



4th World Climate Research Programme International Conference on Reanalyses

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Land Surface Analysis and Reanalysis at the NASA GMAO

Rolf Reichle

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Outline

Introduction

MERRA-Land

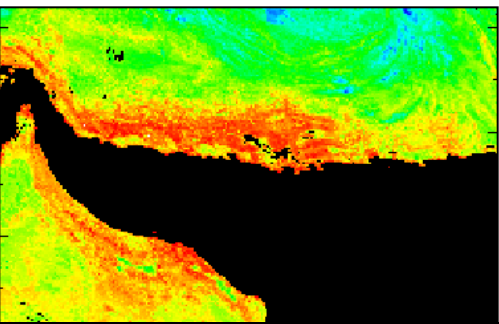
- Motivation and data product design
- Validation

Developments in land surface analysis

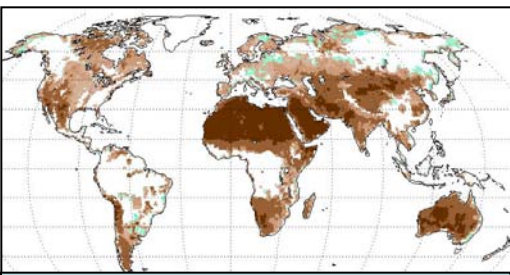
- Soil moisture
- Snow



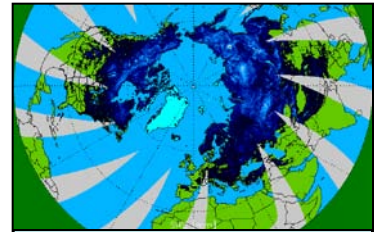
Satellite observations



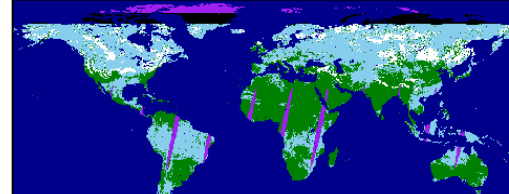
Land surface temperature (MODIS, AVHRR, GOES, ...)



Surface soil moisture (SMMR, TRMM, AMSR-E, ASCAT, SMOS, SMAP)



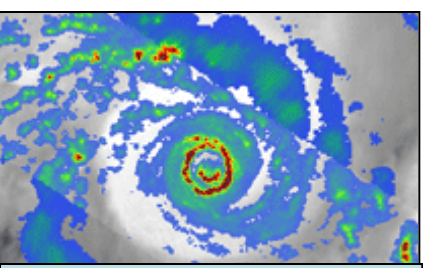
Snow water equivalent (AMSR-E, SSM/I, SCLP)



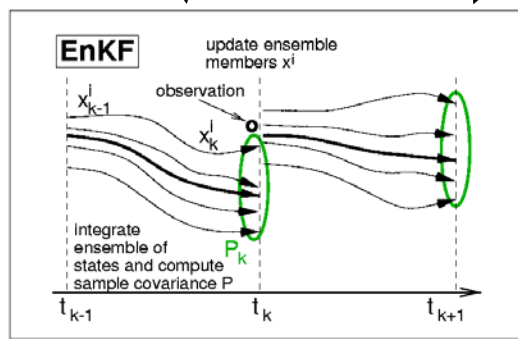
Snow cover fraction (MODIS, VIIRS)



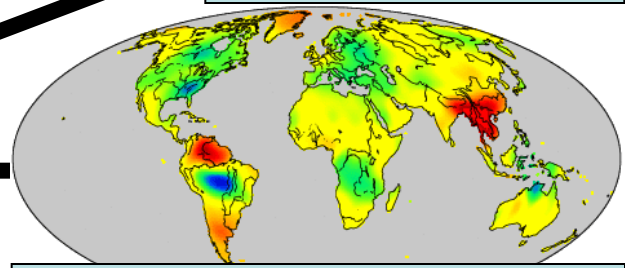
Water surface elevation (SWOT)



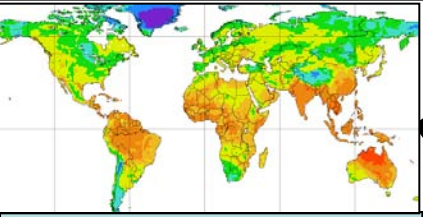
Precipitation (TRMM, GPM)



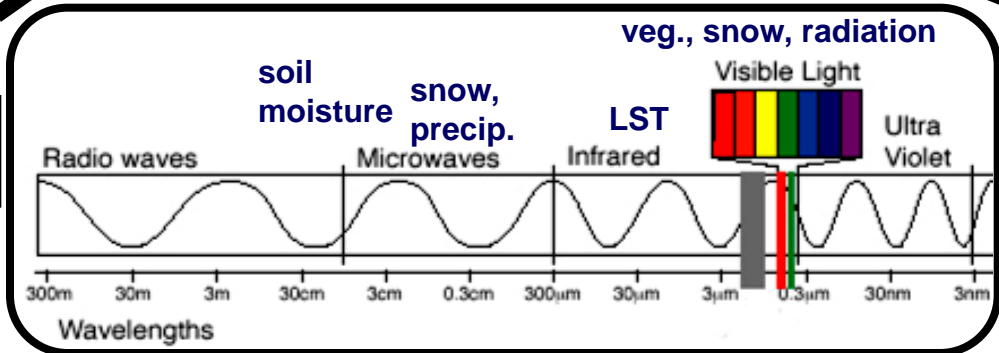
Land data assimilation system



Terrestrial water storage (GRACE)



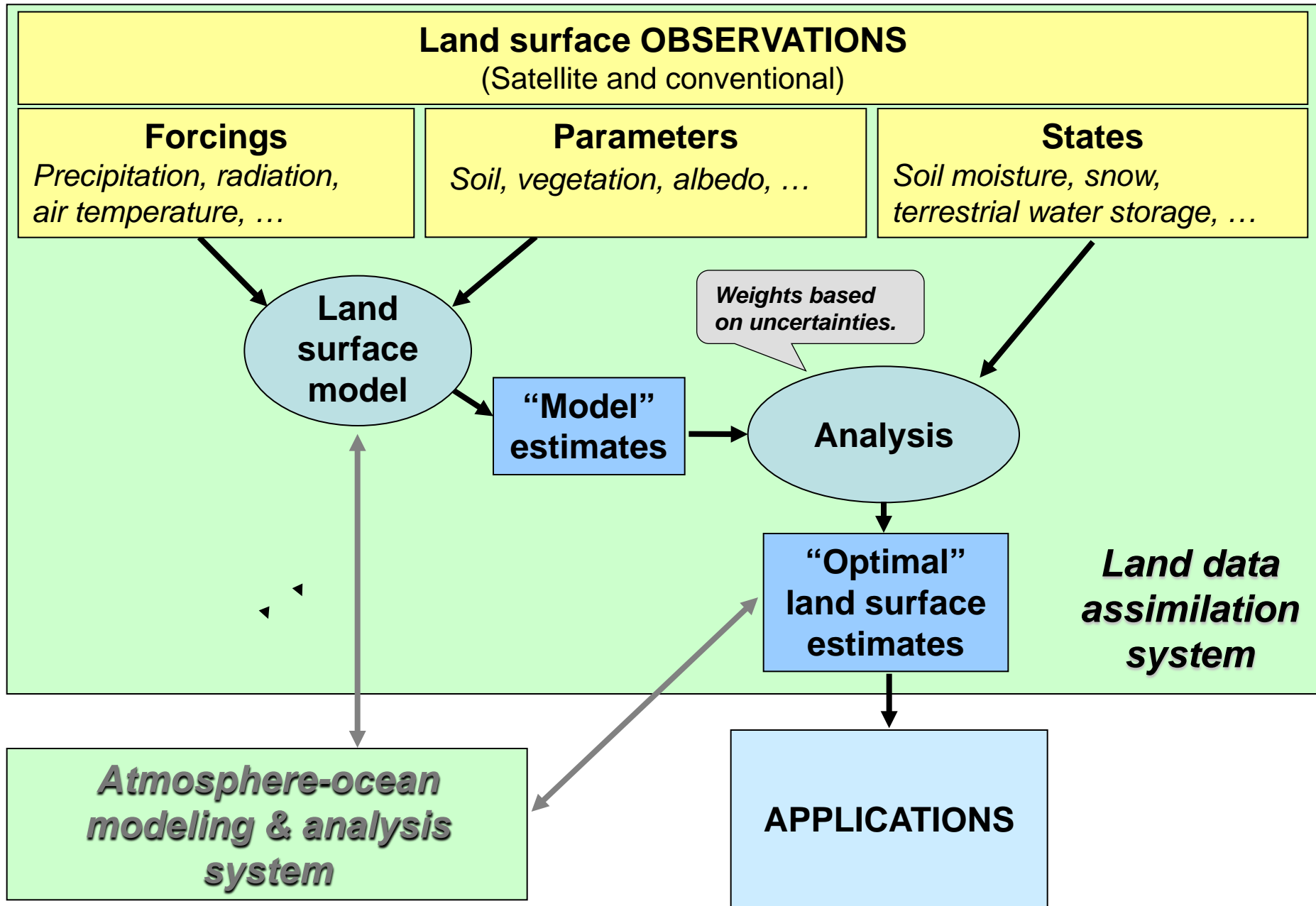
Radiation (CERES, CLARREO)



Vegetation/Carbon (AVHRR, MODIS, DESDynI, ICESat-II, HypSIRI, LIST, ASCENDS)

