Ensemble-based Data Assimilation for Tropical Cyclones: Current Status and Future Prospects

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Frontiers in Ensemble Data Assimilation Workshop 5 August 2015 Boulder, CO

Outline

- Tropical Cyclone Overview
- Unique TC Observations
- Challenges with TCs
- Future prospects



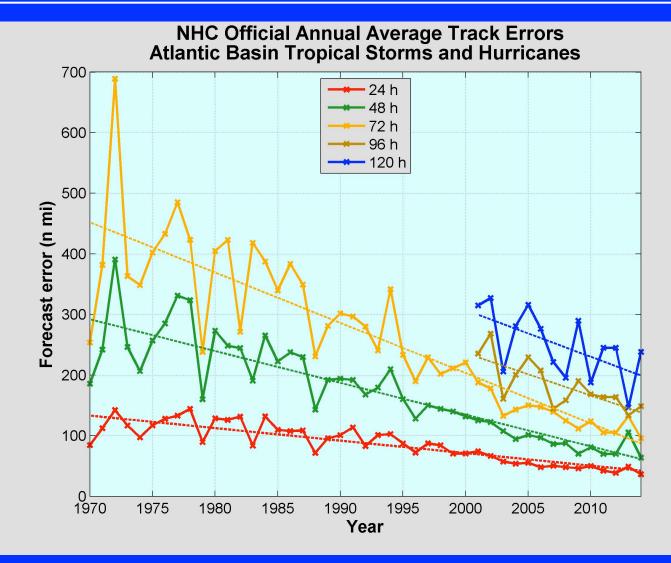
Source: Jim Brooks



Source: Metropolitan Transit Authority

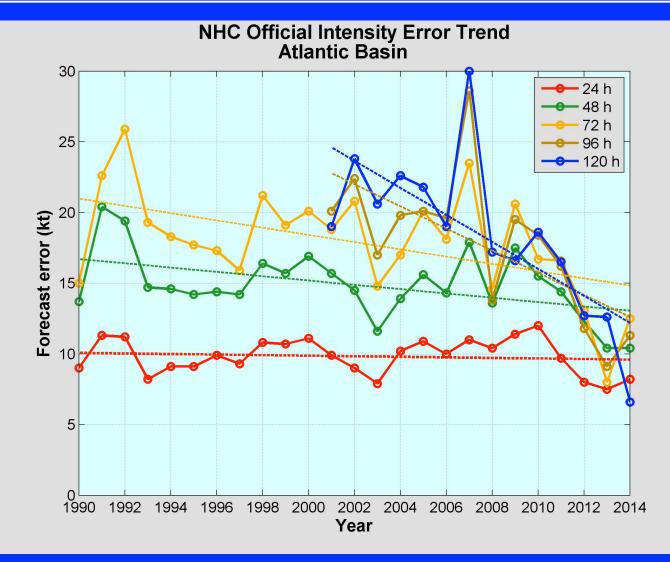
Source: AP

TC Track Errors



National Hurricane Center

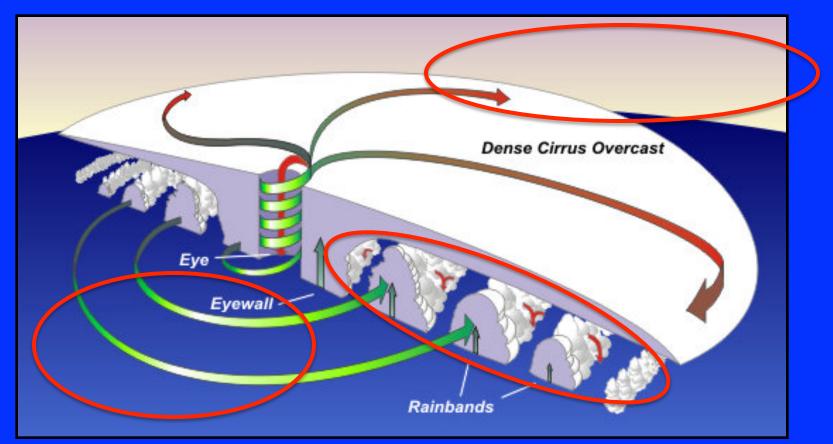
TC Intensity Errors



National Hurricane Center

Intensity Factors

TC Environment



Courtesy National Hurricane Center

Surface Boundary

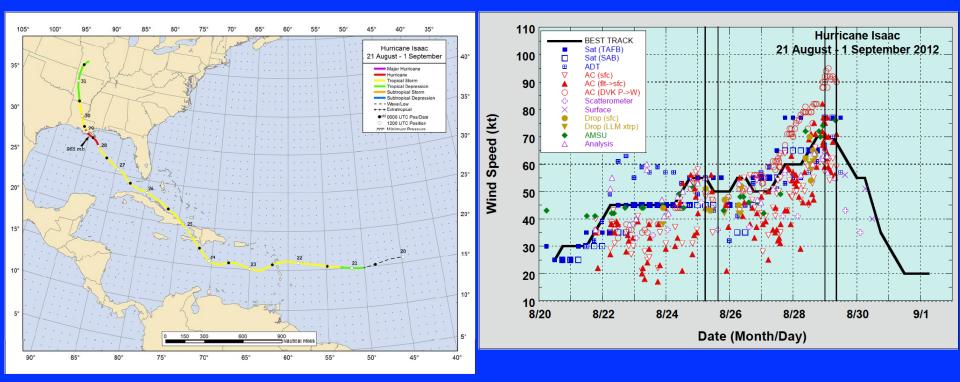
Internal Processes

TC Observations

- One hypothesis for how to improve TC intensity forecasts is to assimilate more observations in their viscinity
- TCs have some unique datasets
 - TC "Vitals"
 - Aircraft Data
 - Satellite-derived wind vectors

TC "Vitals"

Hurricane Isaac (2012)

