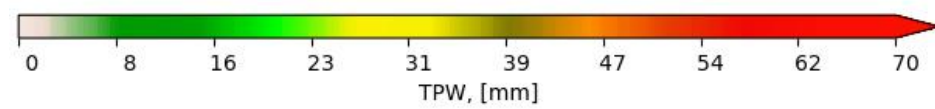
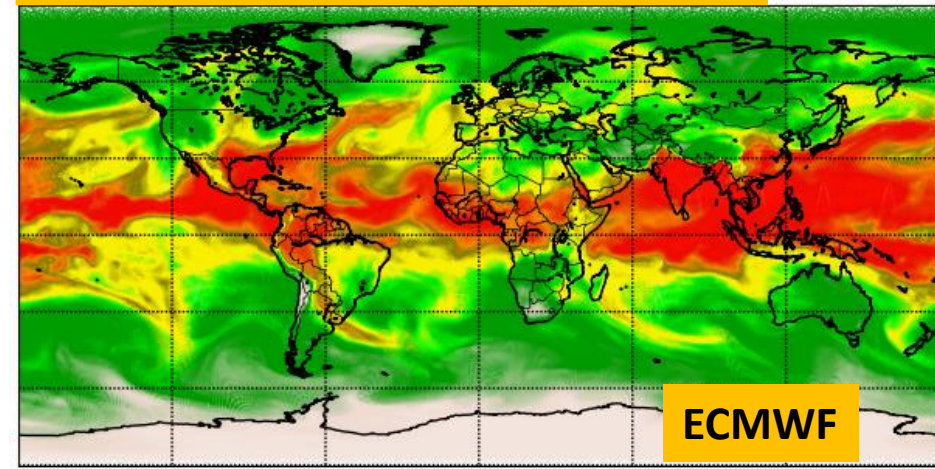
Google TensorFlow Tool used for MIIDAPS-AI

- How to assess that AI-based output (Satellite Analysis) is valid?**
- (1) Assessing quality by comparing against independent analyses
 - (2) Assessing Radiometric Fitting of Analysis
 - (3) Assessing analysis spatial coherence
 - (4) Assessing inter-parameters correlations