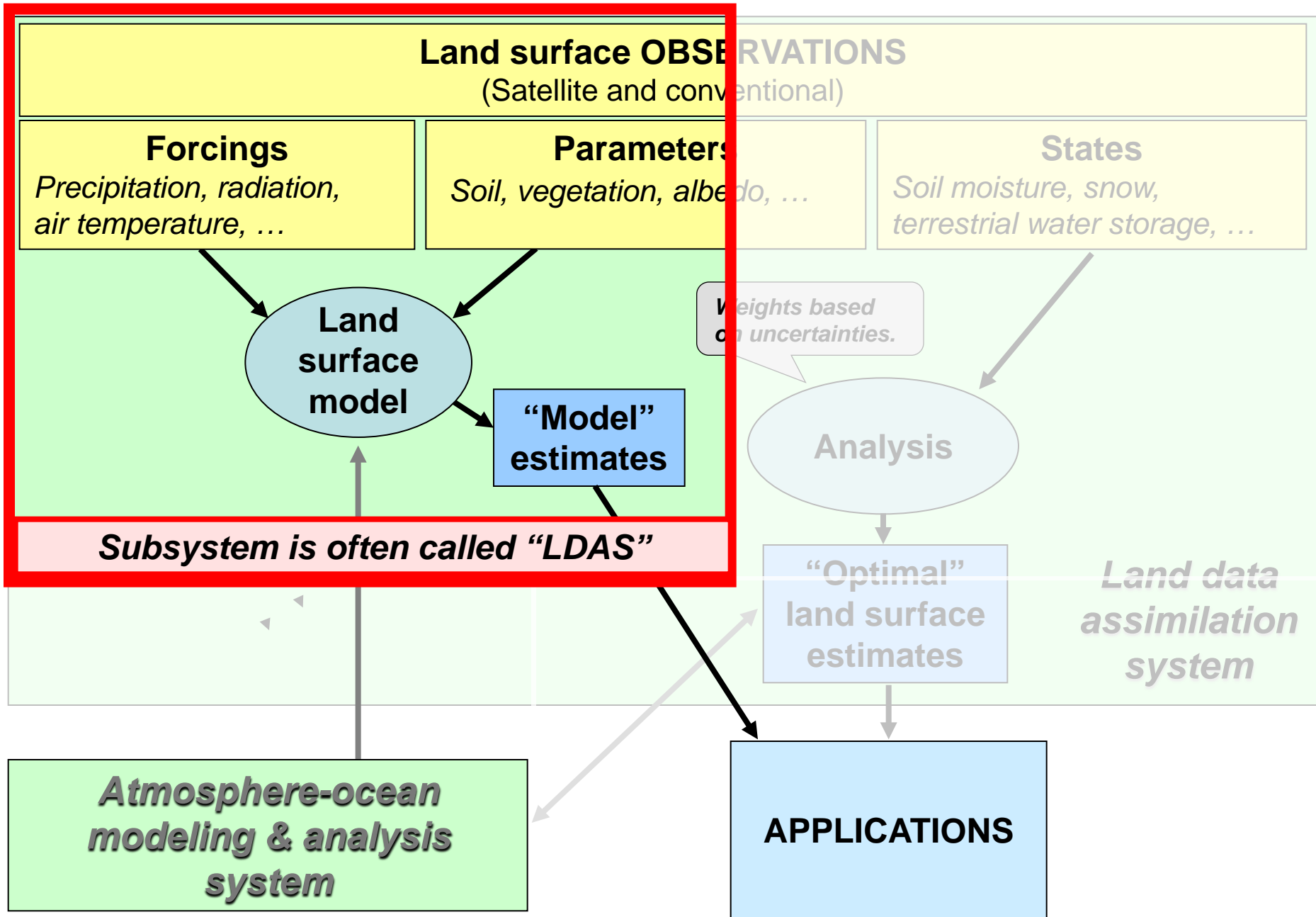


A generic land data assimilation system



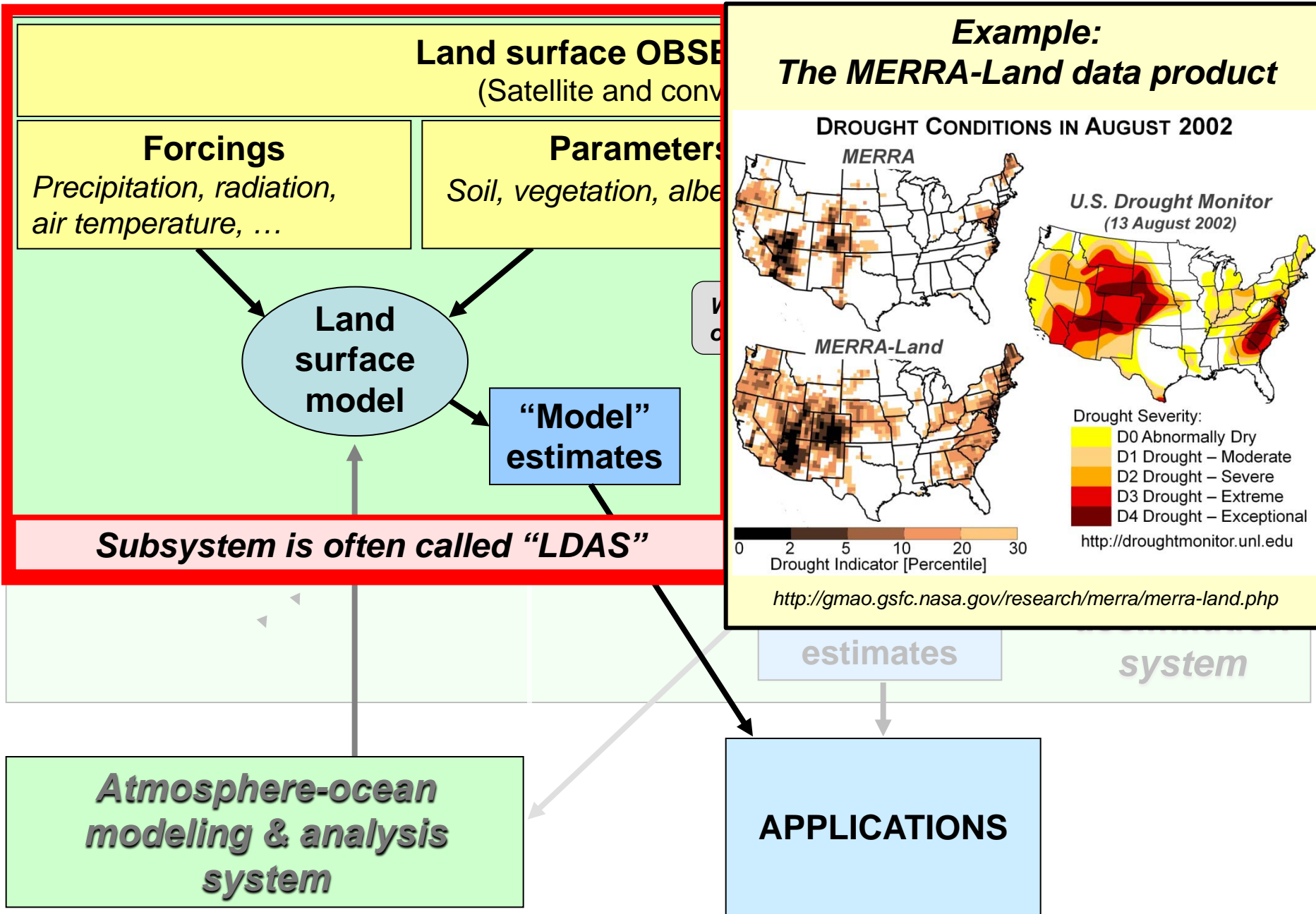


A generic land data assimilation system





A generic land data assimilation system



Land surface OBSERVATIONS
(Satellite and conventional)

Forcings

Precipitation, radiation,
air temperature, ...

Parameters

Soil, vegetation, albedo, ...

Land surface model

“Model” estimates

Subsystem is often called “LDAS”

Atmosphere-ocean modeling & analysis system

APPLICATIONS

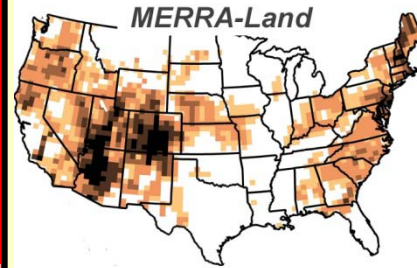
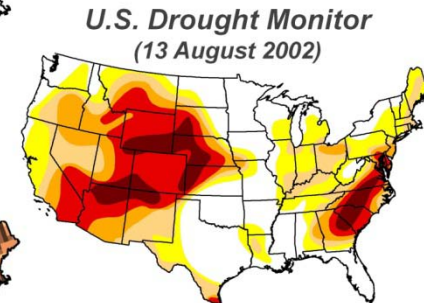
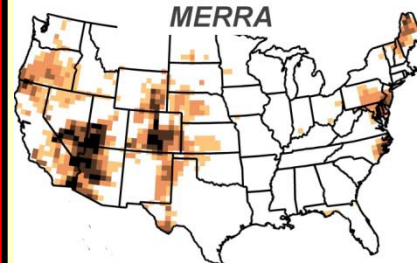
estimates

system

Example:

The MERRA-Land data product

DROUGHT CONDITIONS IN AUGUST 2002



Drought Severity:

- D0 Abnormally Dry
- D1 Drought – Moderate
- D2 Drought – Severe
- D3 Drought – Extreme
- D4 Drought – Exceptional