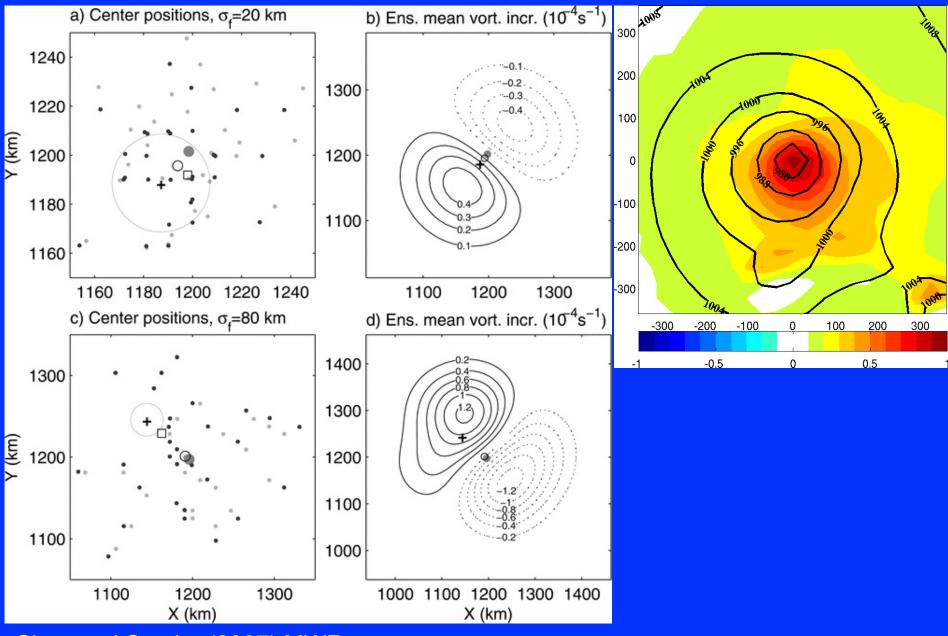


NHC 09L ISAAC

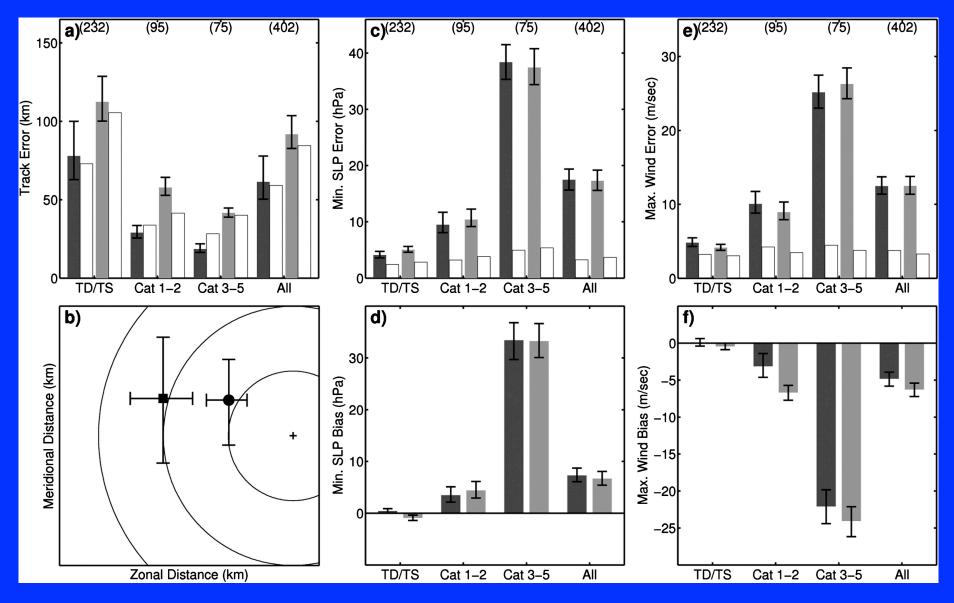
20120828 1200 278N 0882W 305 031 0976 1012 0556 31

Position Assimilation

MSLP Assimilation



Chen and Snyder (2007) MWR



Torn 2010, MWR

ZCZC MIATCMAT3 ALL TTAA00 KNHC DDHHMM

HURRICANE SANDY FORECAST/ADVISORY NUMBER 17 NWS NATIONAL HURRICANE CENTER MIAMI FL AL182012 1500 UTC FRI OCT 26 2012

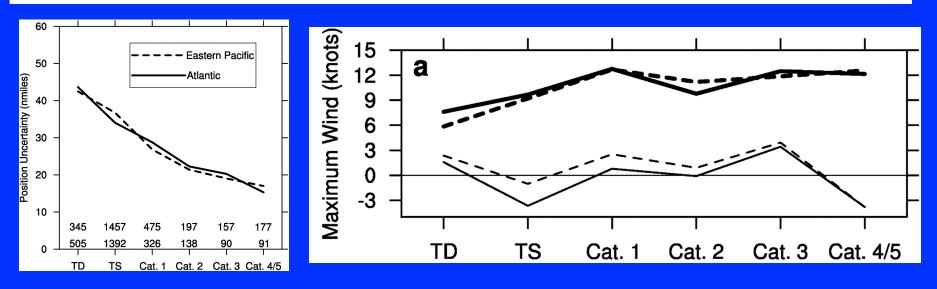
CHANGES IN WATCHES AND WARNINGS WITH THIS ADVISORY ...

THE GOVERNMENT OF THE BAHAMAS HAS DISCONTINUED THE TROPICAL STORM WARNING FOR THE CENTRAL BAHAMAS...AND REPLACED THE HURRICANE WARNING WITH A TROPICAL STORM WARNING FOR THE NORTHWEST BAHAMAS EXCEPT FOR GREAT ABACO AND GRAND BAHAMA ISLANDS.

THE TROPICAL STORM WATCH FOR THE FLORIDA KEYS SOUTH OF OCEAN REEF TO CRAIG KEY AND FOR FLORIDA BAY HAS BEEN DISCONTINUED.

HURRICANE CENTER LOCATED NEAR 26.7N 76.9W AT 26/1500Z POSITION ACCURATE WITHIN 25 NM

PRESENT MOVEMENT TOWARD THE NORTH OR 360 DEGREES AT 5 KT



Torn and Snyder (2012), WAF

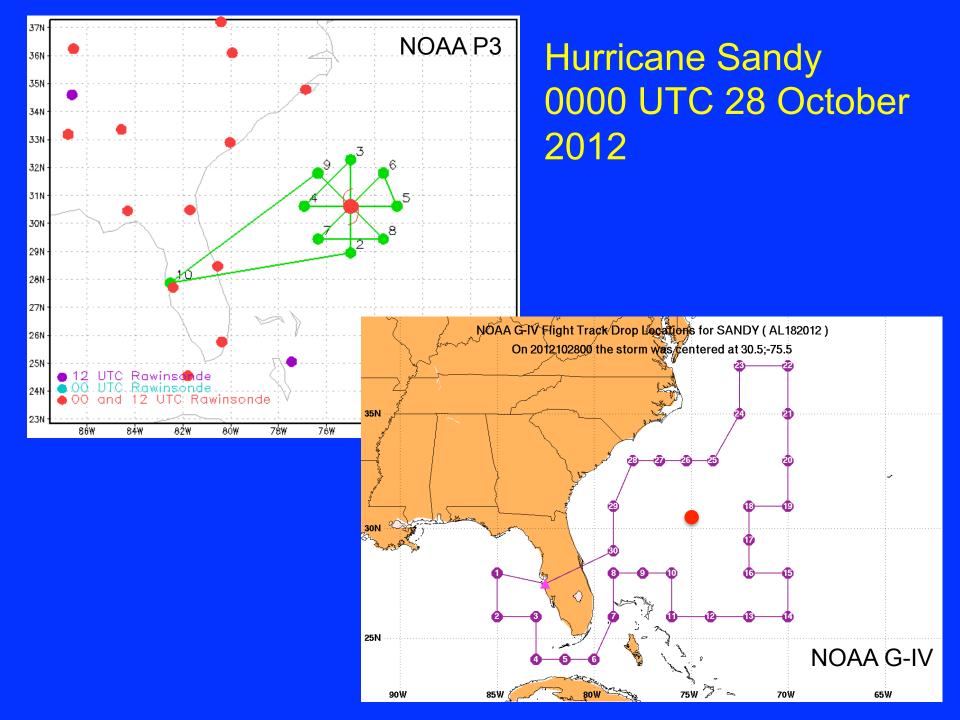
Aircraft Platforms





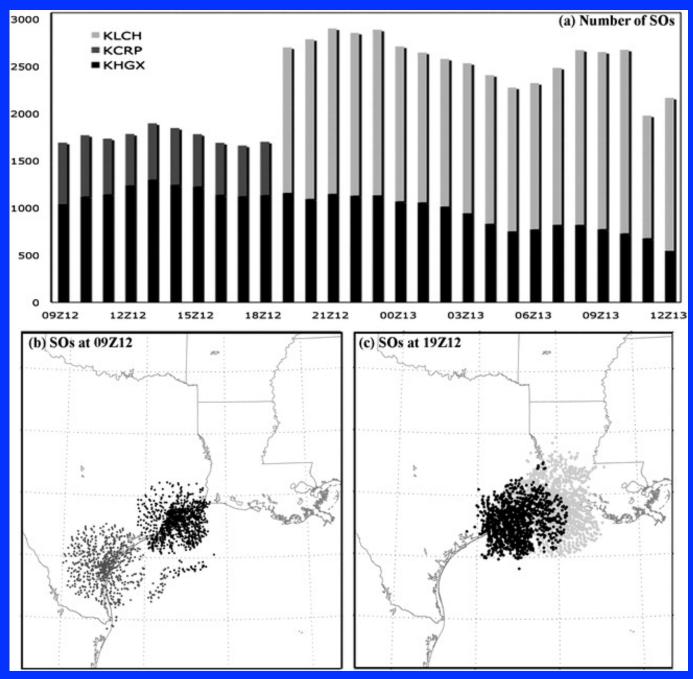
- Flight-level wind, temperature, moisture
- Dropsondes
- Tail-mounted Doppler Radar
- Stepped-Frequency Microwave Radiometer (SFMR; P3)