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

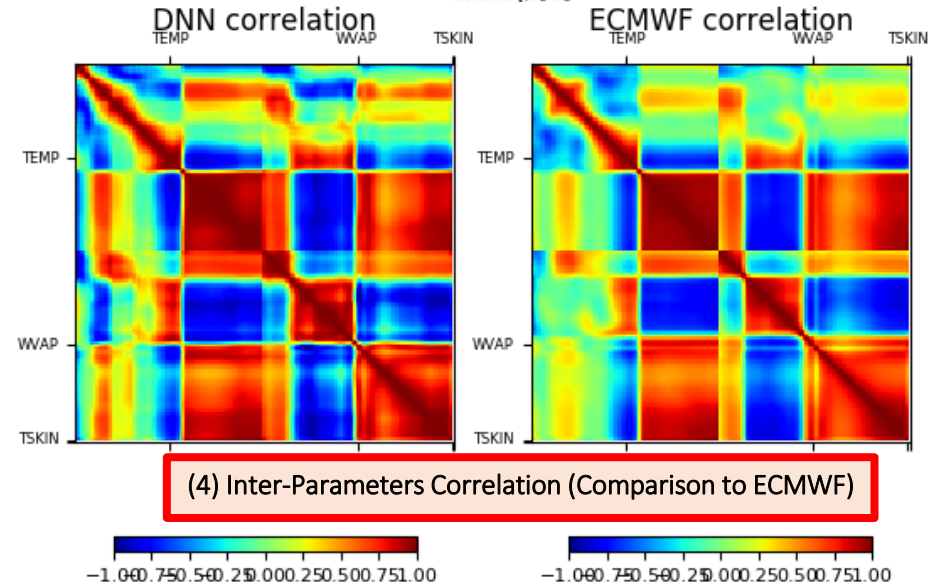
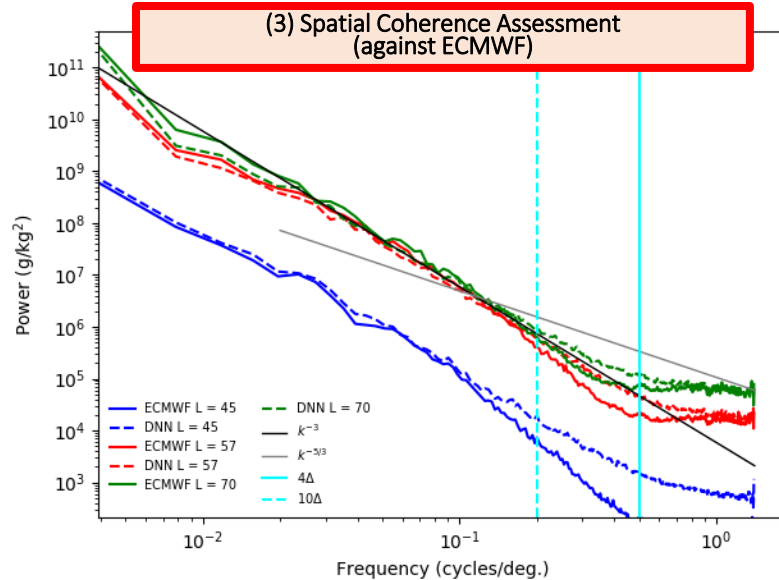
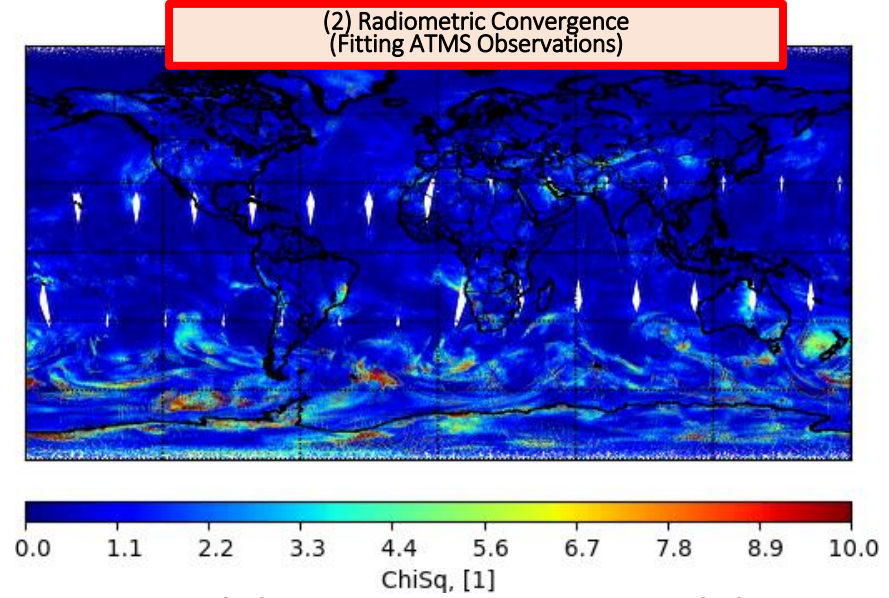
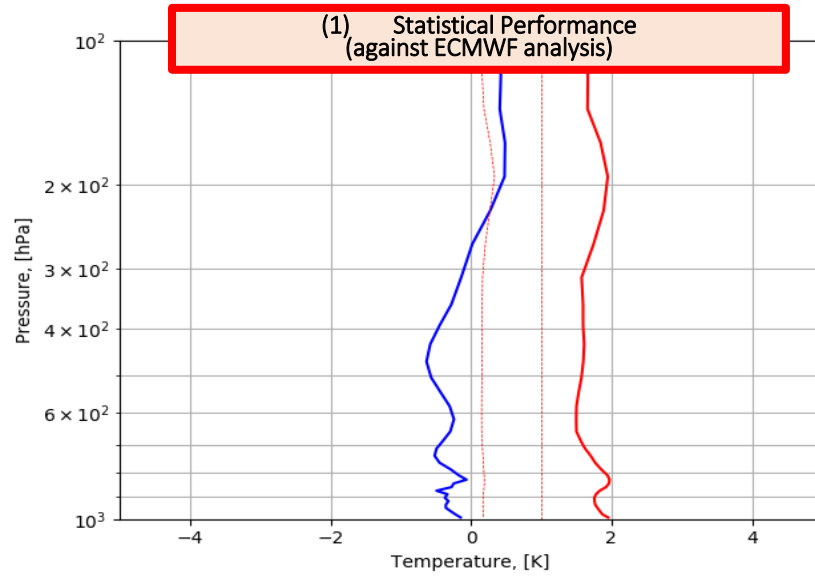
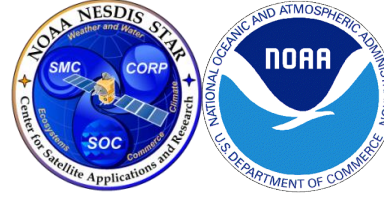


Reference source of TPW: ECMWF Analysis

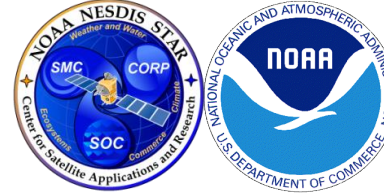


	MIIDAPS-AI	MIIDAPS
Processing Time for a full day data. A single sensor (ATMS). Excluding I/O	~5 seconds	~ 2 hours

