Quantifying Sensor Web Capabilities Through Simulation: Recent Results

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Project Goals

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Demonstrate the value of implementing sensor web concepts for meteorological use cases

Quantify cost savings to missions **Quantify** improvement in achieving science goals

Design and Build an integrated simulator with functional elements that will allow multiple "what if" scenarios in which different configurations of sensors, communication networks, numerical models, data analysis systems, and targeting techniques may be tested





Improvements in predictive skill over the past several decades have been gradual; the sensor web provides an opportunity for a "revolutionary" impact

Source: Fanglin Yang, Environmental Modeling Center, National Centers for Environmental Prediction, NOAA

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Example: Societal Impact and Predictive Skill

4.000

5

Errors in temperature forecasts lead to errors in the prediction of electrical loads for large utilities



Source: Mary G. Altalo (SAIC, Inc) and Leonard A. Smith (London School of Economics): "Using Ensemble Weather Forecasts to Manage Utilities Risk", *Environmental Finance*, October 2004

Example: Societal Impact and Predictive Skill

Peak load temperatures

2.000

6

4.000

Errors in temperature forecasts lead to errors in the prediction of electrical loads for large utilities



Source: Mary G. Altalo (SAIC, Inc) and Leonard A. Smith (London School of Economics): "Using Ensemble Weather Forecasts to Manage Utilities Risk", Environmental Finance, October 2004

Use Case: Decadal Survey Mission 3D Wind Lidar



Application of Sensor Web Concepts

Simulation 1: Extend Mission Life via Power Modulation

- Conserve power / extend instrument life by using aft shots only when there is "significant" disagreement between model first guess line-of-sight winds and winds measured by fore shots
 - Lidar engineers have recently suggested reduced duty cycles may increase laser lifetimes
 - Duty cycles that are on the order of 10 minutes "on" and 80 minutes "off" may be very beneficial to mission lifetime
- Will require model's first guess fields be made available on board the spacecraft -- requires engineering trades be performed for on-board processing, storage, power, weight, communications



Main Architecture ("Science Layer")



Simulation 1 Results



Lidar data deleted when there is "adequate" agreement with the numerical model's first guess wind fields

Designed to simulate suppression of the aft shot of the lidar

Result: Nearly 30% of the lidar's duty cycle may be reduced -- IF there is no discernible impact to forecast skill!



Simulation 1 Results



Application of Sensor Web Concepts

Simulation 2: <u>Better Science</u> via Targeted Observations

- Goal is to target two types of features to help improve predictive skill:
 - "Sensitive regions" of the atmosphere: those regions where the forecast is highly responsive to analysis errors
 - Features of interest that may lie outside of the instrument's nadir view
 - Tropical cyclones
 - Jet streaks
 - Rapidly changing atmospheric conditions
- Would require slewing
- Would require optimization to choose between multiple targets
- Studies have shown that targeted observations can improve predictive skill (difficult to implement operationally)



Source: D. Emmitt and Z. Toth, 2001: Adaptive targeting of wind observations: The climate research and weather forecasting perspectives. Preprints, 5th Symposium on Integrated Observing Systems, AMS.

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Calculating "Sensitive Regions"

Differences between two forecasts launched 72 hours apart and valid at the same forecast hour. Largest differences ("sensitive regions") depicted in colored shading.



Studies have shown the adjoint technique to be effective for adaptive targeting[†]. Testing with this technique will occur during years 2-3 in coordination with NASA's Global Modeling & Assimilation Office.

[†]Leutbecher, M., and A. Doerenbecher, 2003: Towards consistent approaches of observation targeting and data assimilation using adjoint techniques. Geophysical Research Abstracts, Vol. 5, 06185, European Geophysical Society.