- Flight-level wind, temperature, moisture
- Dropsondes
- Stepped-Frequency Microwave Radiometer (SFMR)

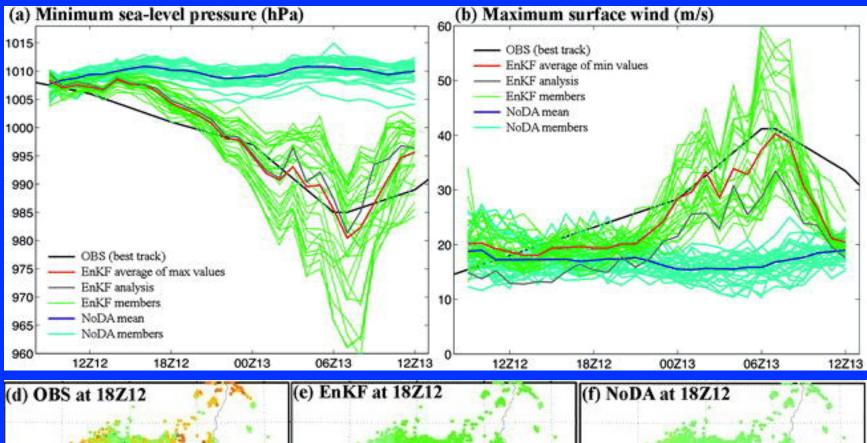


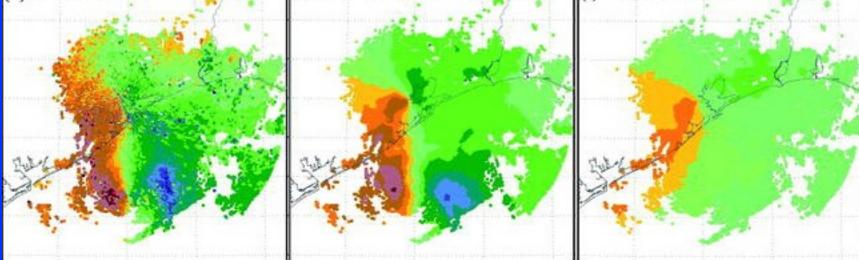
Radar Assimilation

- Dropsonde and flight-level data assimilation is fairly straightforward to assimilate and is more complete
- Radar velocity assimilation considered potentially more valuable due to greater spatial coverage

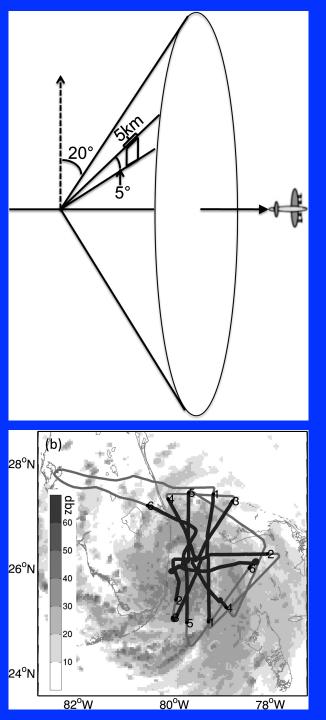


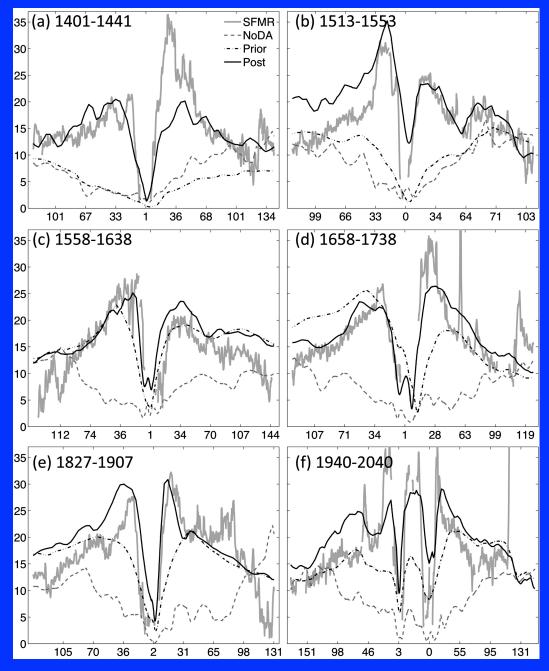
Zhang et al. (2009) MWR





Zhang et al. (2009) MWR





Wang and Zhang (2012), MWR

Comprehensive Recon. Tests

F	orecast Hour	0	6	12	18	24	30	36	42	48	54	60	66	72	78	84	90	96	102	108	114	120
	HWCT-HWAR	-0.7	0	-3	-2.3	-1.8	-1.2	-1.3	-1.2	-3	-3.5	-4.3	-5.5	-5.8	-6.4	-6	-2.5	4	2.8	3	5.3	4.7
	Track	-8%	0%	-13%	-8%	-5%	-3%	-3%	-2%	-5%	-6%	-6%	-7%	-7%	-6%	-6%	-2%	3%	2%	2%	3%	2%
i,	Land/Water	0.803	0.018	0.99	0.899	0.693	0.465	0.465	0.445	0.817	0.8	0.819	0.88	0.919	0.892	0.768	0.36	0.504	0.318	0.3	0.429	0.321
ă	HWCT-HWAR	-5.3	-7.6	-5	-2.6	0.4	1.2	1.8	1.4	1.1	0.7	0.8	1	-0.1	1.1	0.2	1.1	-0.4	0.3	0	0.1	-0.1
tic	Intensity	-151%	-104%	-57%	-25%	3%	10%	16%	12%	9%	5%	6%	8%	-1%	8%	2%	8%	-3%	2%	0%	1%	-1%
an	Land/Water	0.999	0.999	0.992	0.877	0.177	0.654	0.889	0.94	0.68	0.656	0.648	0.77	0.132	0.886	0.158	0.747	0.39	0.234	0.001	0.116	0.105
At	HWCT-HWAR	-5.3	-8.2	-5.7	-3	0.8	1.6	2.1	1.8	1.3	0.4	1.3	0.5	0.2	-0.4	-1	-0.5	2	2.2	1.6	3.8	0
	Intensity	-153%	-115%	-65%	-29%	7%	13%	17%	14%	10%	3%	9%	4%	2%	-3%	-7%	-3%	12%	13%	10%	21%	0%
	Water Only	0.999	0.999	0.989	0.83	0.27	0.585	0.812	0.728	0.479	0.177	0.705	0.214	0.153	0.264	0.512	0.256	0.816				