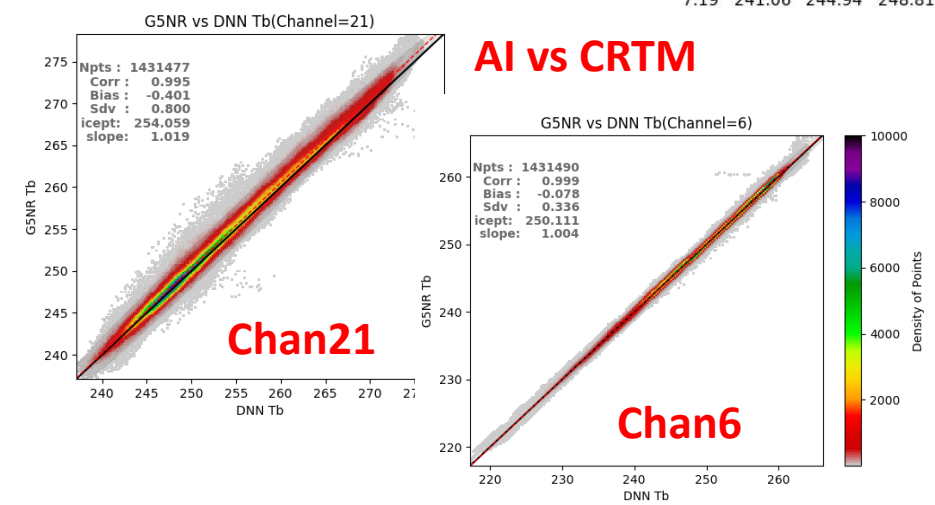
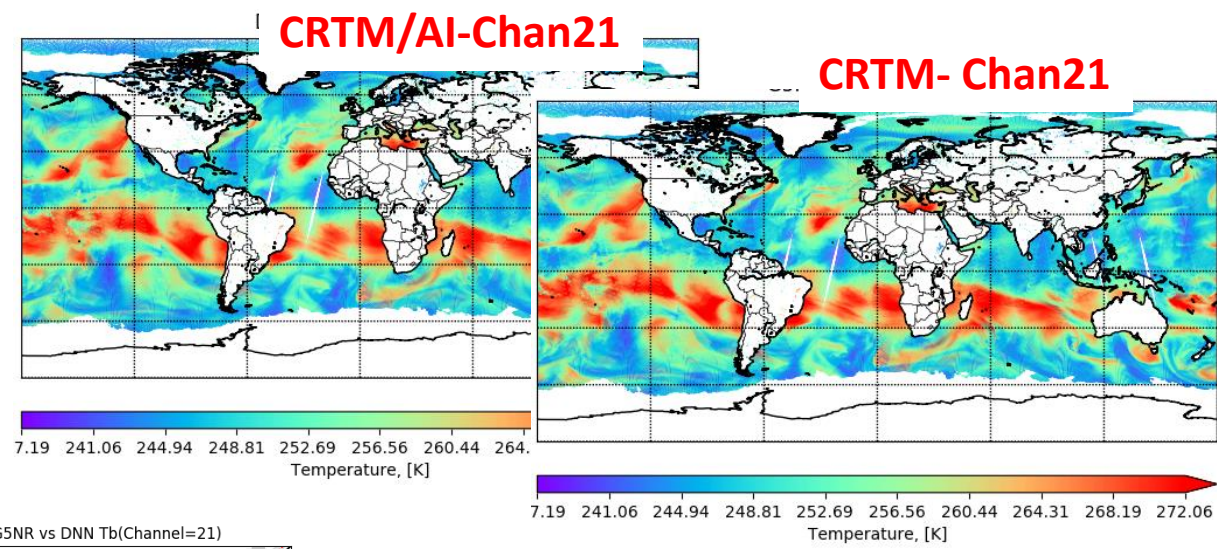
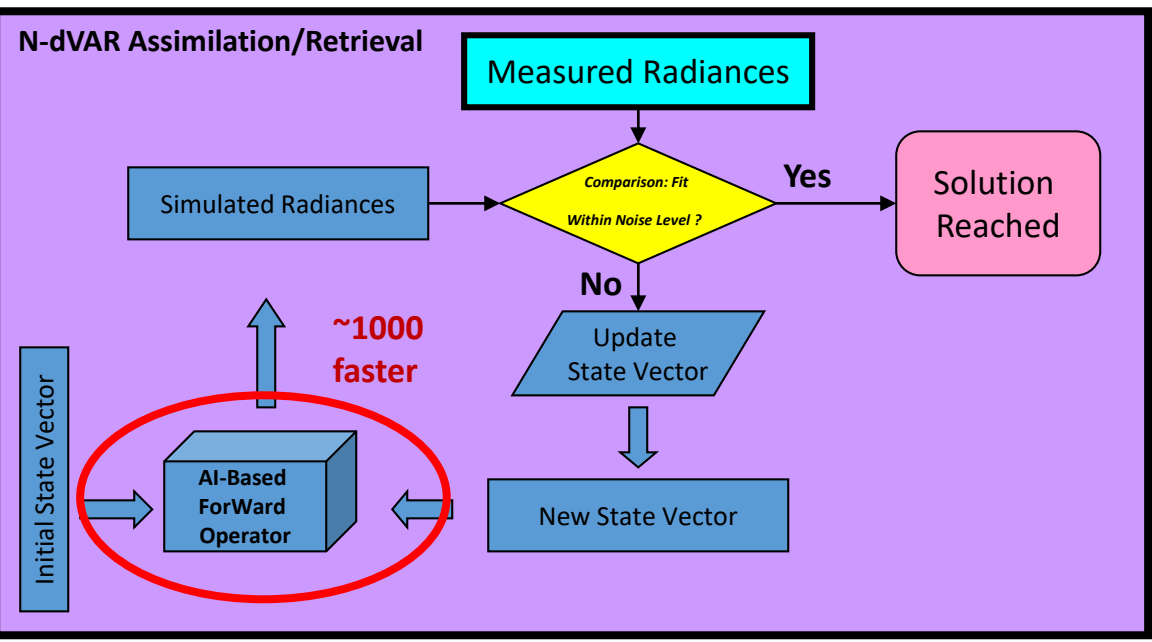
How do we know if this AI-based satellite-analysis is scientifically valid?



Can AI Be Used as Forward Operator?