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Selecting Specific Targets



Simulation 2: Adaptive Targeting

CHOS_

• Click here to play movie



NASA Experiment Design

NED Features

- GUI allows simple configuration of experiment workflow variables
- "Design-mode" to create templates for new experiment configurations
- Save configured experiments to files or a searchable database
 - Re-run past experiments
 - Share with other users
- Compare experiment
 configurations for changes
- Version-controlled scripts



Offer development branches Automatic workflow updates



Simulator Example

X NED 1.0.1.11 User: rburns Mode: USER

File Search Tools Database Submission

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Configuration Tree Configuration 🖥 Sensor Web Simulator . Group Property Information Name Lidar 🖕 🗂 Experiment Description default description ∲
─ □ Observing System Output Type BASH Shell Transform 🔶 🥅 Lidar Output File ExperimentVars.bash lidarType lidarGeometr dutvCvcle. telecopeDiam lidarSlewing aftFireThresh lidarDataPath 🗂 Simpson Weather Name Description Data Type Values lidarType Type of lidar STRING LIDAR_COHERENT quikscat Method of observation STRING lidarGeometry LIDAR_SCANNING SSM_I_Speeds gathering AMSR_E_Speeds How often observations dutyCycle STRING are taken (seconds) TMI_Speeds Size of telescope telecopeDiameter STRING Data Assimilation Svi (millimeters) 🔶 🚍 Forecast Model lidarSlewing Turn on slewing ability INTEGER V Targeting System Percent of failed 30 aftFireThreshold observations needed to INTEGER 🔶 🗂 Feature Detection trigger aft response advection Source of lidar data \$archivePath/\$NED_USE lidarDataPath STRING cvclone tropicalCyclor Status deepening Status Affects Description jet isUsingFeatur 👻 Ш •

User-Directed Targeting



Options for Uplink Comms Requirements

Class	Architecture	Forward Link Capacity
Non-TDRS	S-band; AK ground station	1 Mb/s
	S-band; AK, SVB, MGS ground stations	1 Mb/s
	SafetyNet S-band Augmentation	1 Mb/s
	SafetyNet Ka-band Augmentation	25 Mb/s
TDRS	SSA	300 Kb/s, 1Mb/s
	SSA, KaSA	7 Mb/s
	KaSA	25 Mb/s
Hybrid	RF uplinks, Optical Cross Links	100-200 Mb/s



Non-TDRS "Classical" Ground Stations



TDRS Forward Link Architectures



2030 Architecture



Acronym List

ABI	Advanced Baseline Imager	
DAS	Data Assimilation System	
ECS	External Control System	
EPOS	Earth Phenomena Observing System	
FTE	Full Time Equivalent	
GEOS5	Fifth-Generation Goddard Earth Observing System (numerical model)	
GOES-R	Geostationary Operational Environmental Satellite "R" Series	
GMAO	Global Modeling & Assimilation Office	
GSI	Gridpoint Statistical Interpolation	
GWOS	Global Wind Observing System	
NHC	National Hurricane Center	
NOAA	National Oceanic & Atmospheric Administration	
OSSE	Observing System Simulation Experiment	
SAIC	Science Applications International Corp	
TEC	Targeting & External Control	
UTC	Coordinated Universal Time	
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Approach

Sensor Web Simulator Design

- During 2007 most elements of the lidar use case (1-5) were executed "by hand" to help aid in the design of the simulator prototype
- Five separate Observing System Simulation Experiments (OSSEs) were conducted that concluded:
 - Under certain situations¹, the lidar duty cycle may be reduced 30% without impacting forecast skill
 - Under certain situations, having the model task the lidar to perform a roll maneuver improves detection of features of interest 30% (tropical cyclones, jet streaks, rapidly changing atmospheric conditions)

SIVO Workflow Tool ("NASA Experiment Design")

 Selected as the "glueware" to sequentially execute components 1-6 and manage data flow



1 The OSSEs performed were based upon a 20 day assimilation cycle during September 1999. Although the use cases have been examined by GMAO scientists they have not undergone a rigorous scientific review and the results should not be considered scientifically valid. OSSEs presented here are to validate engineering processes of the simulator.