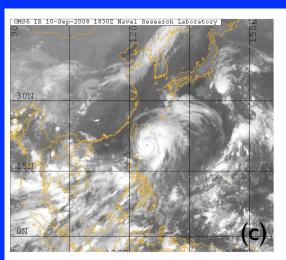
F	orecast Hour	0	6	12	18	24	30	36	42	48	54	60	66	72	78	84	90	96	102	108	114	120
	APCT-APAR	8.3	5.7	5.2	7.9	7.7	8.4	8.4	8.6	13.3	15.7	21.1	24.9	30.8	29.5	14.1	17.8	8.8	9.9	13	21.3	32.8
	Track	33%	19%	15%	20%	17%	16%	15%	13%	17%	18%	21%	22%	24%	22%	12%	14%	8%	8%	9%	12%	16%
U.	Land/Water	0.999	0.985	0.98	0.999	0.998	0.982	0.921	0.82	0.863	0.917	0.893	0.758	0.482	0.548	0.923	0.867	0.79	0.773	0.805	0.934	0.967
ä	APCT-APAR	-2.4	-1.2	-0.5	0.3	0.7	1.6	1.5	0.7	1.8	1.7	1.5	1.4	0	0.2	0.6	-0.1	-1.1	0	-1	-0.8	-1
iti	Intensity	-23%	-12%	-5%	3%	6%	14%	12%	5%	14%	14%	12%	11%	0%	2%	5%	-1%	-10%	0%	-9%	-7%	-8%
a	Land/Water	0.982	0.631	0.331	0.183	0.414	0.822	0.752	0.378	0.859	0.878	0.967	0.889	0.008	0.164	0.555	0.059	0.762	0.02	0.713	0.436	0.342
At	APCT-APAR	-2.4	-1.5	-0.7	0.4	1	2.5	3.2	1.2	2.8	2	1.1	0.7	-0.9	-0.7	1	0	-0.6	-2.1	-1.2	0.8	-5.2
	Intensity	-23%	-15%	-6%	3%	8%	19%	23%	9%	21%	15%	9%	6%	-7%	-5%	7%	0%	-4%	-22%	-16%	9%	-62%
	Water Only	0.982	0.7	0.356	0.171	0.419	0.856	0.901	0.498	0.788	0.912	0.731	0.314	0.561	0.386	0.642	0.001					1 1

F	orecast Hour	6	12	18	24	30	36	42	48	54	60	66	72	78	84	90	96	102	108	114	120
	HECT-HEAR	12.1	9.1	9.4	7.9	9.6	4.6	0.1	0.7	5.5	6.1	3.6	4	6.5	1.6	4.3	9.9	12.7	7.3	1.1	4
	Track	37%	21%	19%	14%	15%	7%	0%	1%	6%	6%	3%	3%	4%	1%	2%	5%	6%	3%	0%	2%
Si-	Land/Water	0.991	0.99	0.996	0.968	0.98	0.73	0.018	0.173	0.89	0.89	0.635	0.664	0.807	0.234	0.558	0.881	0.923	0.735	0.119	
Ba	HECT-HEAR	7	4.1	4.7	2.9	2.3	1.4	0.2	1.8	-0.4	0.4	-0.9	-0.3	-0.1	-1.7	-4.4	-2.6	-1.6	-0.1	1.8	-4.2
ţi	Intensity	38%	24%	30%	20%	16%	9%	1%	12%	-3%	3%	-6%	-2%	-1%	-12%	-30%	-14%	-8%	0%	8%	-17%
a	Land/Water	0.999	0.999	0.996	0.992	0.972	0.667	0.131	0.871	0.244	0.252	0.531	0.166	0.071	0.49	0.8	0.707	0.431	0.035	0.356	
At	HECT-HEAR	7.1	4.6	5.4	3.2	2.6	1.6	-0.2	2.2	-0.8	0.9	-1.8	-2.4	-2.8	-5.9	-9.8	-8.8	-5.4	-3.9	-1.2	-2.6
	Intensity	38%	27%	32%	21%	17%	10%	-1%	13%	-4%	4%	-10%	-14%	-17%	-38%	-63%	-55%	-29%	-20%	-5%	-10%
	Water Only	0.999	0.999	0.996	0.987	0.953	0.466	0.1	0.776	0.253	0.289	0.546	0.65	0.796	0.932						

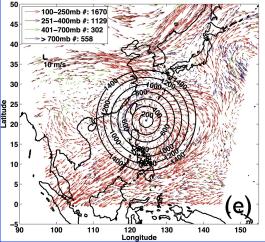
- 101 cases over 5 different seasons
- 3 Different groups, two different models
 - NCEP Environmental Modeling Center (HW*)
 - Penn State University (AP*)
 - NOAA Hurricane Research Division (HE*)

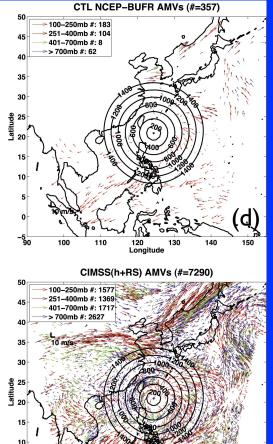
Nance et al. (2013) Report

Atmospheric Motion Vectors



CIMSS(h) AMVs (#=3659)



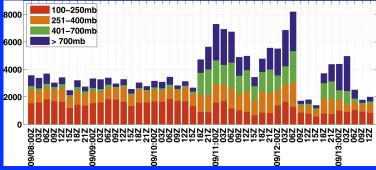


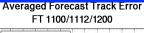
100

110

120

Lonaitude



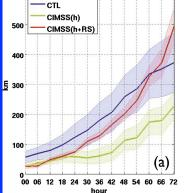


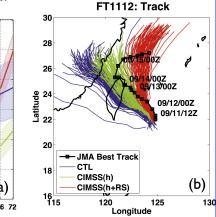
600

(†)

150

140



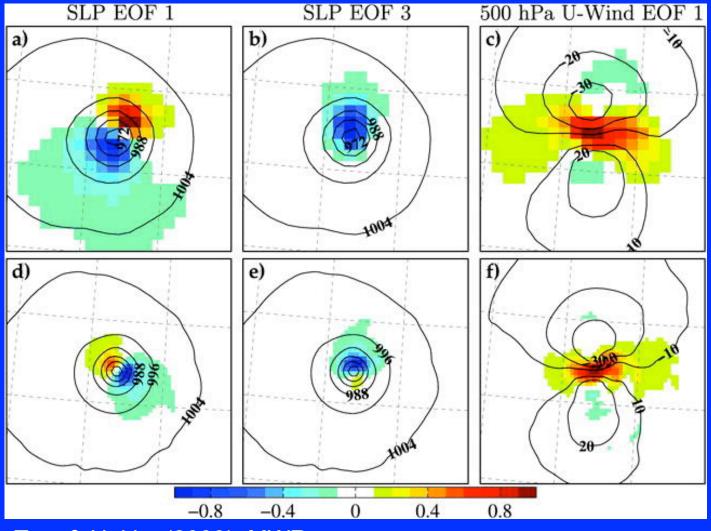


Wu et al. (2014), MWR

Challenges

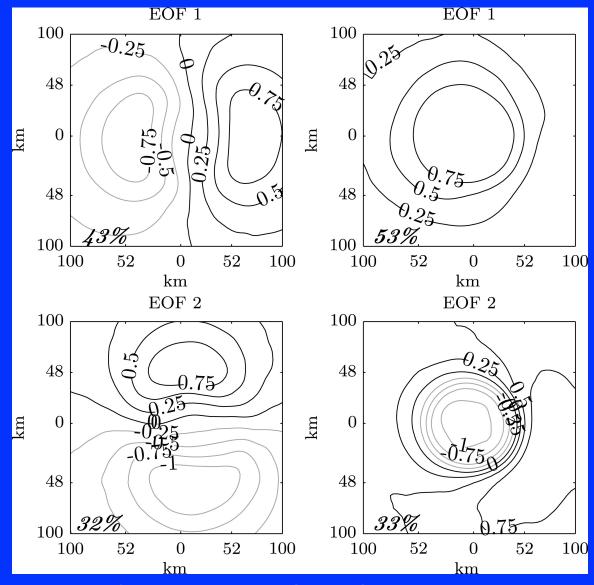
- TC data assimilation has a number of challenges, some of which are unique to the application
 - Modes of variability
 - Impact of sampling errors
 - Representativeness
 - Variable observation densities
 - Model biases
 - Domain sizes