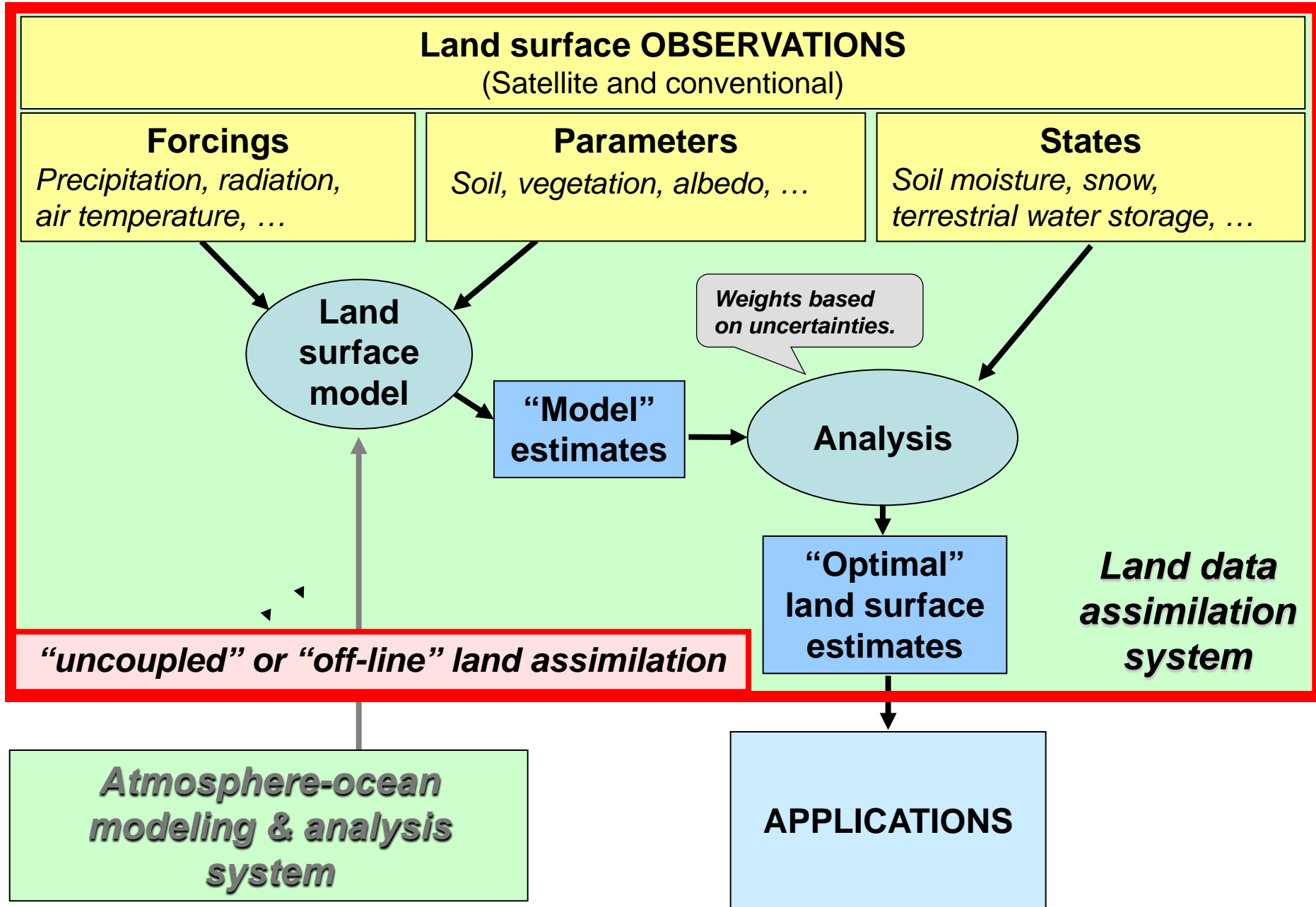
<http://droughtmonitor.unl.edu>



<http://gmao.gsfc.nasa.gov/research/merra/merra-land.php>

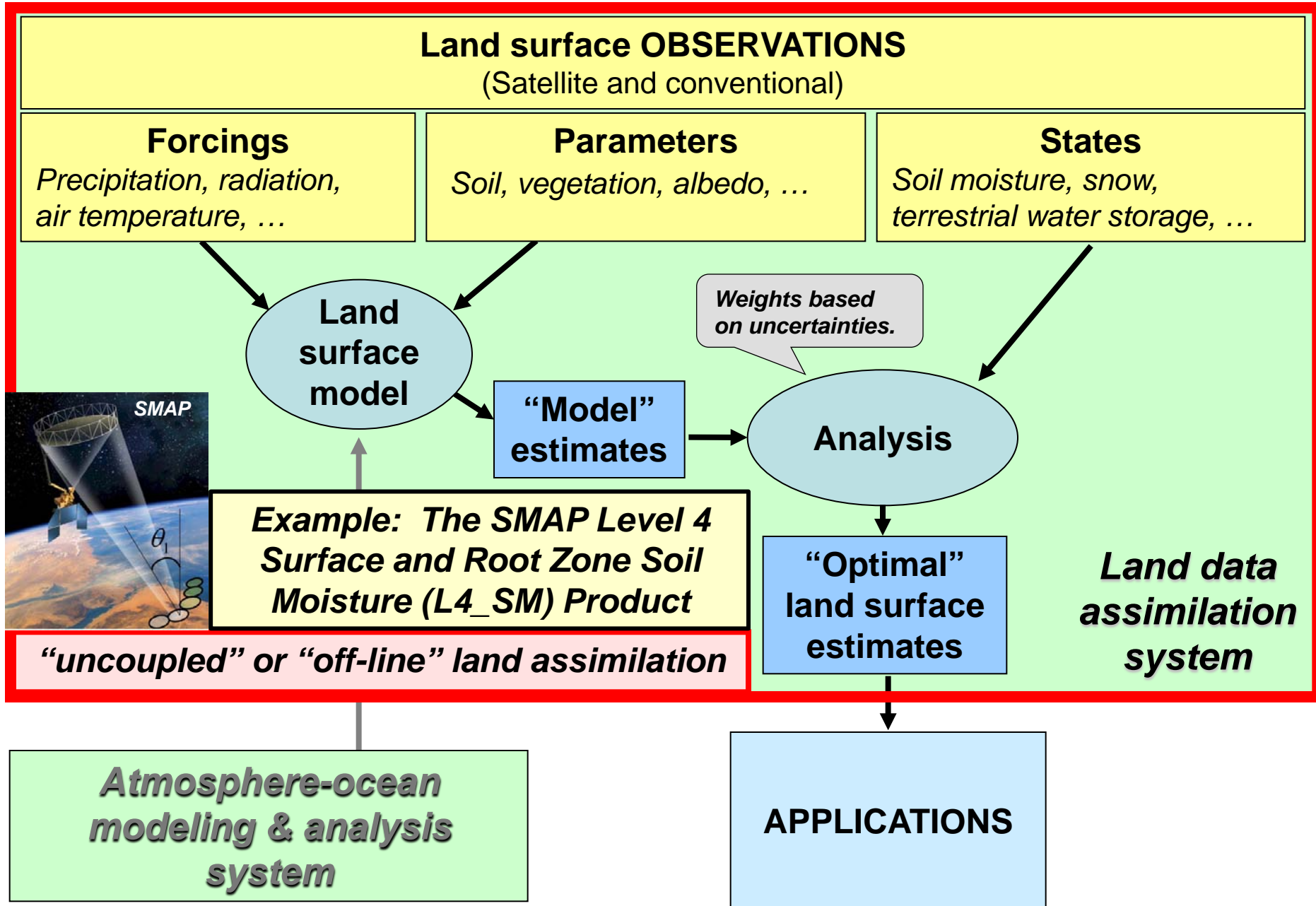


A generic land data assimilation system



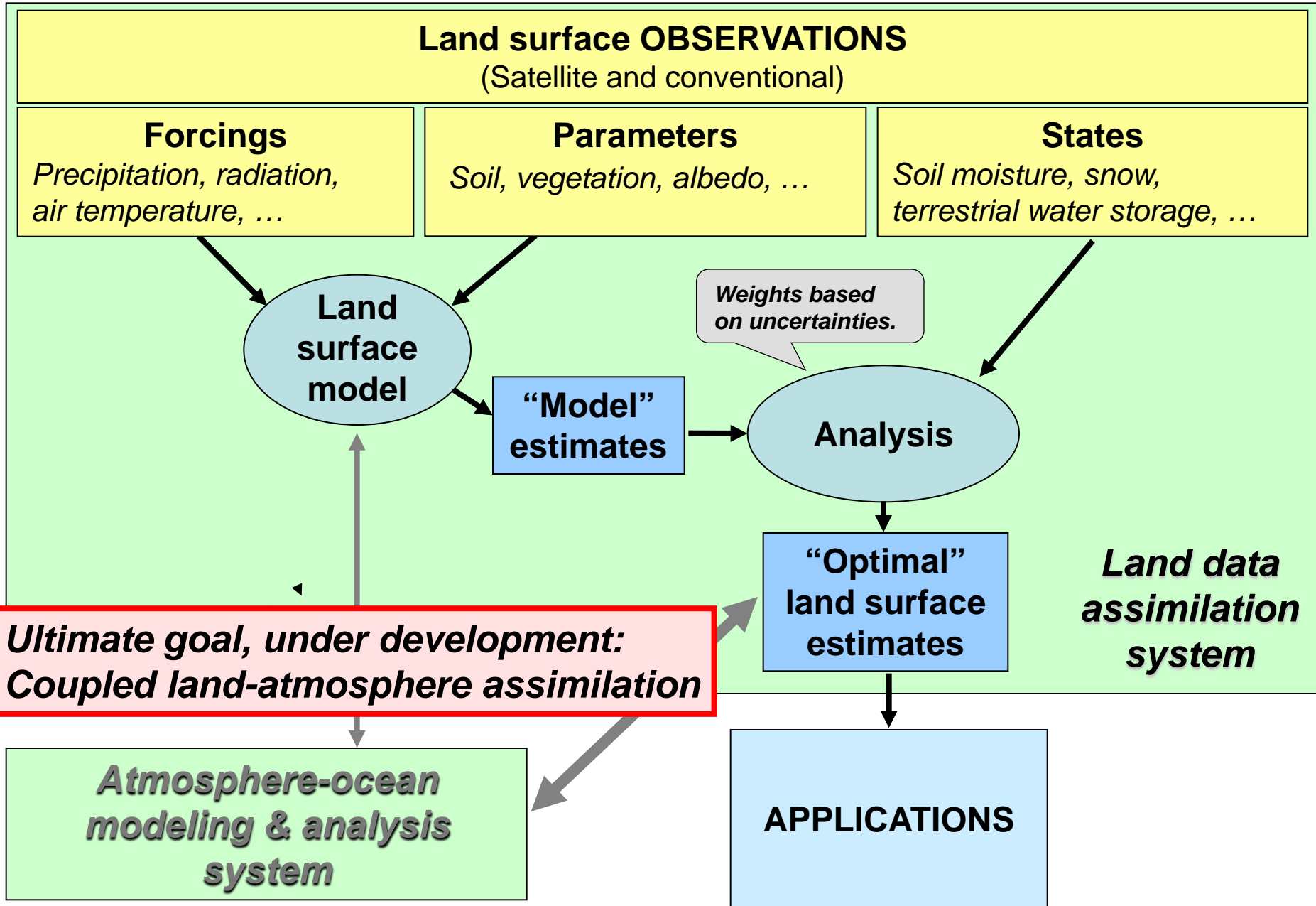


A generic land data assimilation system





A generic land data assimilation system





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MERRA-Land

- Motivation and data product design
- Validation

Developments in land surface analysis

- Soil moisture
- Snow

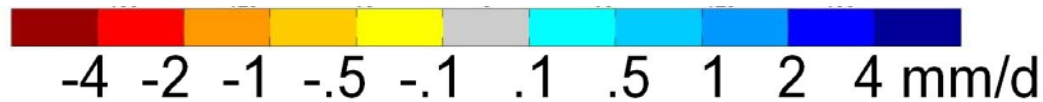
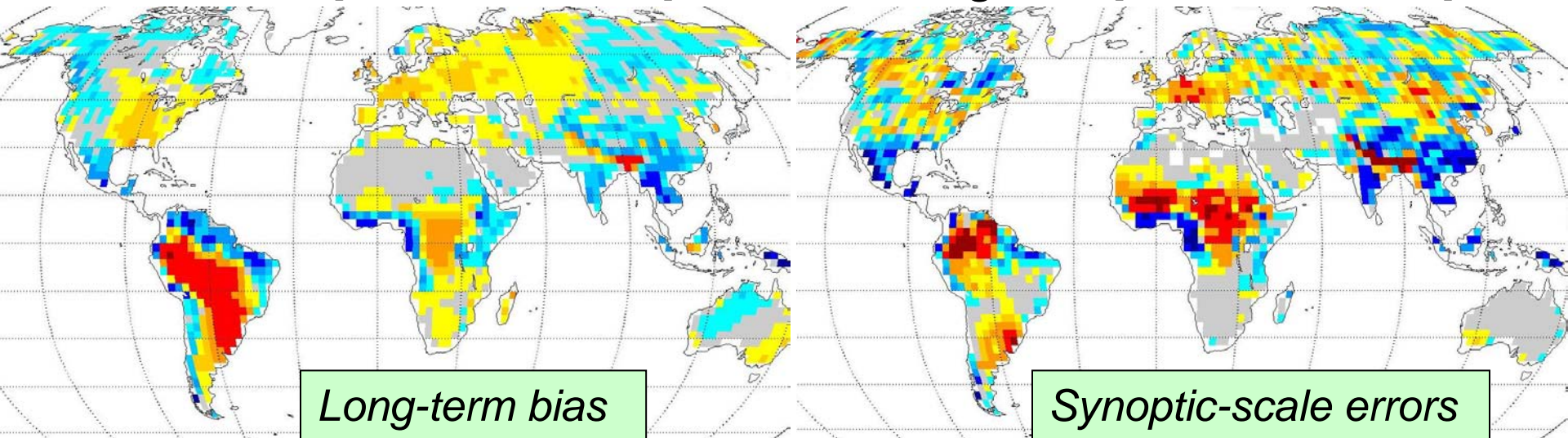


Motivation for MERRA-Land: Precipitation

MERRA – GPCPv2.1

1981-2008 [mean=-0.03 mm/d]

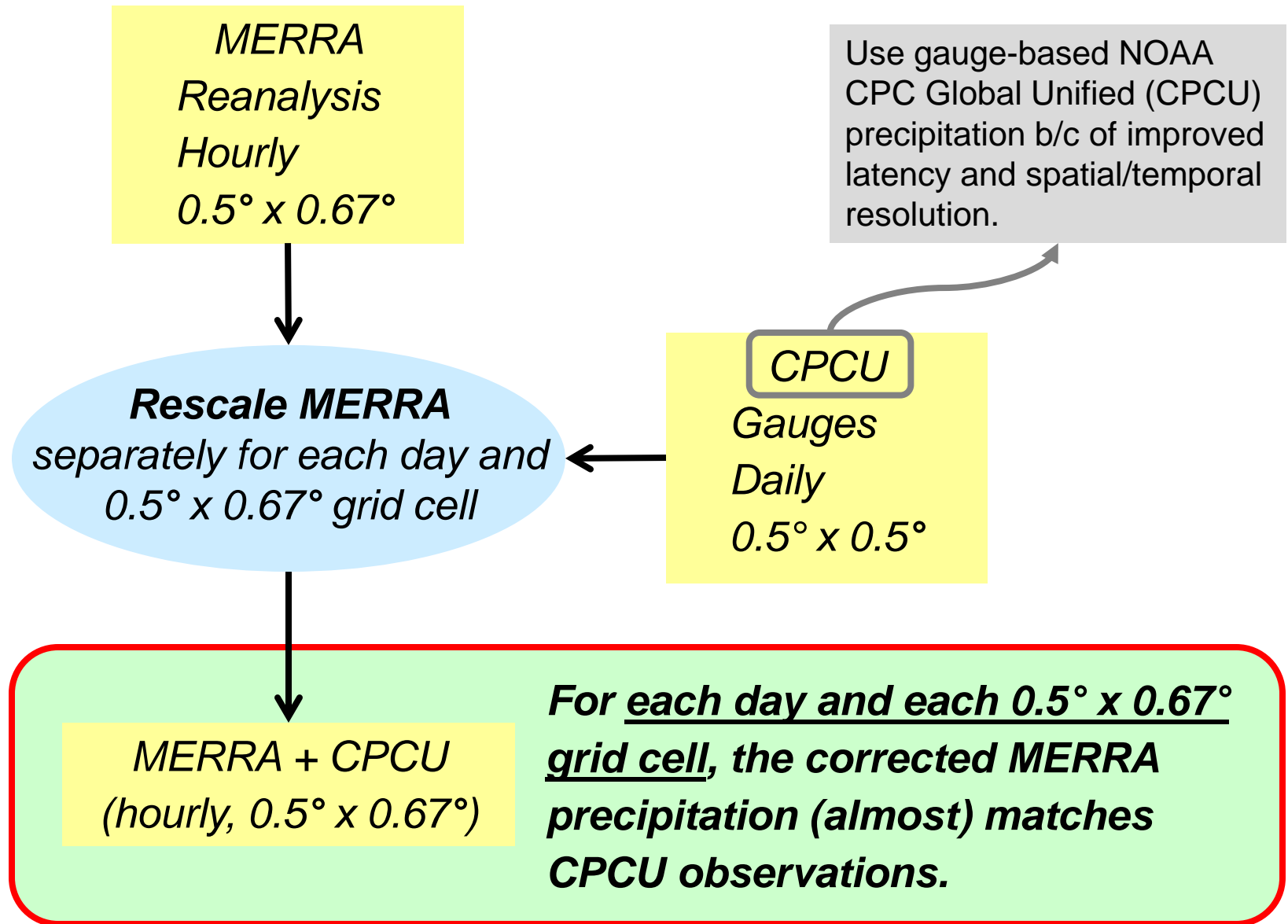
Aug 1994 [mean=-0.04 mm/d]



→ Correct MERRA precipitation with gauge-based precipitation observations to the extent possible.