Use of Deep Neural Network (DNN) for Radiative Transfer Modeling Purposes



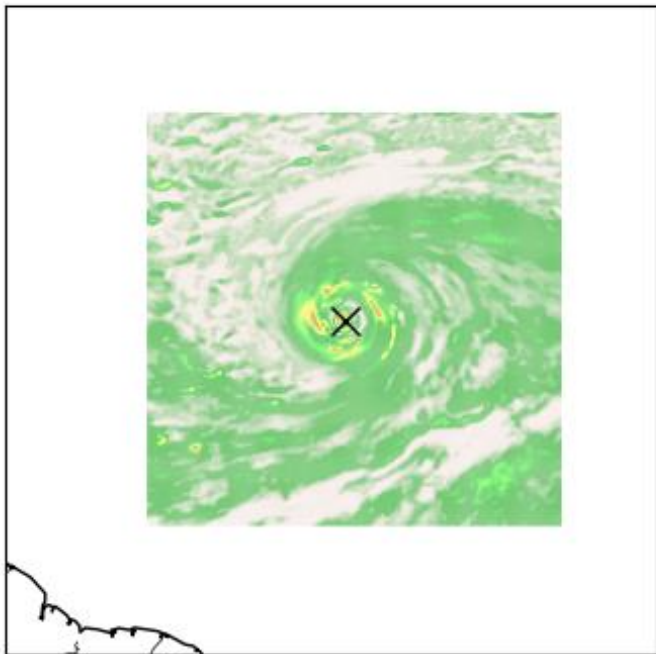
	CRTM-AI	CRTM
Processing Time for a full day data. A single sensor channel(ATMS). Excluding I/O	<1 second	~ 1.3 hours

Can AI Tools Be Used for Data Fusion & Data Assimilation?

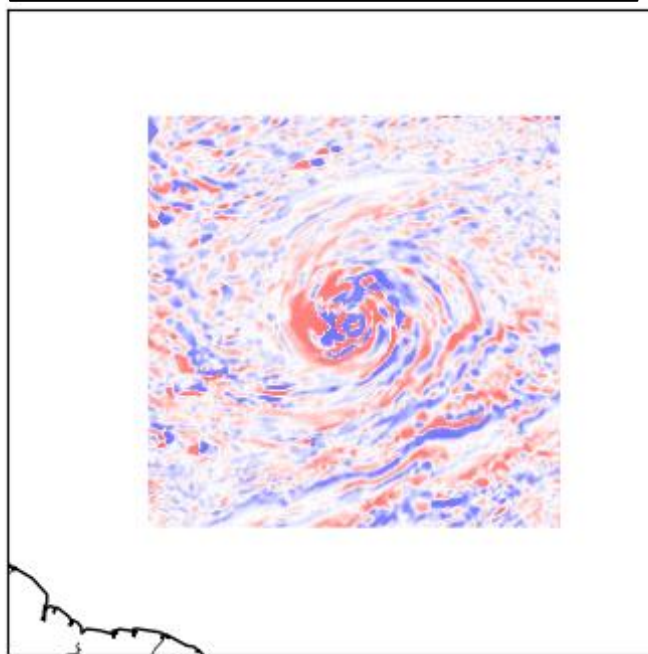
Use of GPR (Gaussian Process Regression) AI Model for Data Fusion/Assimilation (Case of Hydrometeors/Clouds)



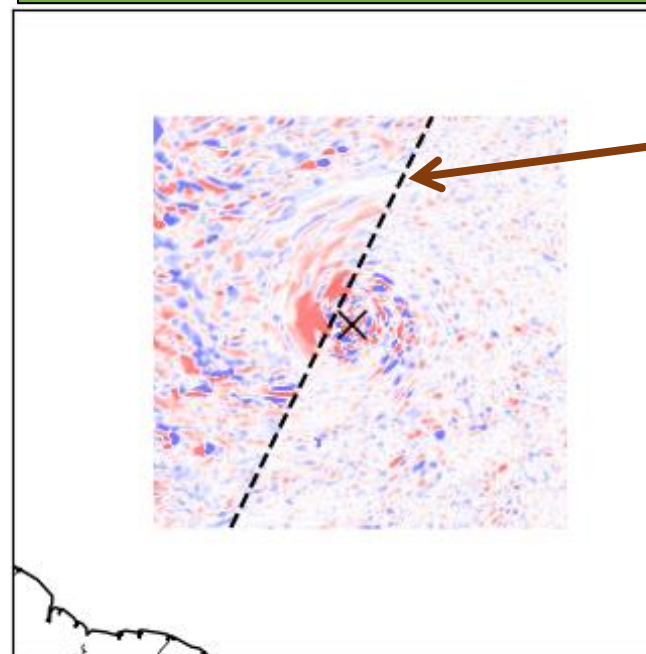
GPR Analysis



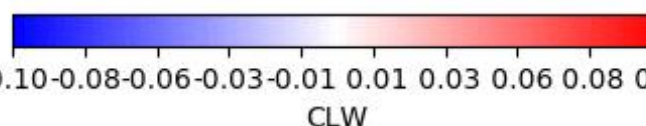
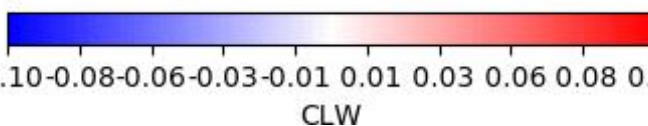
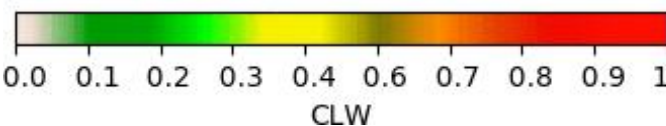
Analysis - Background



Analysis - GPR



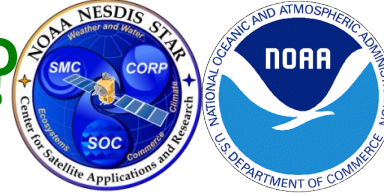
Simulated observations sampled on right side of dashed line.



- Fused GPR analysis of total cloud water matches analysis where observations are dense and relaxes to the background field where they are sparse.
- Some distortion near the center of the hurricane is evident in the GPR fields and is due to the sampling.