Tropical Cyclone Variability



Torn & Hakim (2009), MWR

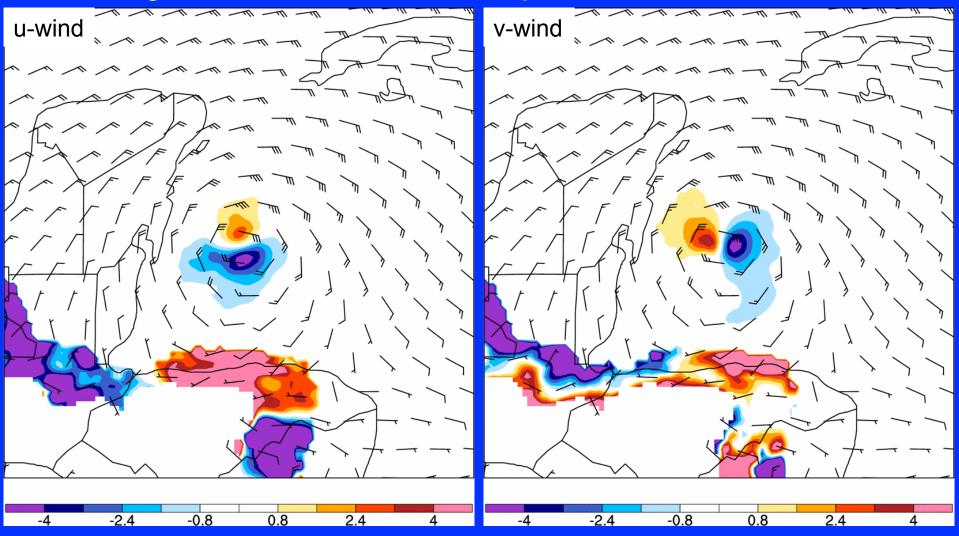
Tropical Cyclone Variability



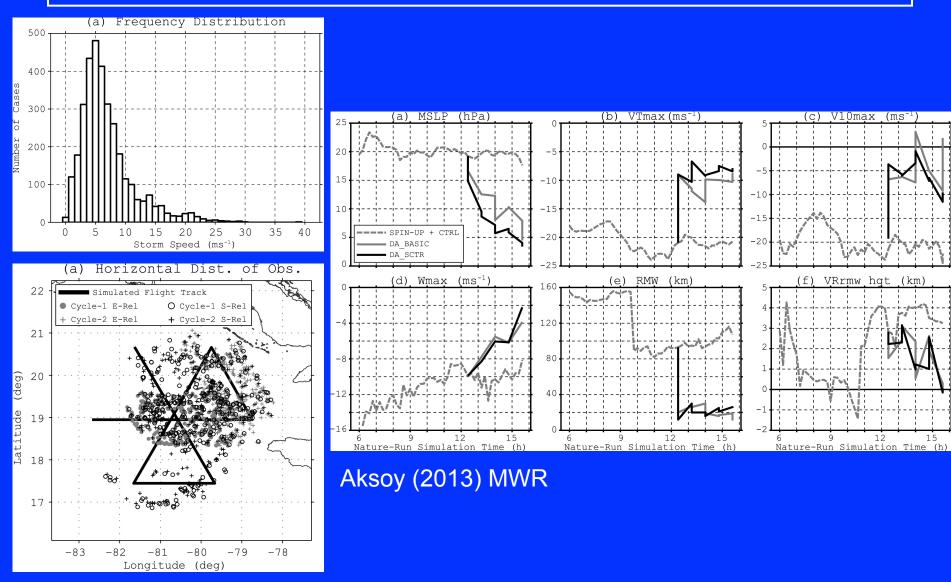
Navarro & Hakim (2013) QJRMS

Storm-centered Assimilation

Regression Between TC sea-level pressure and 950 hPa wind



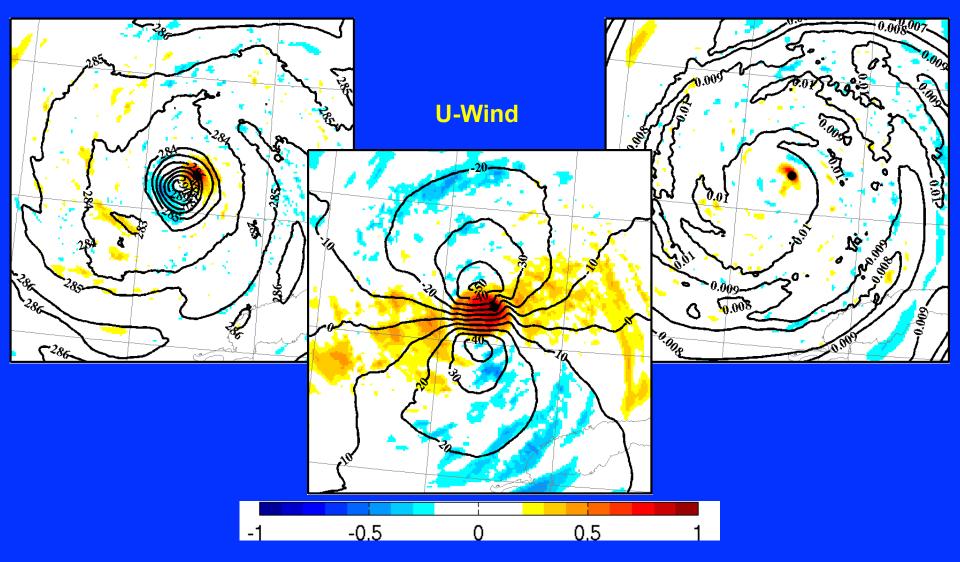
Storm-relative Obs.