MERRA-Land precipitation corrections



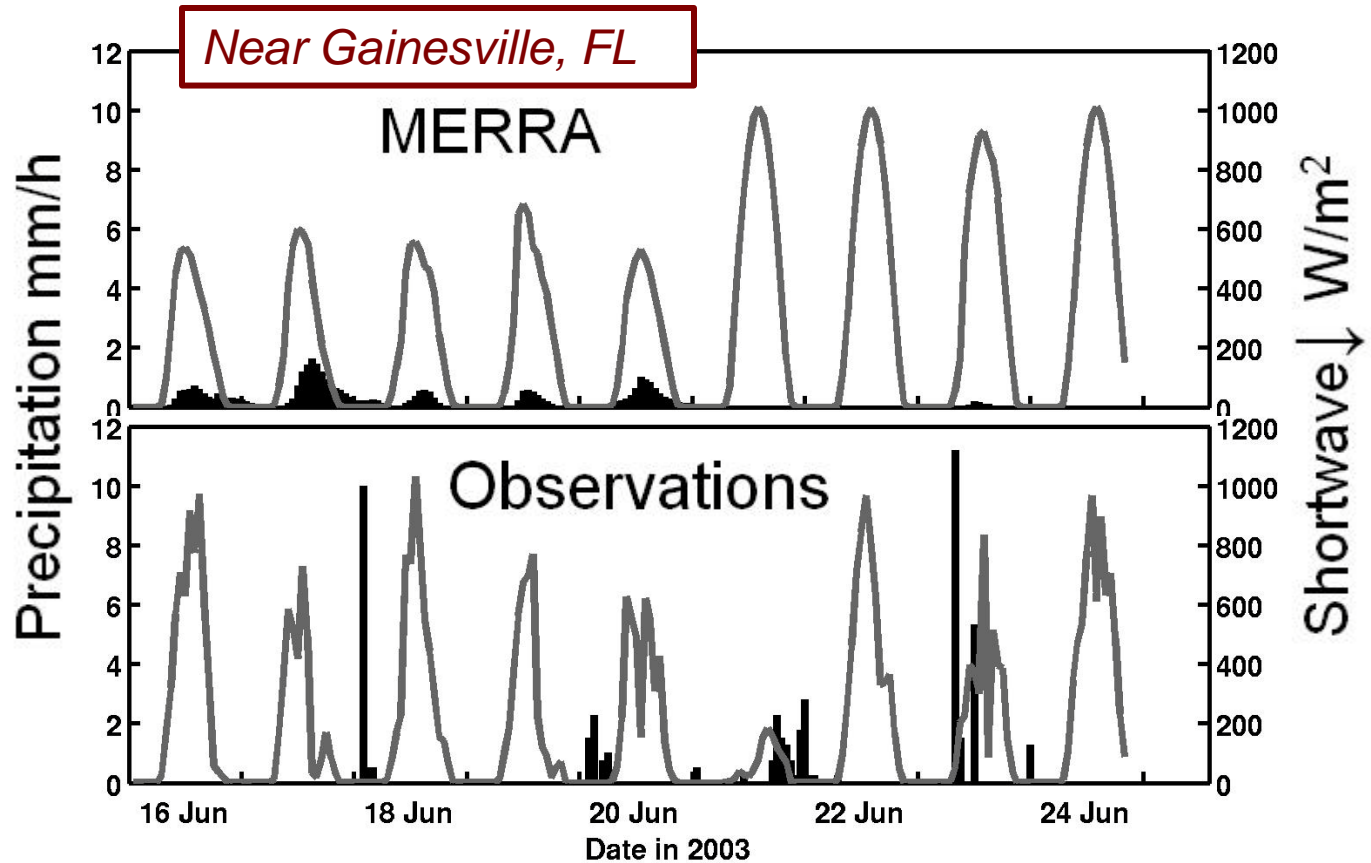


MERRA precipitation and radiation forcing

MERRA precip. has short-term errors in

1. intensity,
2. timing, and
3. consistency with SW radiation forcing.

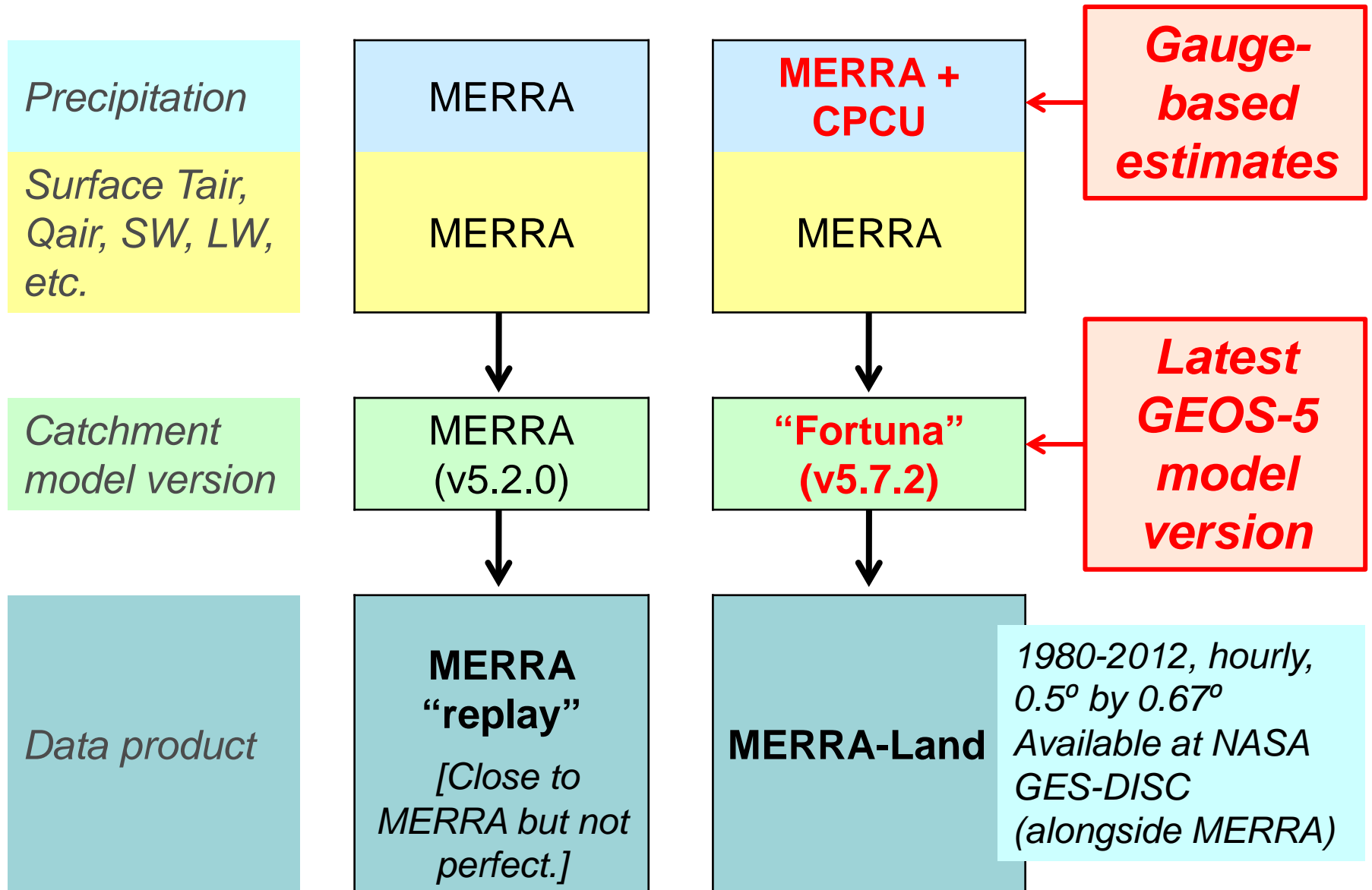
Too much canopy evaporation
→ not enough water reaches the soil!



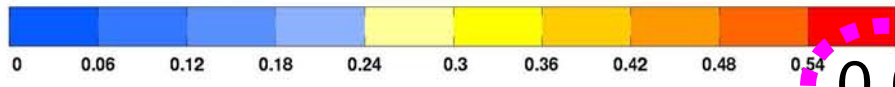
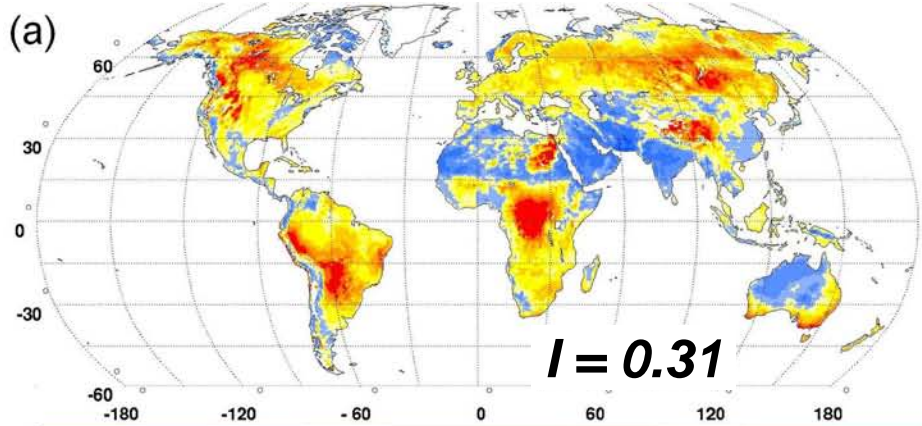
Additional fix:
Change Catchment model interception parameters.



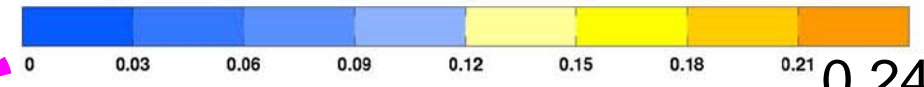
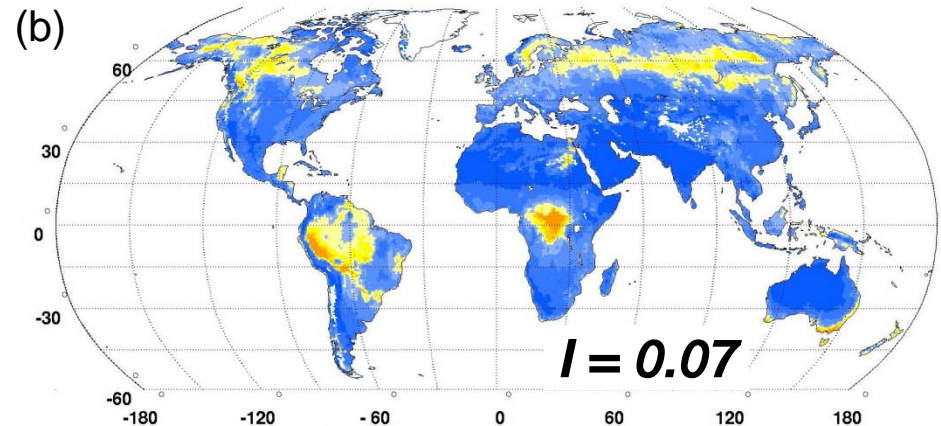
Land-only (“off-line”) replay



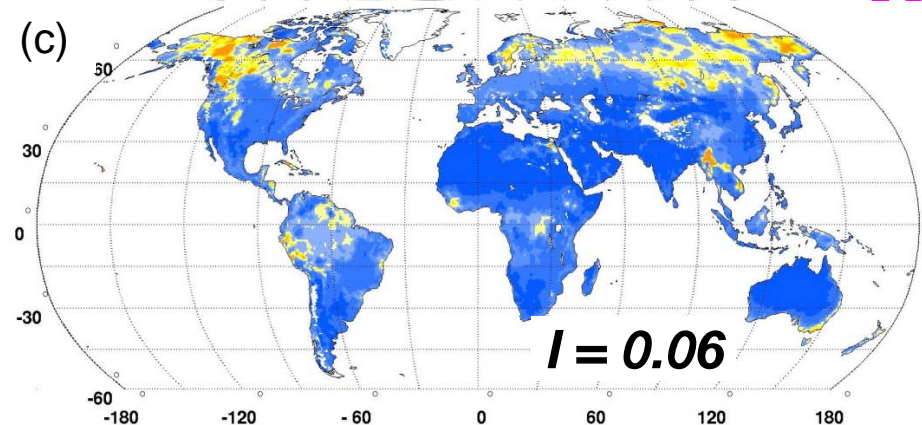
MERRA



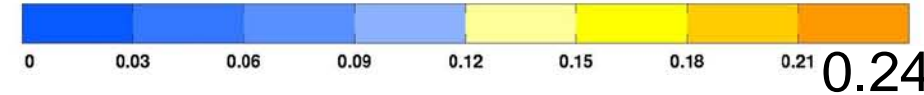
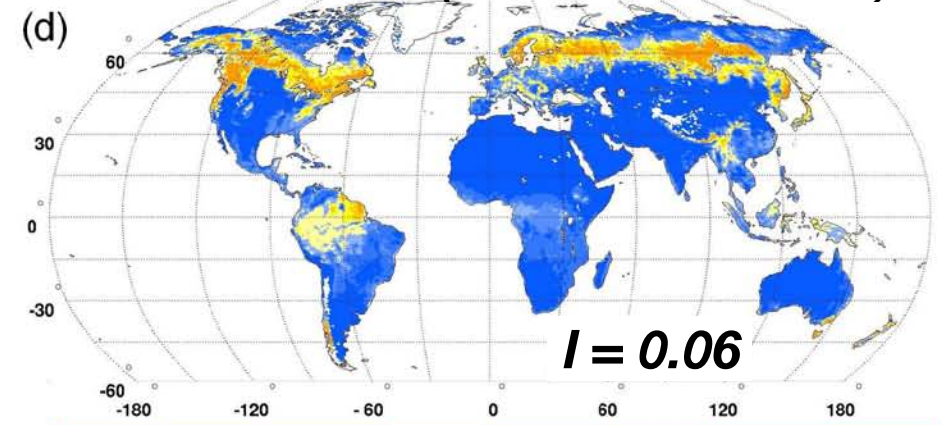
Revised Catchment model