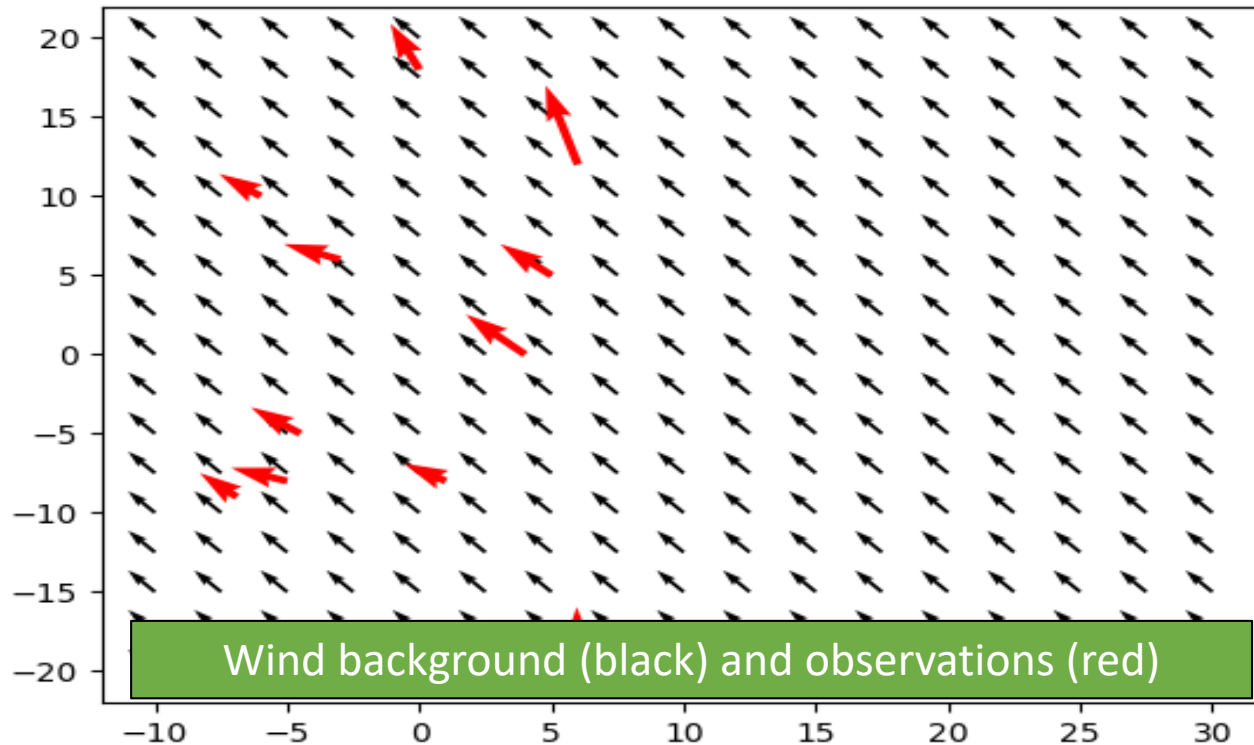
Can AI Tools Be Used for Data Fusion & Data Assimilation?

Use of GPR (Gaussian Process Regression) AI Model for Data Fusion/Assimilation (Case of AMV)