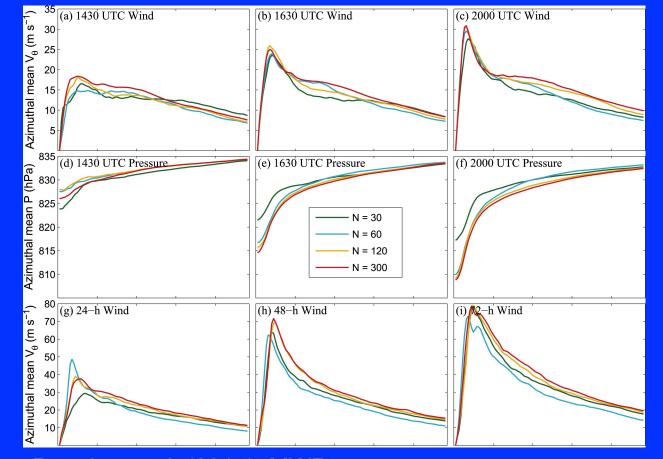


Covariances

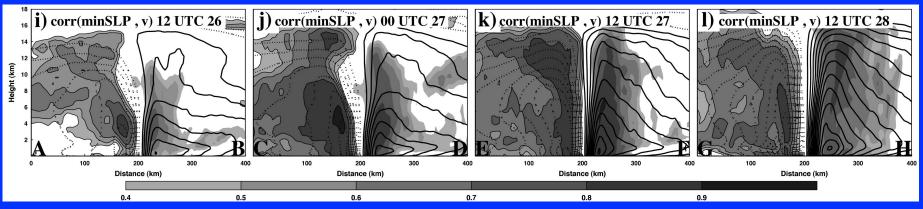
Temperature

Mixing Ratio



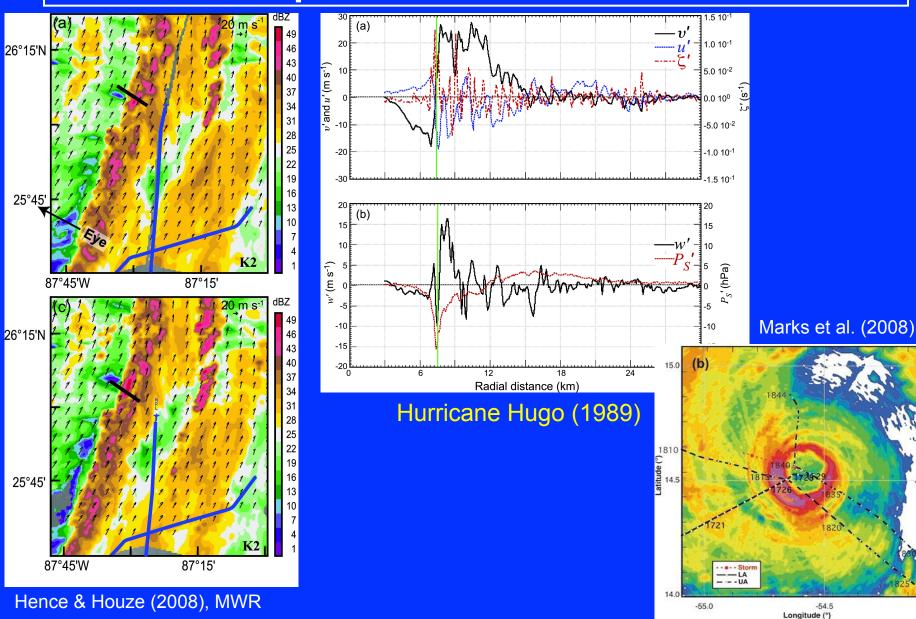


Poterjoy et al. (2014), MWR

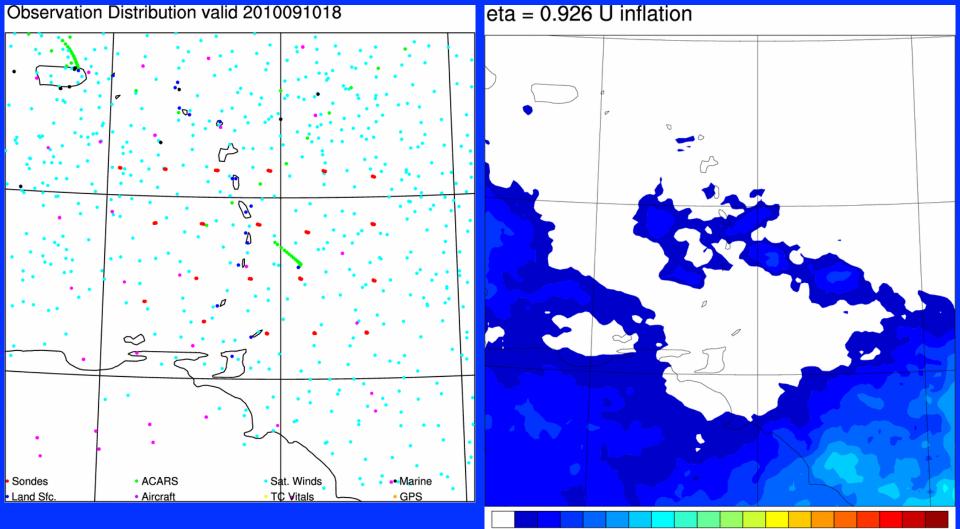


Poterjoy and Zhang (2011), MWR

Representativeness



Observation Variability



1.3

1.5

1.7 1.9

2.3

2.1

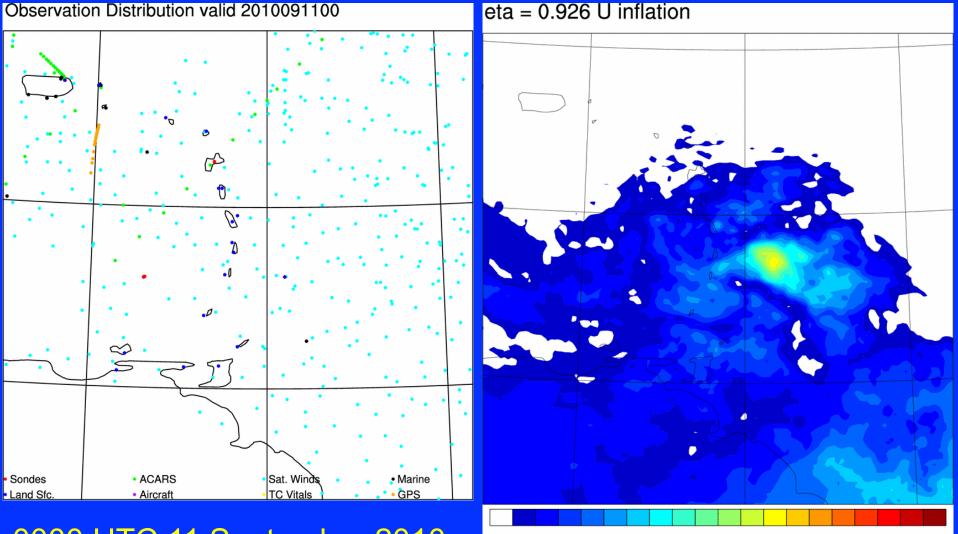
2.7

2.9

2.5

1800 UTC 10 September 2010

Observation Variability



1.5

1.3

1.9

2.1

2.3

2.5

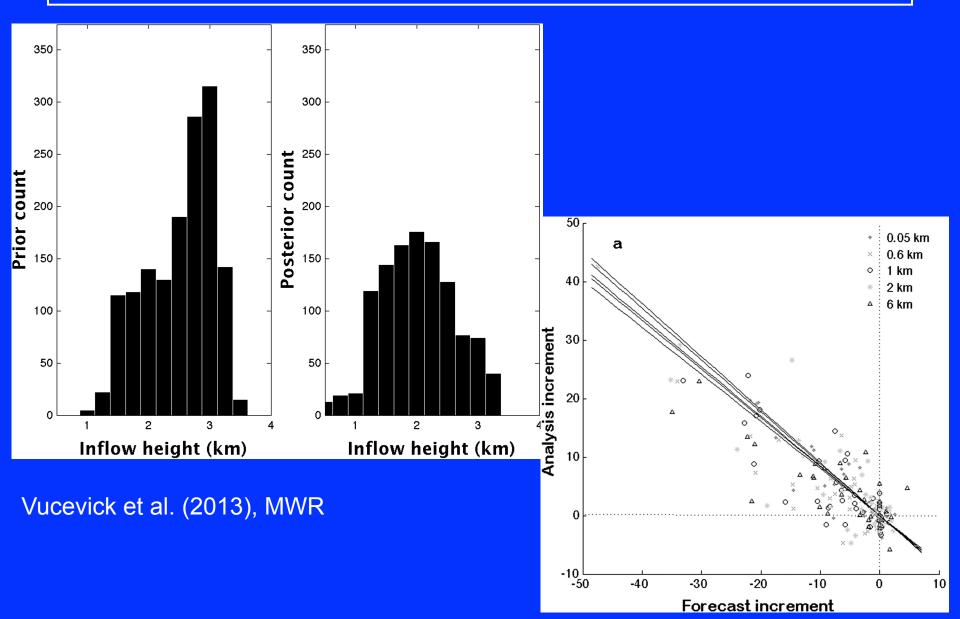
2.7

2.9

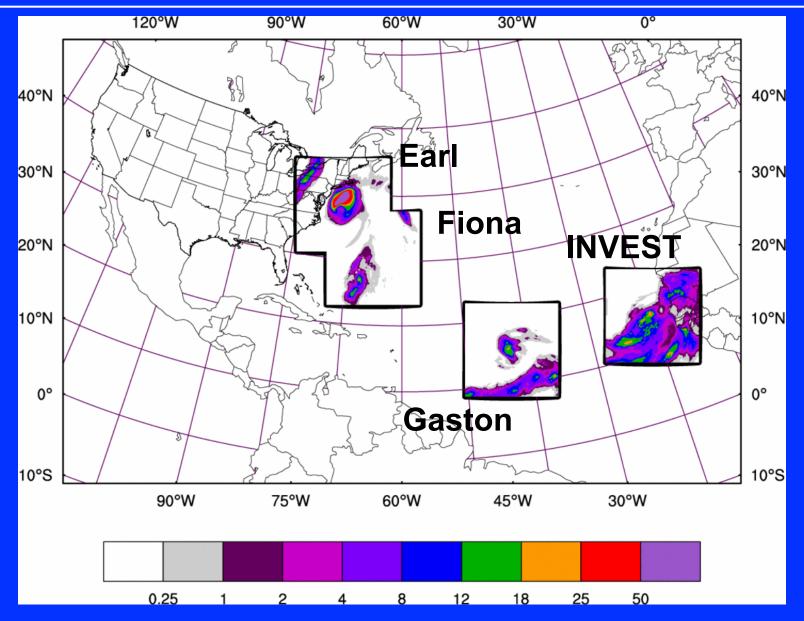
1.7

0000 UTC 11 September 2010

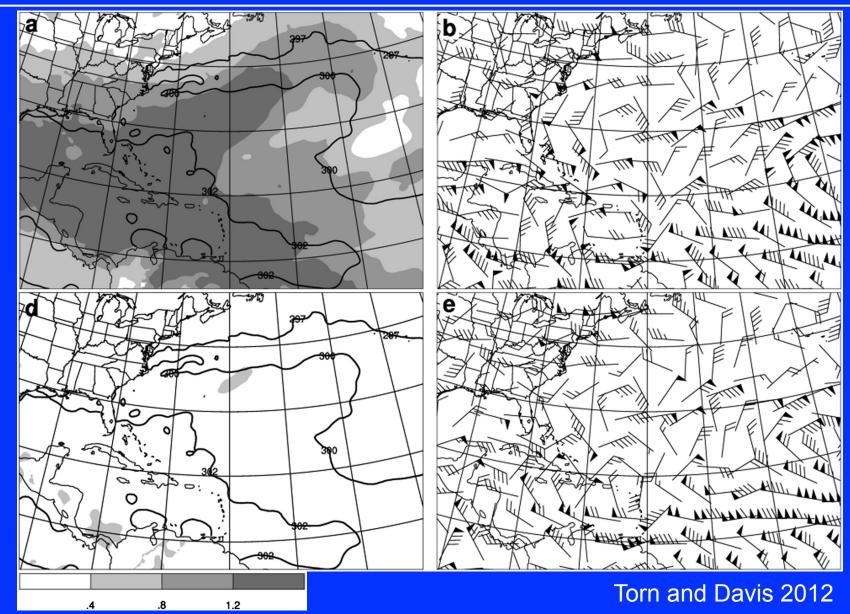
Model Biases