MERRA-Land ("final")



"Observations" (Miralles et al. 2010)

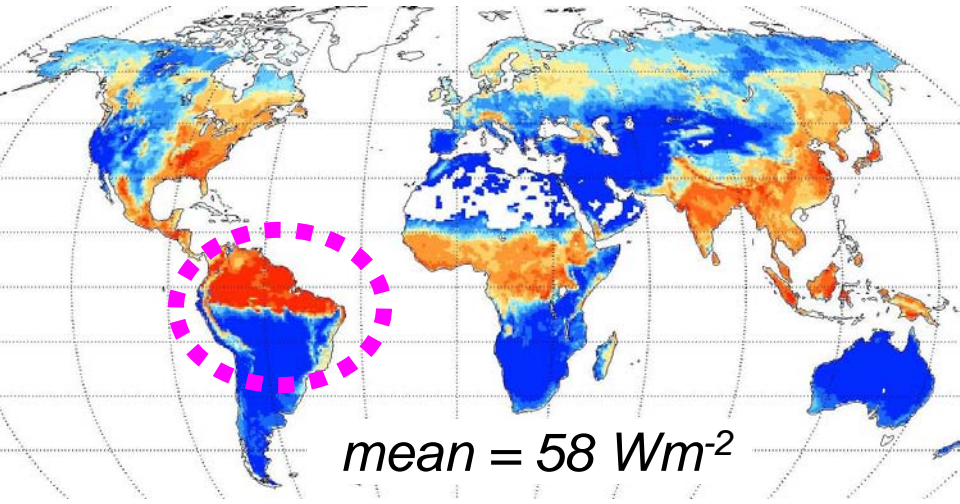


*Improvement everywhere from revised interception parameters (b).
Additional improvement from precipitation corrections (c).*



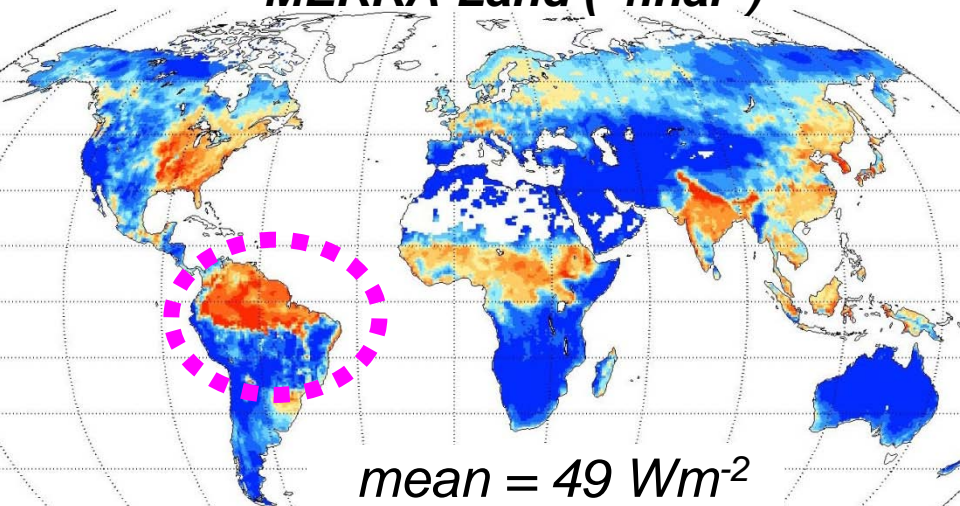
Latent heat flux (August 1994)

MERRA

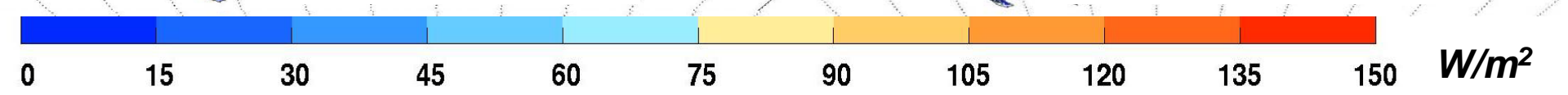
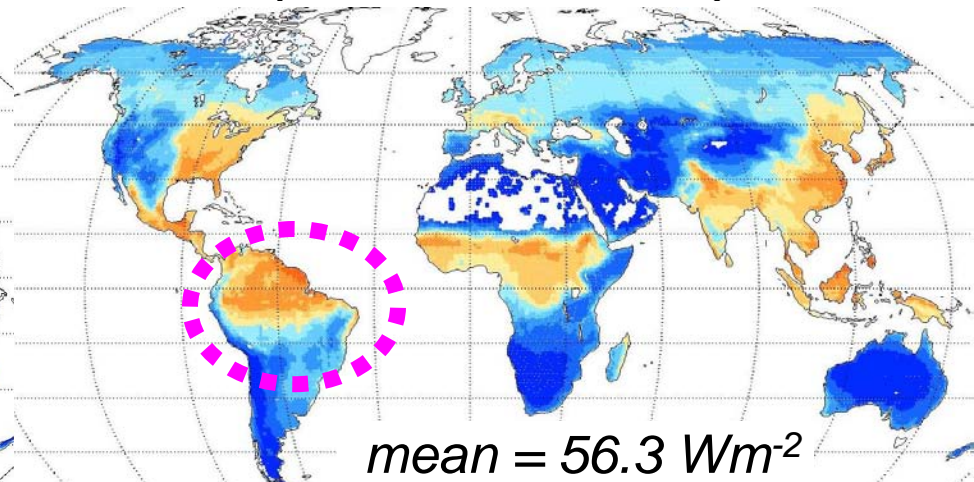


MERRA-Land has more realistic LH over Amazon during dry season.
Attributed to revised interception parameters and precipitation corrections.

MERRA-Land ("final")



**Multi-product average
(Jimenez et al. 2011)**





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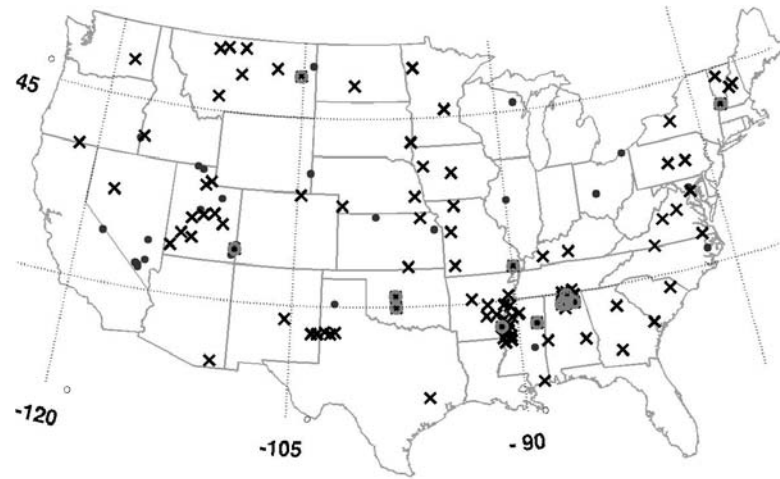
Developments in land surface analysis

- Soil moisture
- Snow



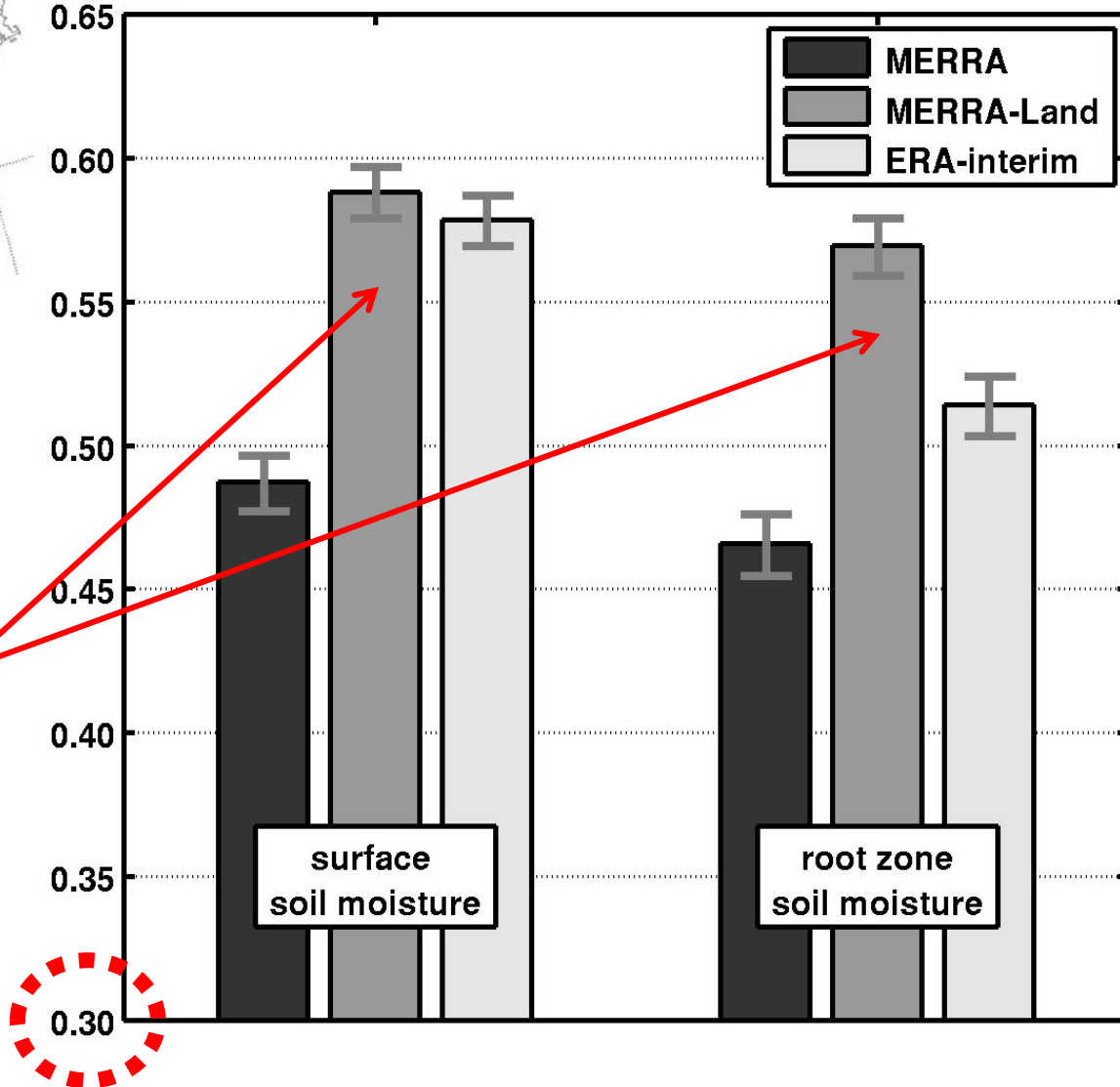
Soil moisture validation (2002-2009)

*Skill (pentad anomaly R)
v. SCAN in situ observations*



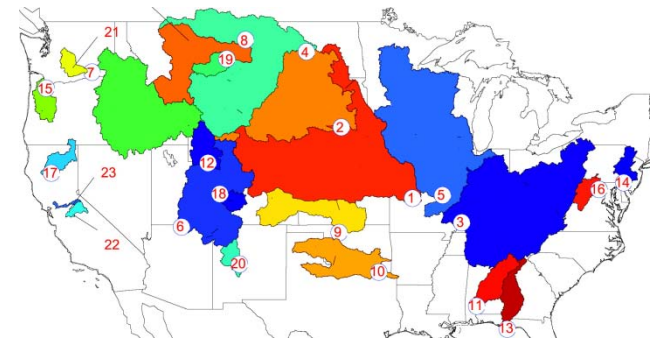
MERRA-Land has better soil moisture anomalies than MERRA (attributed to precipitation corrections).

MERRA-Land root zone skill better than ERA-Interim.