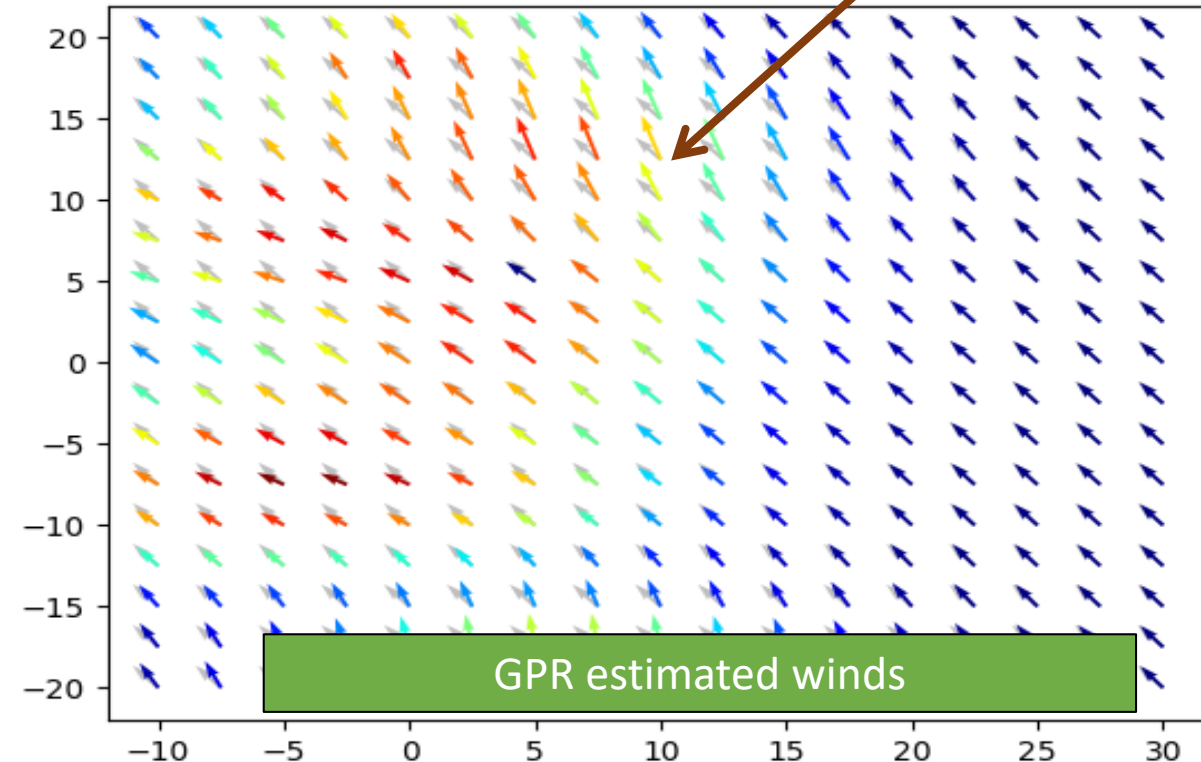


Color confidence/error estimates

Background and Measurements



GPR-Based Analysis



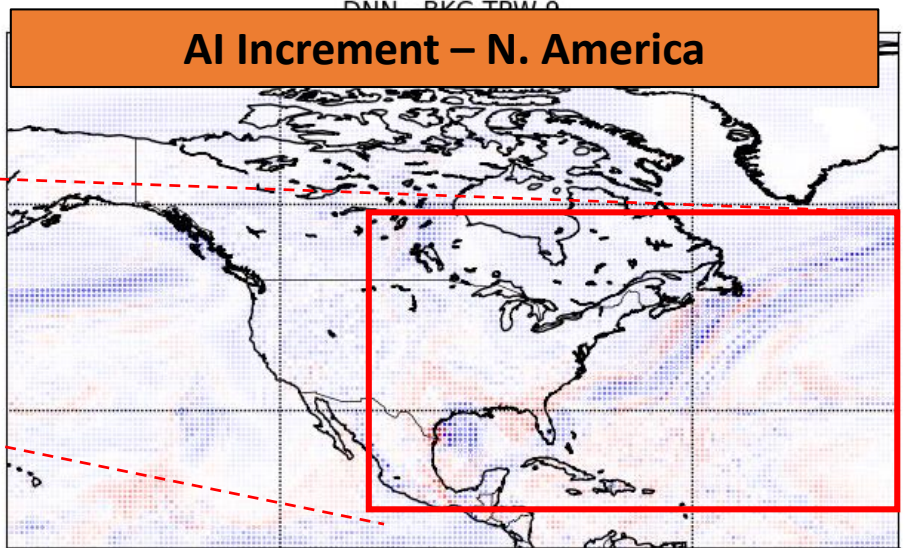
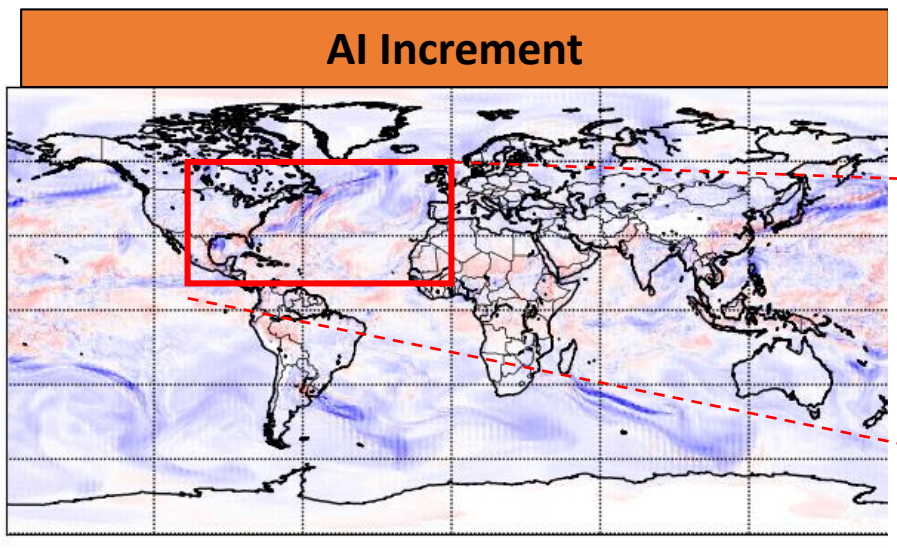
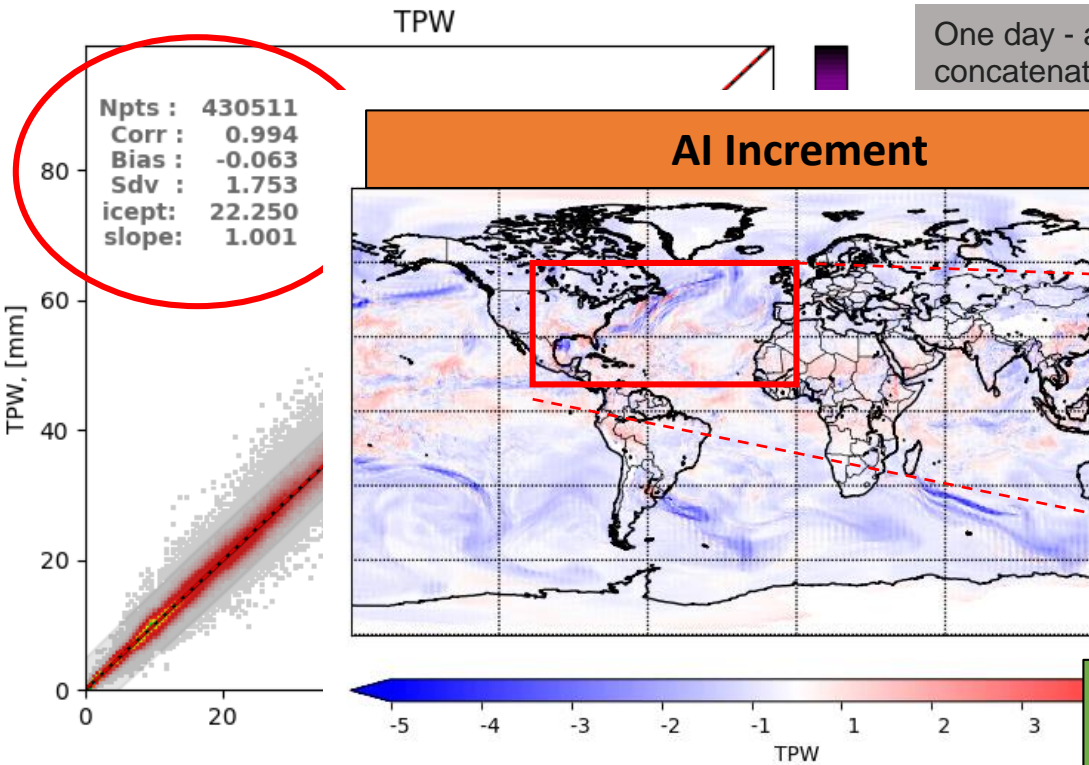
- Synthetic wind observations (red) are injected onto background (black) fields and GPR used to “fuse” the two.
- Color code corresponds GPR confidence – warmer colors reflect high confidence, while colder colors reflect low confidence estimates – and are consistent with observation locations.