Interaction & Basin-Scale Models



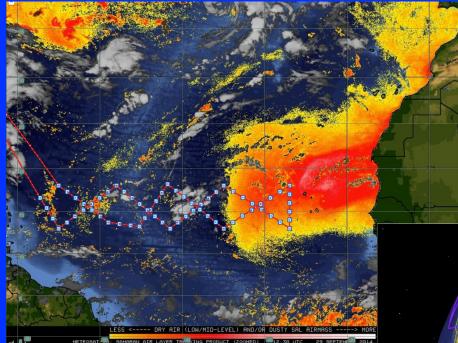
Prior Biases



New Directions

- Most of the above issues can be addressed with more development & computing
- New observation platforms are becoming available that will provide better coverage
- New methods of determining where to direct scarce observation resources

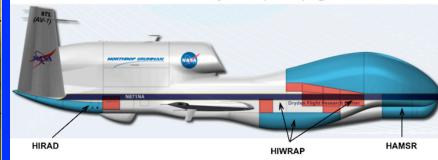
Unmanned Vehicles

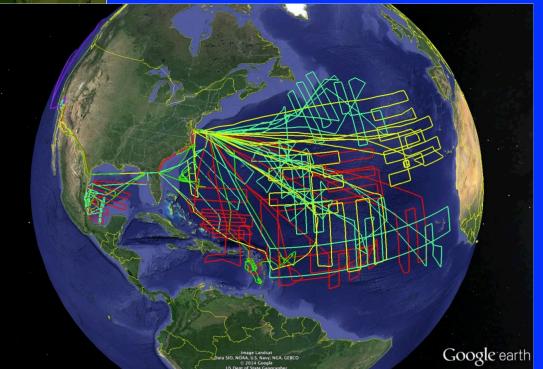


NASA Global Hawk

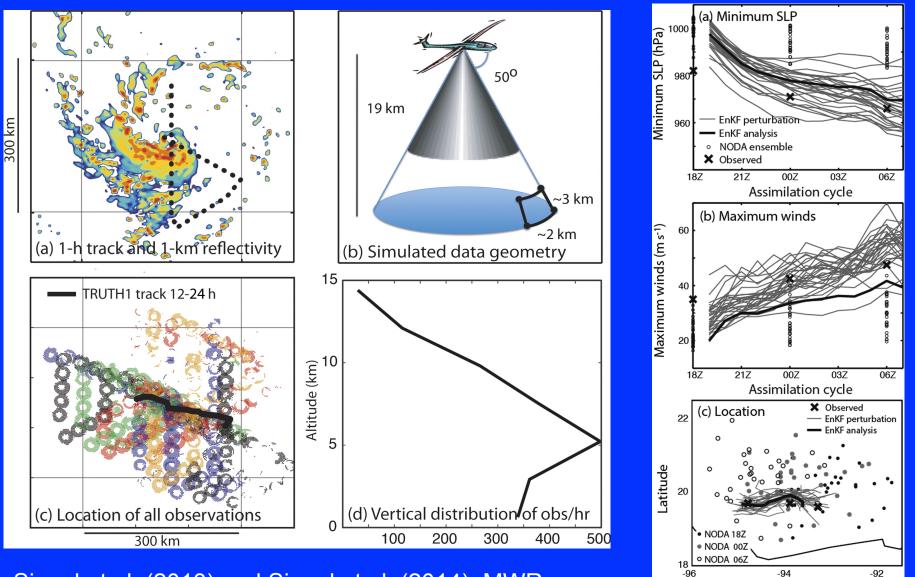
- Has similar observation capability to manned aircraft (including radar)
- Long duration flights (> 20 h)
- Used in Hurricane Sentinel (HS3) project