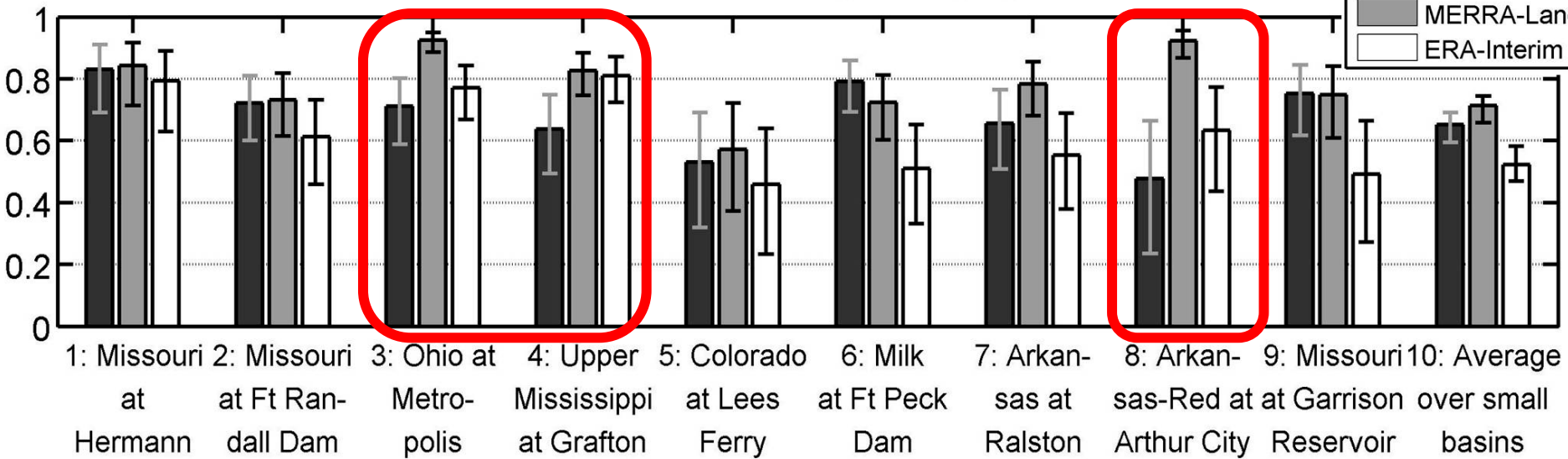
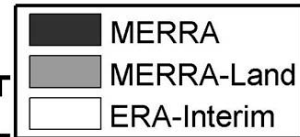
Runoff



NB: Numbering does not match figure below.

Validation against naturalized streamflow observations from 9 "large" and 9 "small" basins (~1989-2009).

Streamflow skill (anomaly R)

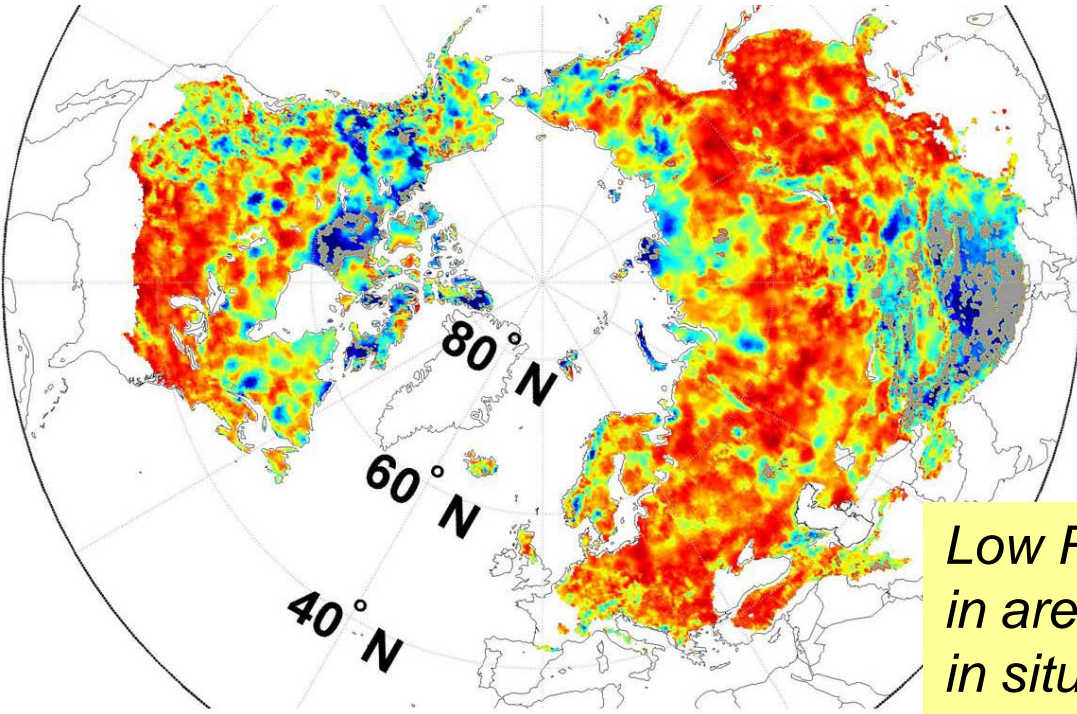


Precipitation corrections yield significantly better runoff for 3 basins. MERRA and MERRA-Land (0.5 deg) better than ERA-Interim ("1.5 deg"). Not shown: In all cases the revised interception parameters yield improved runoff anomalies (albeit not significant).

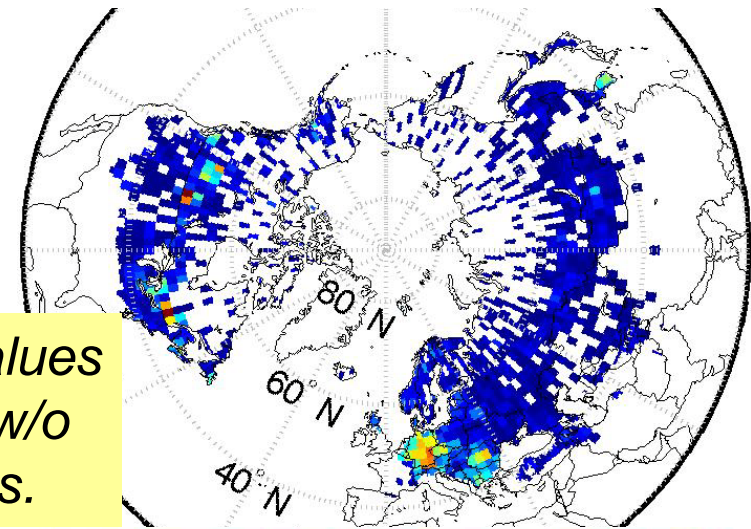
Snow depth

MERRA-Land v. CMC snow analysis Pentad anomaly R (2002-2009)

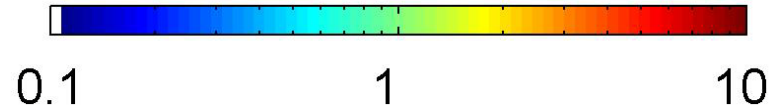
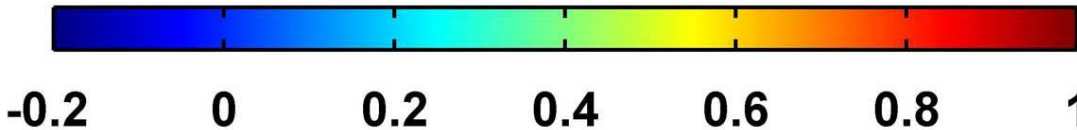
Note: **No** snow analysis in MERRA or MERRA-Land.



CMC snow analysis Density [stations/10,000 km²]



Low R values
in areas w/o
in situ obs.



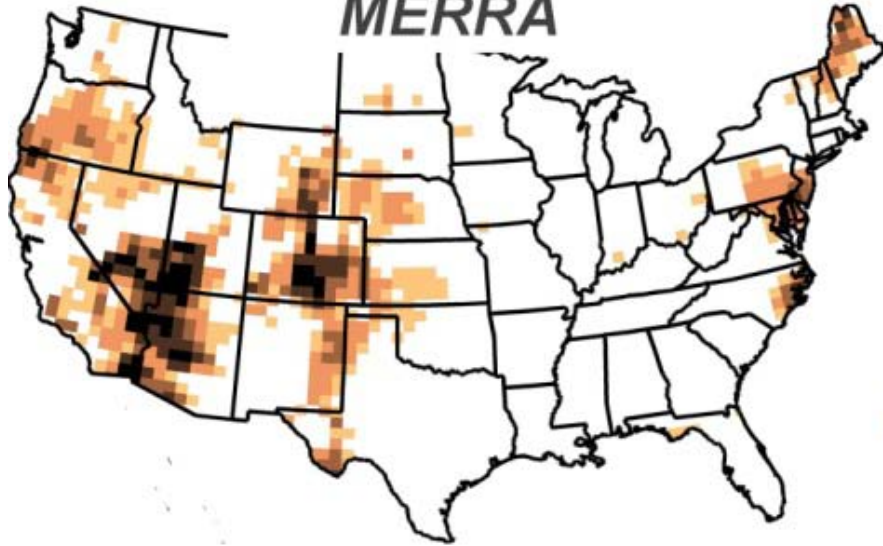
**NOT
SHOWN**

- MERRA and MERRA-Land have similar skill.
- Similar results for comparison vs. in situ obs. (583 stations).
- Similar results for snow water equivalent (SWE).
- Step in 1998/99 in CPCU high-lat avg precip. → step in snow mass.

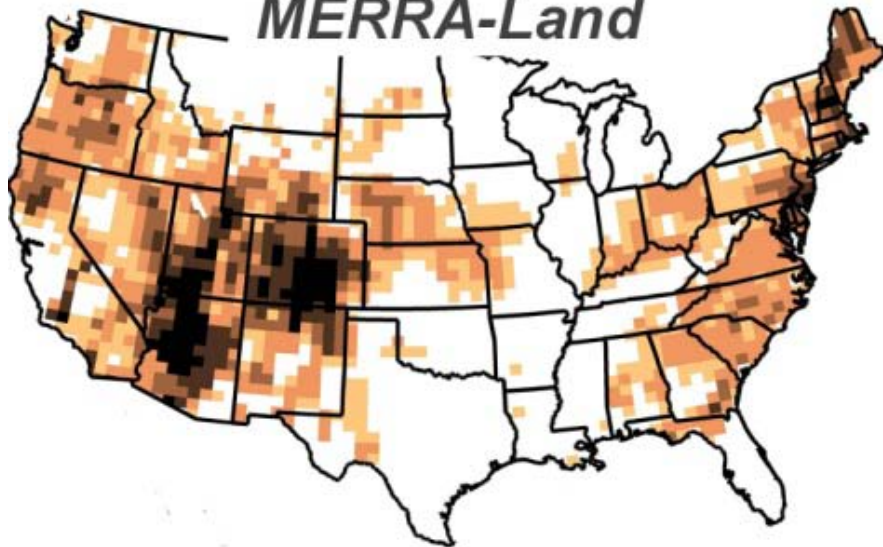


DROUGHT CONDITIONS IN AUGUST 2002

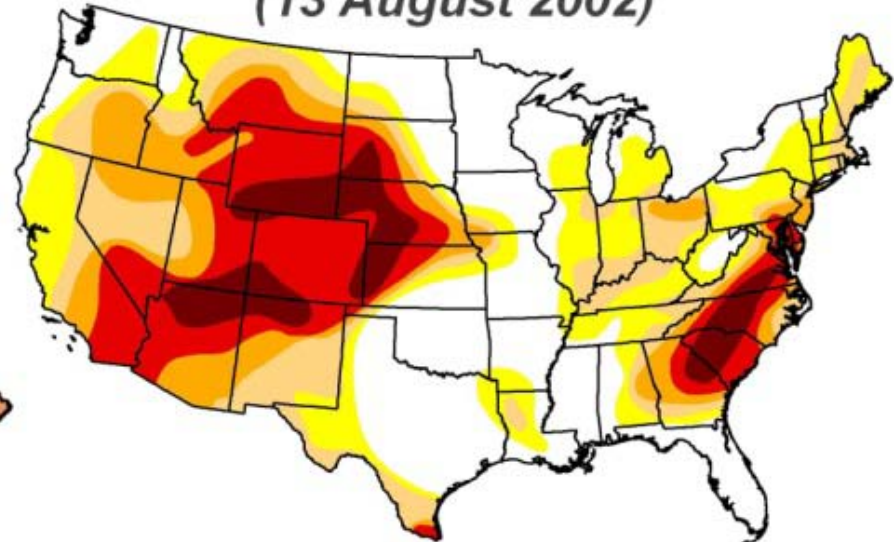
MERRA



MERRA-Land



*U.S. Drought Monitor
(13 August 2002)*



Drought Severity:



<http://droughtmonitor.unl.edu>



<http://gmao.gsfc.nasa.gov/merra/merra-land.php>



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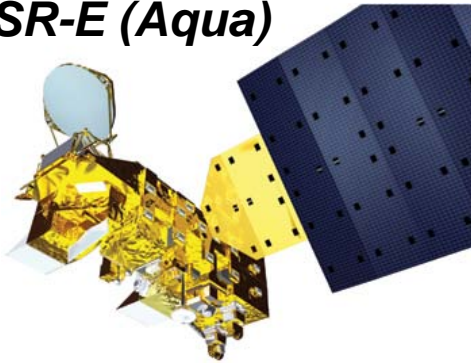
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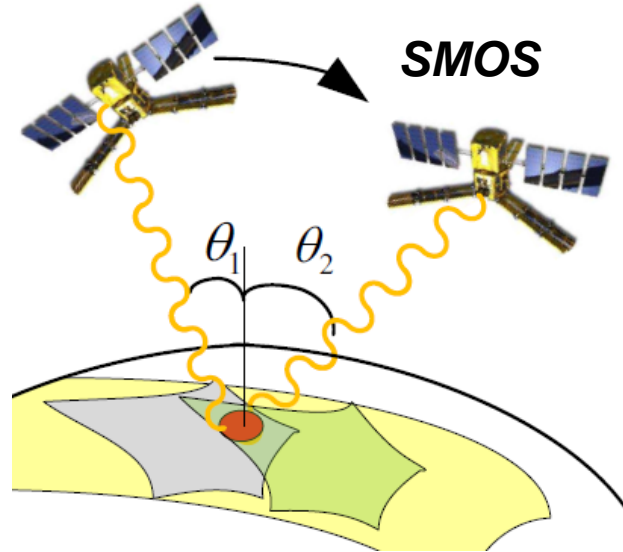


