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Towards Multi-variate Land Data Assimilation in the NASA GEOS-5 System

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With contributions from:

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- GEOS-5 land data assimilation system
- Multi-source observations and multi-variate analysis

Multi-source observations

- Geostationary skin temperature retrievals
- Active and passive soil moisture retrievals
- Precipitation and soil moisture retrievals

(LST analysis)

(soil moisture analysis)

(soil moisture analysis)

• Snow water equivalent and snow cover fraction retrievals (snow analysis)

Multi-variate analysis

- Soil moisture and snow analysis
- Soil moisture and soil temperature analysis



The NASA Goddard Earth Observing System, Version 5 (GEOS-5) is a global Earth system modeling and data assimilation framework.

GEOS-5 is being developed to support NASA's Earth science research, including data analysis, observing system modeling and design, climate and weather prediction, and basic research.

GEOS-5 is used to generate the MERRA reanalysis.

GEOS-5 includes a land data assimilation system component, typically run in "off-line" (land-only) mode.



Introduction – GEOS-5 Land Data Assimilation System

Ensemble Kalman filter: Iocal ("1d") or distributed ("3d"), where 3d includes spatial extrapolation, interpolation, and disaggregation of assimilated observations GEOS-5 Catchment land

surface model

Uncertainty estimation is at the heart of the ensemble-based approach.









Session title: "Utilization of Multi-source Observations"

Sensors and assimilated observations: Which types of observations are assimilated? Where do they come from?

Presentation title: "Towards Multi-variate Land Assimilation"

Geophysical (control) variables: Which model variables are updated in the analysis?



Update equation at the heart of the land assimilation system:





Early land data assimilation projects typically used a **single observation type** in a **uni-variate** analysis, e.g.:

- AMSR-E soil moisture retrievals \rightarrow soil moisture analysis
- AMSR-E snow water equivalent retrievals → snow analysis
- MODIS skin temperature retrievals \rightarrow land surface temperature analysis

Towards multi-source observations in a multi-variate analysis:
1.a.) Same type of observations, but from different sensors
1.b.) Different types of observations
2.a.) Single type of observation, multi-variate analysis
2.b.) Different types of observations, multi-variate analysis



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Geostationary Tskin observations

Clear-sky Surface Skin Temperature [K] – 15 May 2012, 12z



Near-real-time, 3-hourly composites of Tskin retrievals from 5 geostationary imagers. Provided by NASA Langley Research Center (LaRC).



Geostationary Tskin observations





Geostationary Tskin Observations

Draper et al. (2014) developed a dynamic *observation* bias estimate based on observations minus forecast (O-F) residuals.

Estimate of Tskin observation bias [K], 1 June 2012



Draper et al., 2014, *JHM*, accepted.



Geostationary Tskin Observations

Diurnal cycle in O-F mean difference





Geostationary Tskin Observations



1:30pm LST



ubRMSD open loop minus ubRMSD with assimilation



The bias-aware filter yields improved assimilation estimates (when compared to the open loop skill).

 Skill is vs.
 independent satellite observations
 (MODIS).
 Similar results vs.
 in situ obs
 (SURFRAD).

> Draper et al., 2014 *JHM*, accepted.



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Satellite Remote Sensing of (Surface) Soil Moisture



2002-2011 C/X-band passive 40 km resolution



2007-present C-band active 40 km resolution



2009-present L-band passive 40 km resolution interferometric & multi-angular



Launch: 2014 (?) L-band active/passive 3-40 km resolution

Frequency band	Sensing depth		
C/X-band	1 cm		
L-band	5 cm		



Assimilation of Soil Moisture Retrievals

Skill increases significantly a) through data assimilation. 0.6 Similar improvements from AMSR-E and ASCAT. R (surface) **Root-zone** *not* observed by 0.5 satellite. **Improvements** may be critical for applications. Metric: Anom. time series corr. coeff. (R) Anomalies ≡ mean seasonal cycle removed Validated with in situ data b) a) -120 -100'-80 50' 0.6 R (root-zone) 40* 0.5 30" b) 0.4144 148 152 156 FOR -32" MIX GRA CRO -36 BAR URB



Draper et al. (2012), GRL, doi:10.1029/2011GL050655.