HS3 Over-storm Payload (AV-1) @ WFF '14





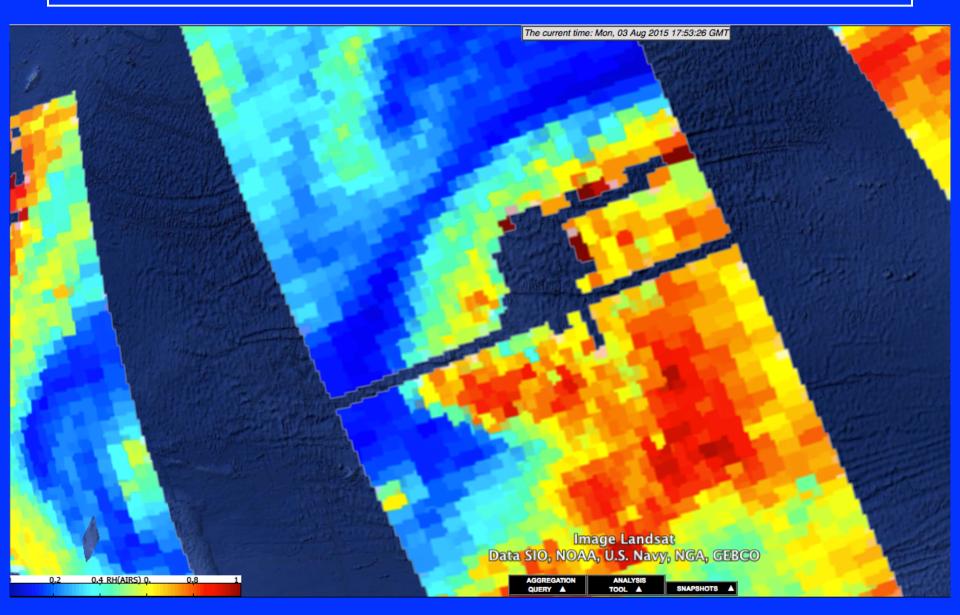
HIWRAP Assimilation



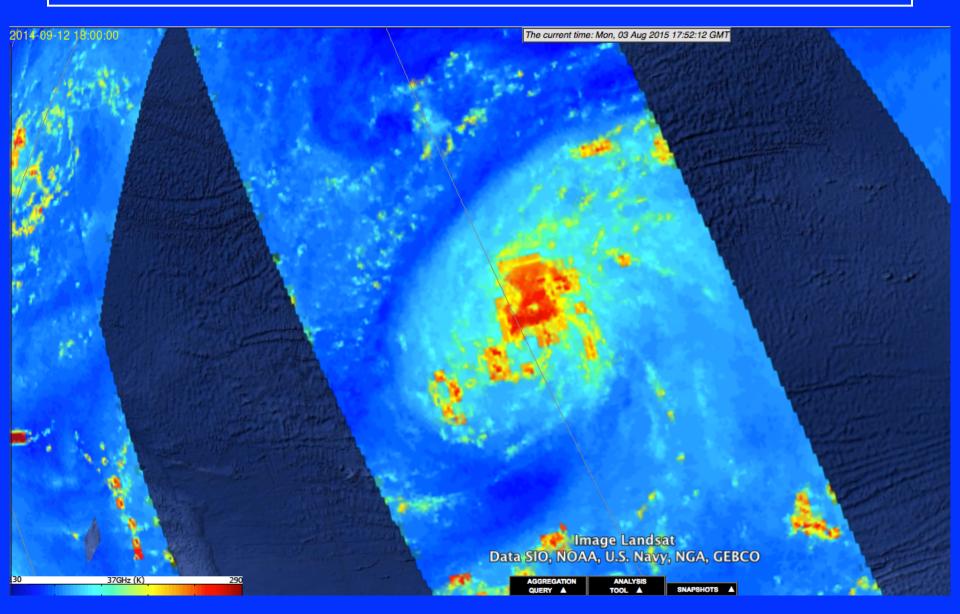
Longitude

Sippel et al. (2013) and Sippel et al. (2014), MWR

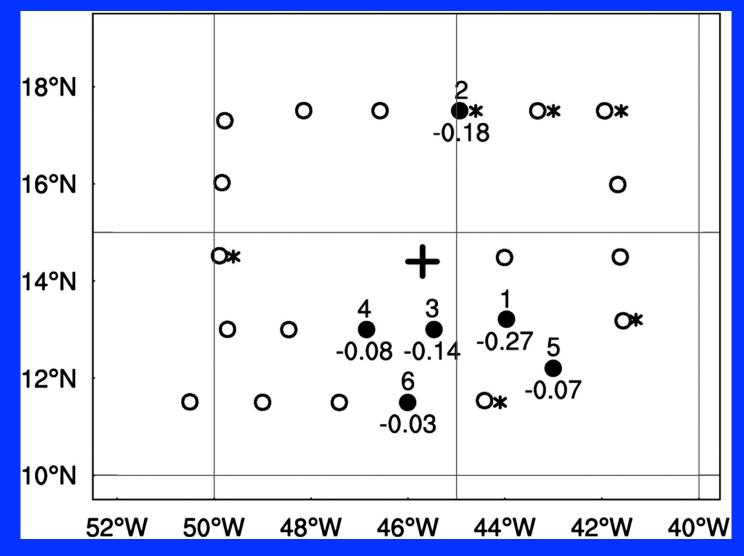
Satellite Data



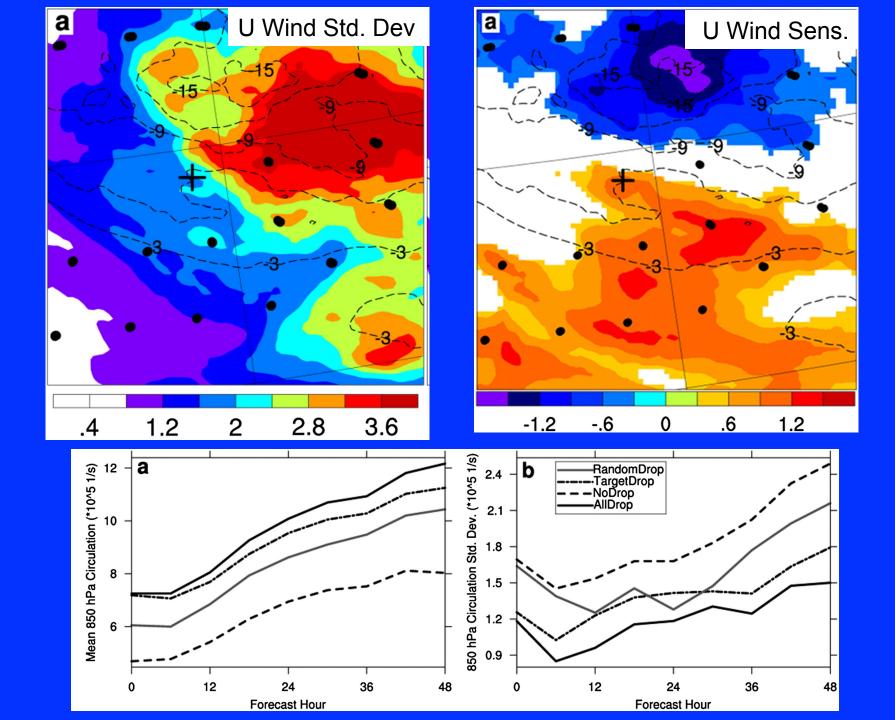
Satellite Data



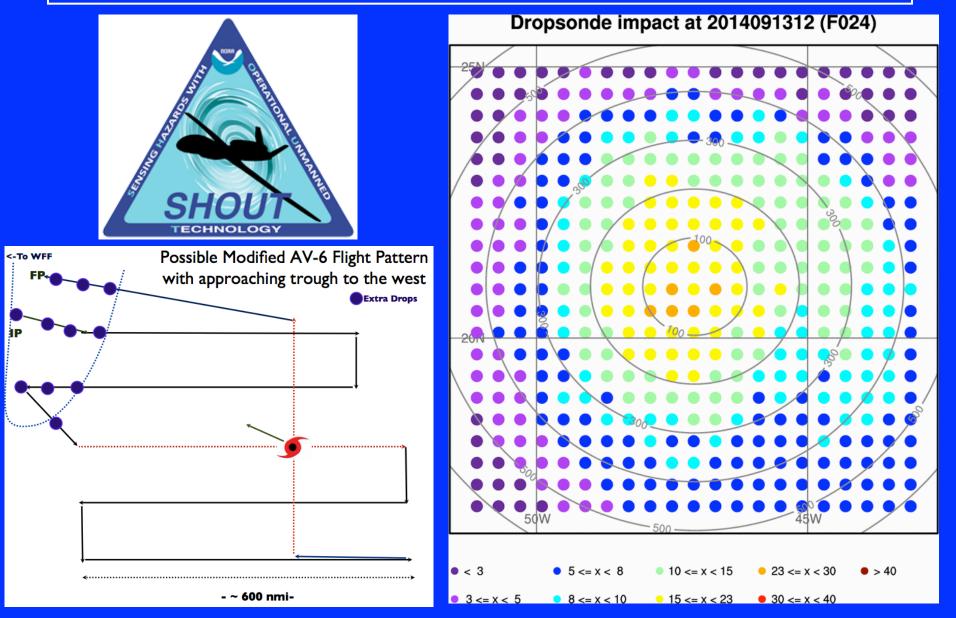




Torn (2014) MWR



SHOUT



Summary

- Greatest challenges in TC data assimilation due to multiscale nature
 - Needs more intelligent methods of dealing with sampling errors
 - Must remove position variability
- Aircraft data is useful for assimilation, but impact has been limited. Likely that UAVs will provide better coverage in next 10-20 years
- Satellites hold promise to provide frequent coverage, but forward operators are problem
- Biggest improvement to assimilation likely to come from model itself. DA systems are too stressed by bias and representativeness problems
- Need to focus forecast metrics away from wind
 - Storm surge & Precipitation