Correcting TPW Forecasting with AI?

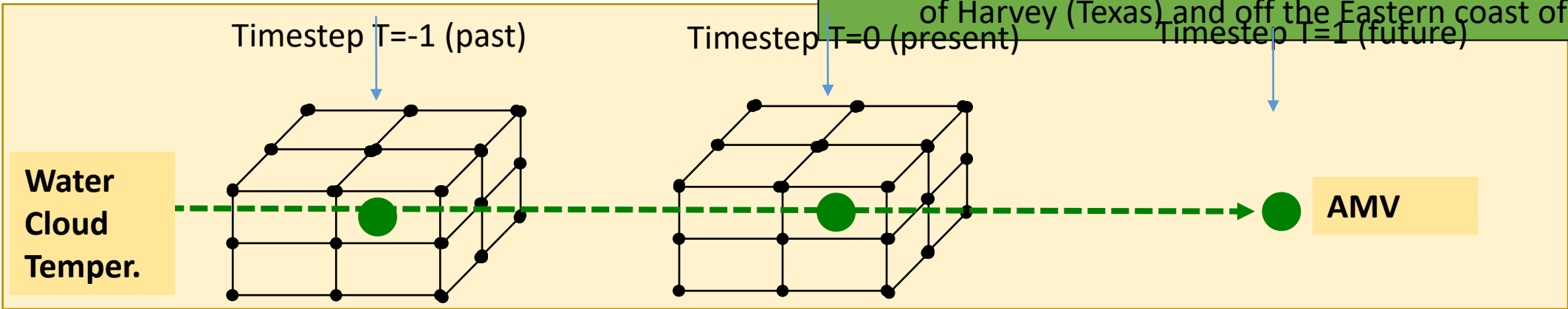


ECMWF vs AI-corrected 6h fcst valid @ECMWF analysis time

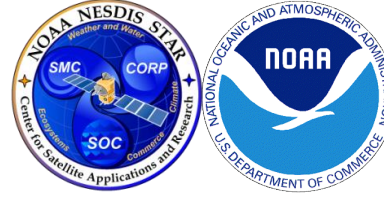
ECMWF vs 6 hr frkst valid @ECMWF analysis time.



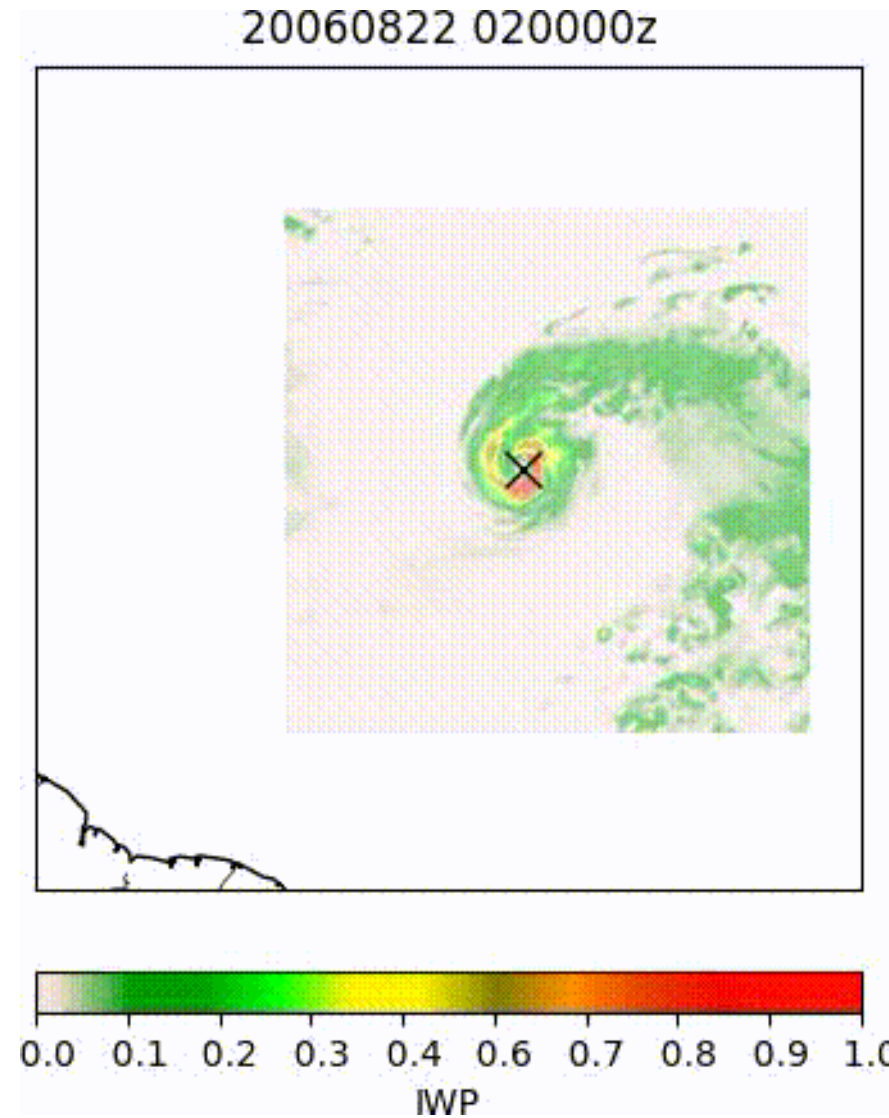
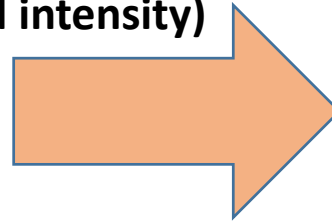
AI increment shows some dipoles indicating that the correction is adjusting the position of some features – Most notably the position of Harvey (Texas) and off the Eastern coast of N.America