Soil Moisture Assimilation and Precipitation Corrections



Different precipitation forcing inputs

Soil Moisture Assimilation and Precipitation Corrections



Precipitation corrections and retrieval assimilation contribute approximately:

- evenly and
- independently to skill improvement.

Results from single sensor per watershed (SCAN data) are consistent with those from distributed CalVal in situ sensors.

Liu et al. JHM (2011) doi:10.1175/JHM-D-10-05000.



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Assimilation of MODIS and AMSR-E Snow Observations

Noah land surface model (1 km resolution) 100 km X 75 km domain in northern Colorado





Multi-scale assimilation of

- AMSR-E snow water equivalent (SWE) and
- MODIS snow cover fraction (SCF).

Validation against in situ obs from COOP (Δ) and Snotel (•) sites for 2002-2010.

Assimilation of MODIS Snow Cover Fraction (SCF)



MODIS SCF also improves timing of onset of snow season (not shown).

De Lannoy et al. (2012) doi:10.1029/2011WR010588.

Solution of AMSR-E Snow Water Equivalent (SWE)



0

- AMSR-E SWE retrievals and MERRA differ **a lot**.
- Working on observation operator based on machine learning.

Assimilation of AMSR-E Snow Water Equivalent (SWE)



Forman and Reichle (2014) doi:10.1109/JSTARS.2014.2325780.



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Assimilation of GRACE Terrestrial Water Storage (TWS)



0.2

0.4

0.6

0.8

Forman et al. WRR (2012) doi:10.1029/2011WR011239.



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Satellite Remote Sensing of (Surface) Soil Moisture

Use SMOS data to prepare for the



2009-present L-band passive 40 km resolution interferometric & multi-angular

SMAP Level 4 Surface and Root Zone Soil Moisture (L4_SM) product.



Launch: 2014 (?) L-band active/passive 3-40 km resolution

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SMAP L4_SM Analysis Overview





L4_SM Soil Moisture Analysis



NASA

L4_SM Radiative Transfer Model Parameters

- Radiance assim. requires unbiased L-band radiative transfer model.
- Locally optimized to minimize differences in long-term mean and std-dev between Tbs from SMOS and GEOS-5 forward modeling
- Areas where SMOS data are not suitable for calibration (e.g., due to RFI) are filled with calibrated parameter values that are (spatially) averaged by vegetation class.



De Lannoy et al., RSE, 2014, doi:10.1016/j.rse.2014.03.030

L-band Brightness Temperature: SMOS vs. GEOS-5

Seasonally varying biases remain even after calibration of the microwave radiative transfer model (RTM) parameters.

Derive climatological scaling parameters (based on 4 years of SMOS data).

After calibration of RTM parameters

After climatological Tb scaling





Calibration of the distributed ("3-dimensional") analysis requires **perturbation parameters** for surface meteorological forcing and soil moisture:

- Std-dev
- Spatial correlation scales
- Temporal correlation scales
- Cross-correlations

Assimilated Tb observations from SMOS are multi-angular, but for now observation error cross-correlations are neglected.



Calibration of the Data Assimilation System

Perturbation	Additive (A) or Multipli- cative	Std- dev	AR(1) time series correlation scale	Spatial correlation scale	Cross-correlation with perturbation in		ation tions
	(IVI)				•	011	
Precipitation (P)	M	0.5	24 h	50 km	n/a	-0.8	0.5
Downward shortwave (SW)	М	0.3	24 h	50 km	-0.8	n/a	-0.5
Downward longwave (LW)	А	20 W/m²	24 h	50 km	0.5	-0.5	n/a
					catdef	srfexc	
Catchment deficit (catdef)	A	0.03 kg/m²	3 h	h 50 km n/a		0.0	
Surface excess (srfexc)	A	0.02 kg/m ²	3 h	50 km	0.0	n/a	

Perturbations applied at every 3 h forcing time step (or 7.5 min model time step).