Satellite remote sensing of (surface) soil moisture

AMSR-E (Aqua)



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



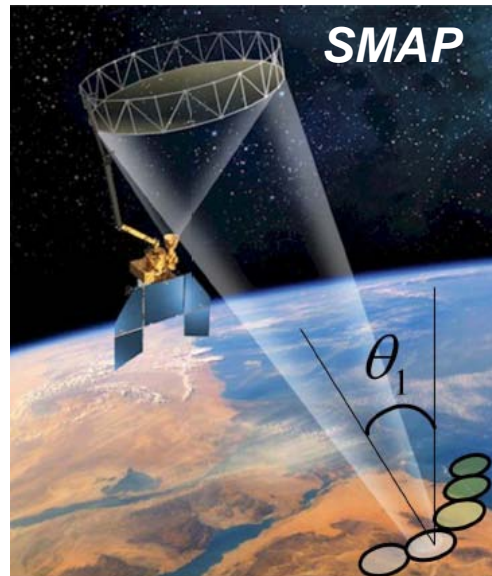
SMOS

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

ASCAT (Met-op)



2007-present
C-band active
40 km resolution



SMAP

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

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



Soil moisture assimilation

Skill increases significantly through data assimilation.

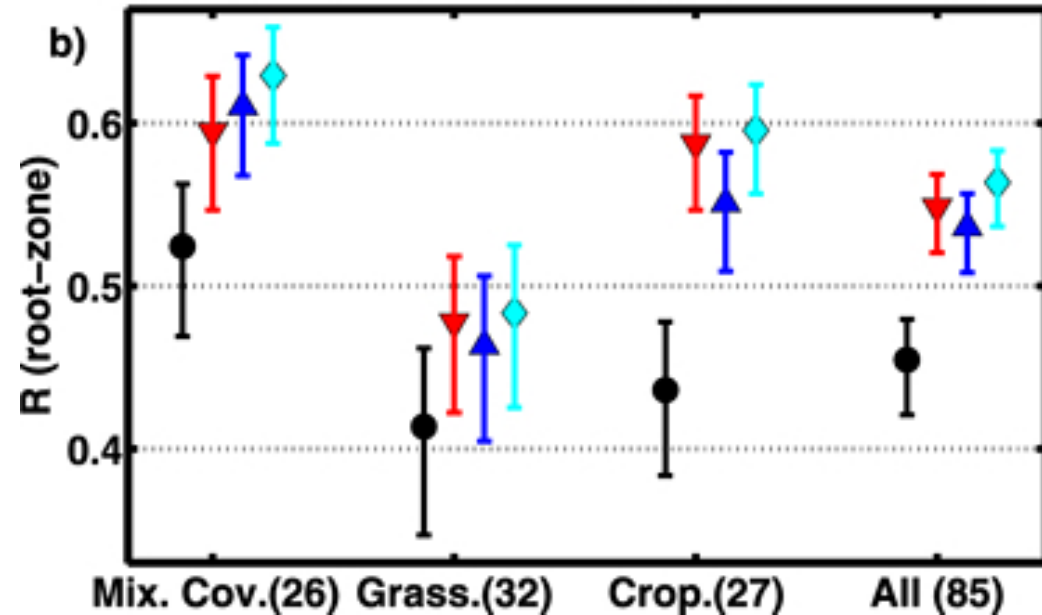
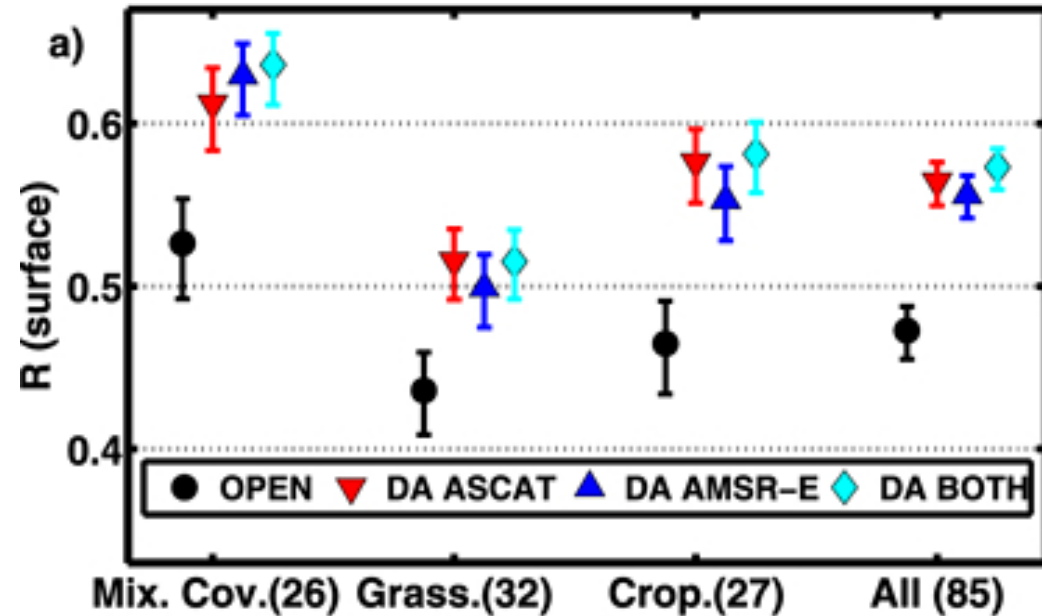
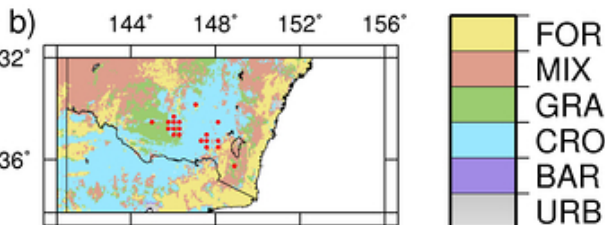
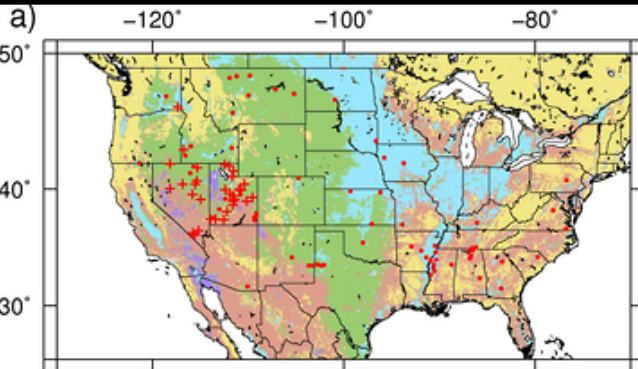
Similar improvements from AMSR-E and ASCAT.

Root-zone *not* observed by satellite. **Improvements** may be critical for applications.

Metric: Anom. time series corr. coeff.