Use of “morphing” AI Tool (“dogs” video morphing software) for Cloud/Precip morphing



Note the potential for morphing both the shape and color (i.e. equivalent of track and intensity)

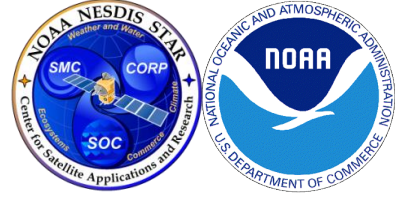


- Used total integrated cloud ice from NASA GEOS-5 Nature Run (G5NR) “AL01” tropical cyclone at two time-steps (0200z and 0600z).
- Morphing software applied as a black box with some hand tuning of transformations between the two images.
 - Image at right sampled using 20 transformations between images

Credit: Example output and software from:

<http://andrew.gibiansky.com/blog/image-processing/image-morphing/>

Conclusions



❖ **Big Data Challenge already here**

❖ **AI/ML approach is a natural evolution of how to exploit data**

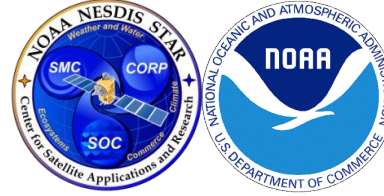
(think evolution of Programming languages: Assembler, Basic, Pascal, F66, C++, F00, Java, Python, ...to TensorFlow, Keras,)

❖ **Goal of this study is not to show AI/ML approach can do better, but that it can provide at least similar quality, much faster, therefore it can process more data.**

❖ **Significant potential to leverage AI tools and models developed in other fields, to our field: for remote sensing, radiative transfer, data fusion, morphing, etc.**

❖ **AI has the potential to be a transformational new approach in our exploitation of Big Environmental data**

For More Information....



Upcoming Workshop on use of AI in Earth Observation and NWP

meeting_2019AIWorkshop.php

Calculate distance an **Step** Computing Distances: Calculation of Distanc: STAR JPSS - Instrum: Plotlyjs Function Refi

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Workshop: Using AI to Exploit Big Data in Satellite Earth Observation & Numerical Weather Prediction

Sponsors: STAR, NWS, and OAR
23-25 April 2019
NOAA Center for Weather and Climate Prediction Conference Center
5830 University Research Court • College Park, MD 20740

[Main meeting page](#)

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Workshop Summary

The workshop will help gather scientists, software engineers, program managers, and leaders, from public and private sectors, versed or interested in the application of AI in the field of environmental data information (from space- and ground-based platforms) and NWP data assimilation and forecasting and other environmental prediction systems. This will facilitate the cross-fertilization of knowledge in the different communities to benefit the environmental observation and numerical prediction using machine learning techniques. This gathering is expected to allow participants to exchange ideas, share learned lessons, and establish collaborations to:

1. Further the scientific objectives of the Earth observation and NWP prediction skills;
2. Improve efficiency of environmental data processing and exploitation;
3. Identify innovative ways to use satellite data and other environmental data to create new products and services and generate new markets;
4. Expand commercial markets of high-level environment-related products and services.

	Tuesday, 23 April	Wednesday, 24 April	Thursday, 25 April
0830 - 0900	Check-in at NCWCP Conference Center	Check-in at NCWCP Conference Center	Check-in at NCWCP Conference Center
0900 - 0930	Opening Remarks	Opening Remarks	Opening Remarks
0930 - 1030	Session I	Session I	Session I
1030 - 1050	Break	Break	Break
1050 - 1200	Session II	Session II	Session II
1200 - 1300	Lunch	Lunch	Lunch
1300 - 1530	Session III	Session III	Session III
1530 - 1550	Break	Break	Break
1550 - 1700	Session IV	Session V	Session V

schedule of sessions (click to enlarge)