Calibration of **model and observation error** parameters guided by validation vs. in situ measurements and by internal assimilation diagnostics.

Validation at SMAP Core Validation Sites De Lannoy et al., Jul 2010 – Jun 2014 ubRMSE anomaly R 2014, in preparation. Surface soil moisture [m³/m³] **Black: Model** Red: Assimilation 0.04 0.5 0.02 0 0 RC1 RC2WG1WG2WG3 LW FC LR1 LR2 SF RC1 RC₂ Root zone soil moisture [m³/m³] 0.04 0.03 0.5 0.02 0.01 0 0 LW FC LR1 LR2 SF LW FC LR1 LR2 SF Surface soil temperature [K] 10.72 8.04 5.36 0.5 2.68 0

LR1 LR2

FC

SF

RC1 RC2WG1WG2WG3

LW

FC

LR1

SF

LR2

RC1

RC2WG1WG2WG3 LW





Calibration of the Data Assimilation System

Std-dev of normalized observation-minus-forecast residuals





Further calibration underway using newly implemented infrastructure for **spatially distributed** perturbation std-devs and observations error std-devs.

De Lannoy et al., 2014, in preparation.



SMAP L4_SM Analysis Overview





Freeze-thaw OSSE

OL = Open loop (no assimilation)DA = Assimilation of synthetic F/T obs. $\Delta RMSE = RMSE(OL) - RMSE(FT)$

Small improvements with realistic classification errors:



	OL	ΔRMSE* [K]				
bs.	RMSE*	Max. Classification Error				
	[K]	0%	5%	10%	20%	
Tsurf	3.08	0.21	0.20	0.18	0.15	
Tsoil	1.97	0.06	0.05	0.04	0.01	

*Excl. times & locations with Tair>7°C or Tair<-7°C





See talk by Leila Farhadi (Wed. am) Farhadi et al., 2014, JHM, conditionally accepted.





Conclusions

Multi-source observations:

- Address sensor-specific observation bias, e.g.,
 - geostationary Tskin.
- Improved skill with assimilation of data from multiple sources, e.g.,
 - active and passive surface soil moisture retrievals,
 - precipitation corrections and soil moisture assimilation,
- but this does not always work, e.g.,
 - MODIS SCF and AMSR-E SWE assimilation, b/c of poor-quality SWE retrievals.

Multi-variate analysis:

- Assimilation of GRACE TWS retrievals yields improvements in soil moisture and snow estimates.
- Assimilation of SMOS L-band brightness temperature yields improvements in soil moisture but not in soil temperature estimates.
- The SMAP Level 4 Surface and Root Zone Soil Moisture (L4_SM) product is based on multi-source observations and a multi-variate analysis.
 In progress:
- Assim. of Tskin & L-band Tb; Assim. of GRACE TWS & L-band Tb.



Thank you for your attention!









NASA

Geostationary Tskin observations





Geostationary Tskin observations



FIG. 4. Histograms of the state update innovations at 21:00 UTC, for the assimilation of geostationary T_{skin} , at the Goodwin Creek (GWN), Sioux Falls (SXF), and Desert Rock (DRA) SURFRAD sites, for a bias-blind assimilation (upper), and for the two-stage observation bias and state estimation bias-aware assimilation with $\tau=20$ days (lower).

Assimilation of AMSR-E snow water equivalent (SWE)



AMSR-E does not observe thin snow packs

 \rightarrow no improvement at start and end of snow season.

Some improvement at lower elevations with shallow snow pack (COOP sites). Problematic in mountains with deep snow packs (SNOTEL sites).

De Lannoy et al. WRR (2011) submitted.