Anomalies \equiv mean seasonal cycle removed

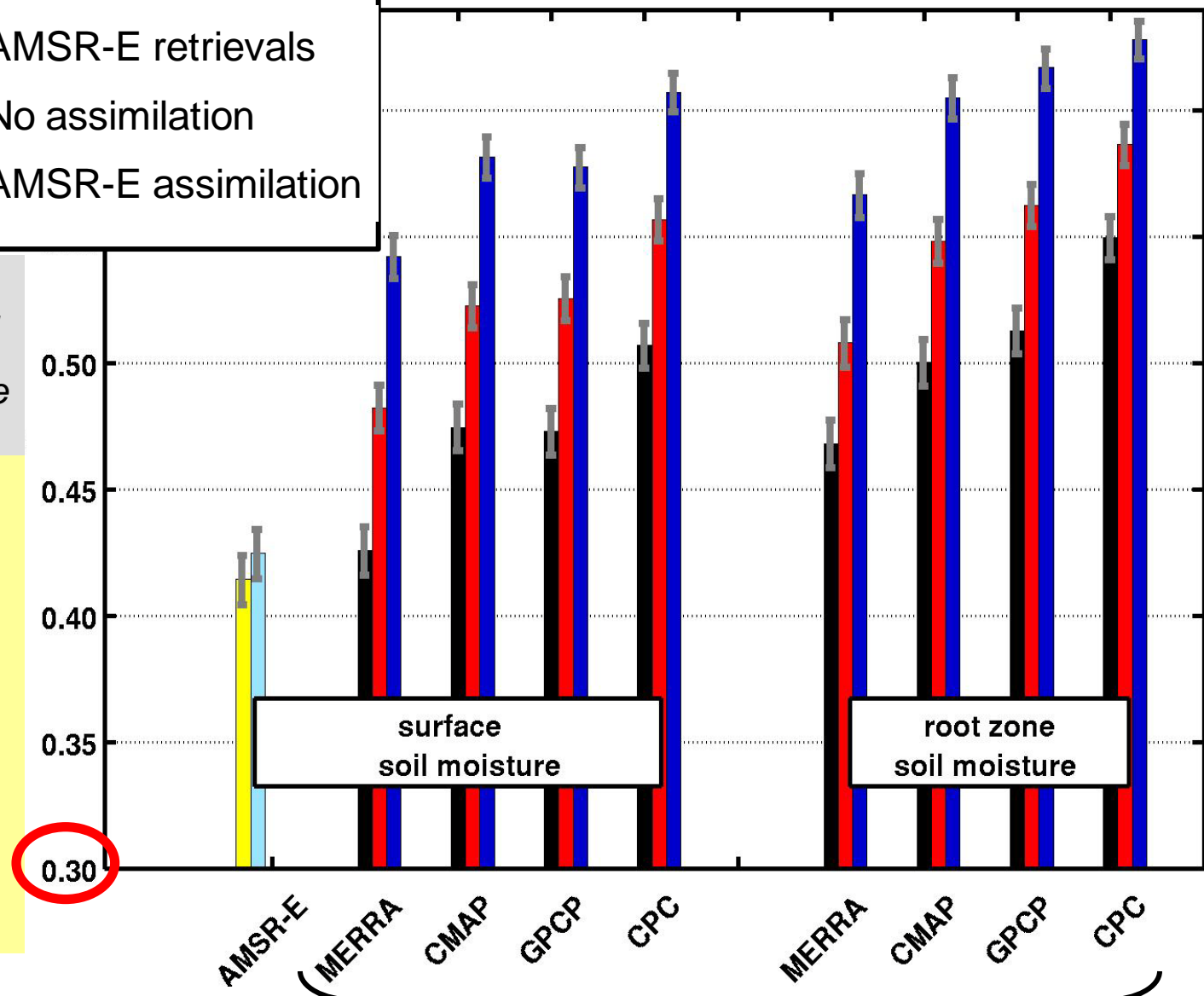
Validated with in situ data





Precipitation corrections v. retrieval assimilation

Skill v. SCAN in situ obs



Anomalies \equiv mean seasonal cycle removed

Skill metric: Anom. time series corr. coeff. R

Soil moisture skill increases with

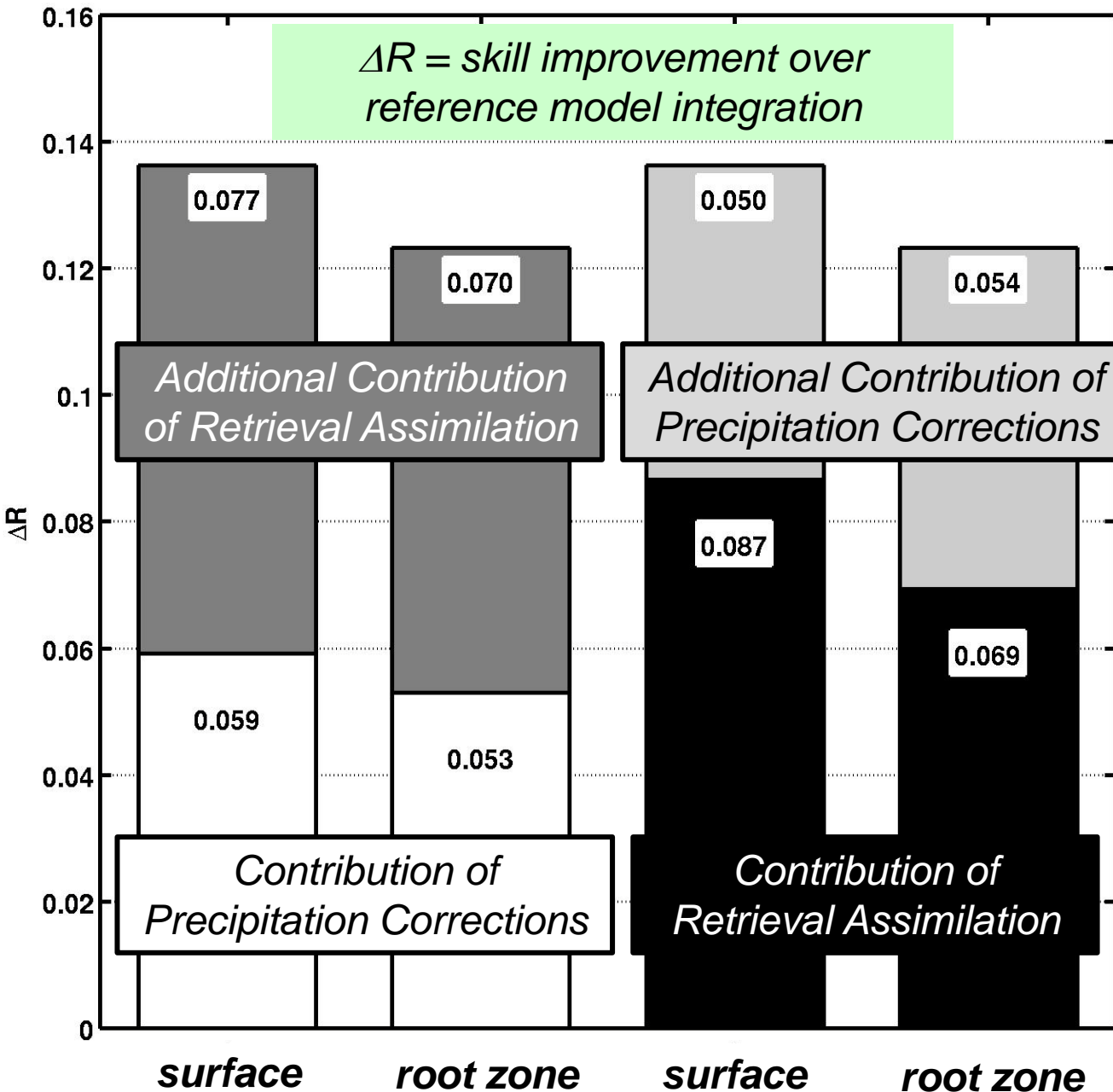
- precipitation corrections and
- assimilation of surface soil moisture retrievals.

Improved root zone soil moisture!

Different precipitation forcing inputs



Precipitation corrections v. retrieval assimilation



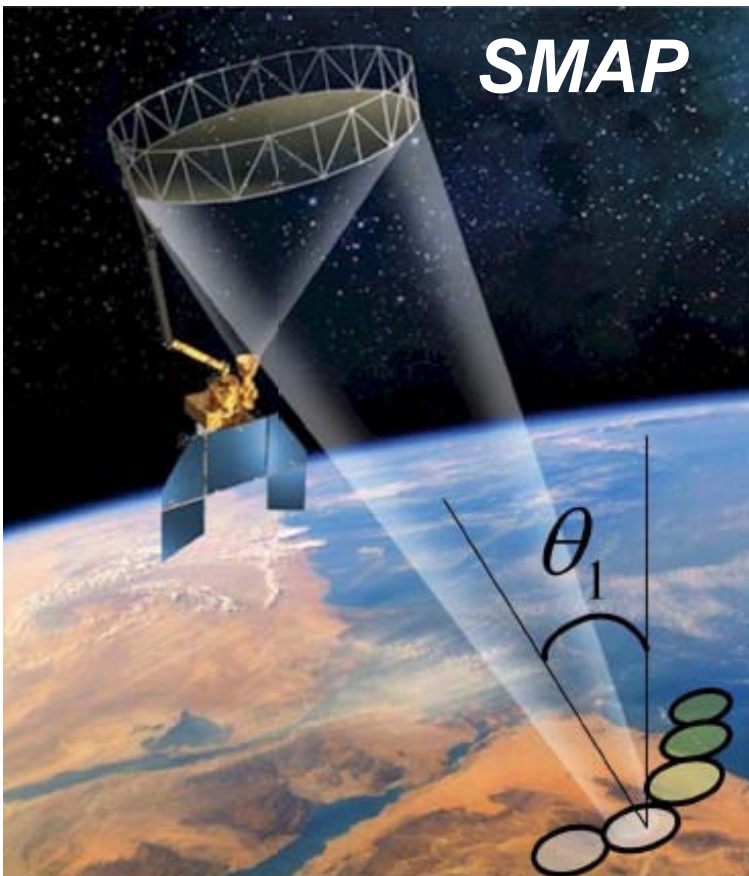
Precipitation corrections and retrieval assimilation contribute approximately:

- evenly and
- independently to skill improvement.

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



Soil moisture assimilation



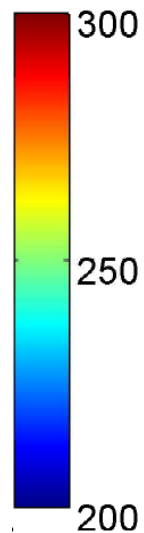
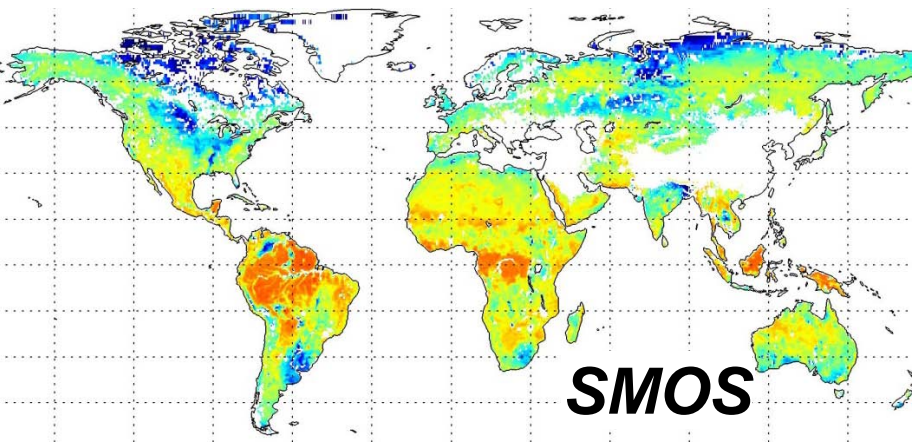
For the SMAP L4_SM product use **brightness temperature (radiance) assimilation**.

Need **L-band** radiative transfer model (RTM).



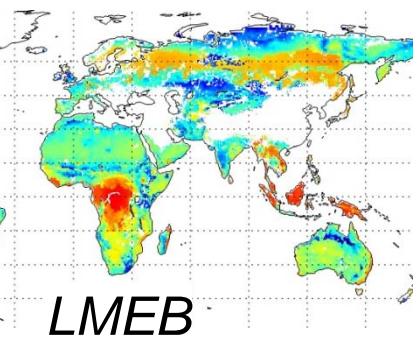
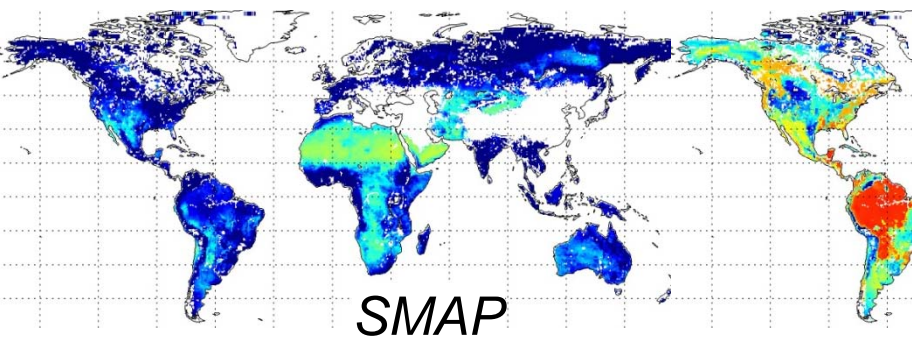
L-band brightness temp.: SMOS vs. Catchment/RTM

Annual mean [K]

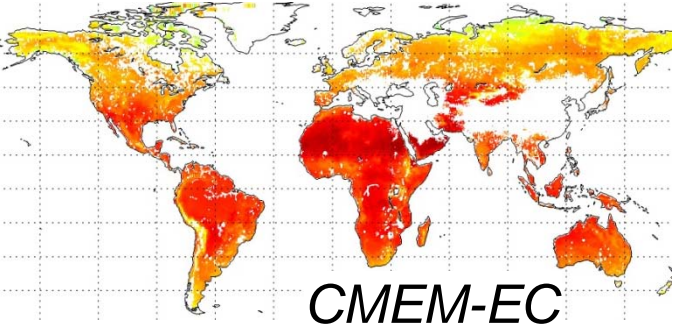


H-pol 1/1/2011 – 1/1/2012
42.5° (validation period)

Literature values for parameters yield strongly biased Tb.



Model

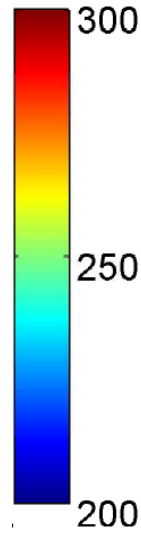
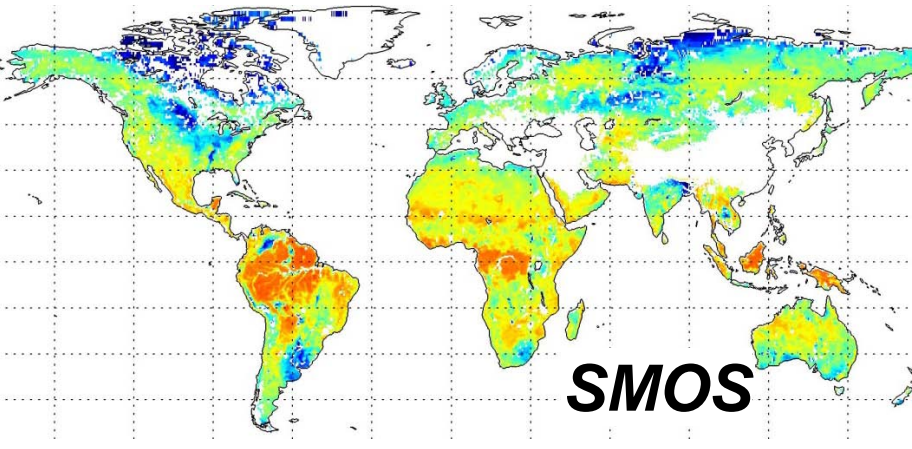


RTM parameters Prescribed:
 SMAP Level2 ATBD
 LMEB literature
 SMOS-monitoring at ECMWF (CMEM-EC)



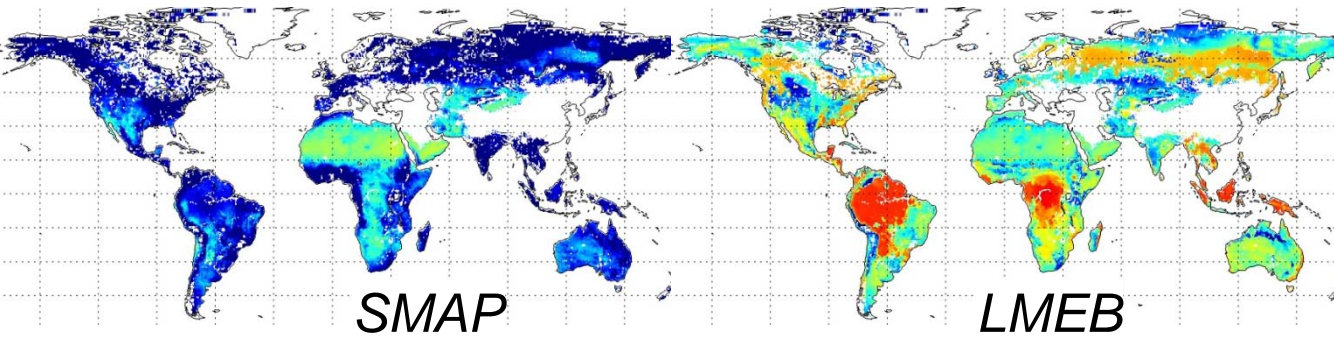
L-band brightness temp.: SMOS vs. Catchment/RTM

Annual mean [K]

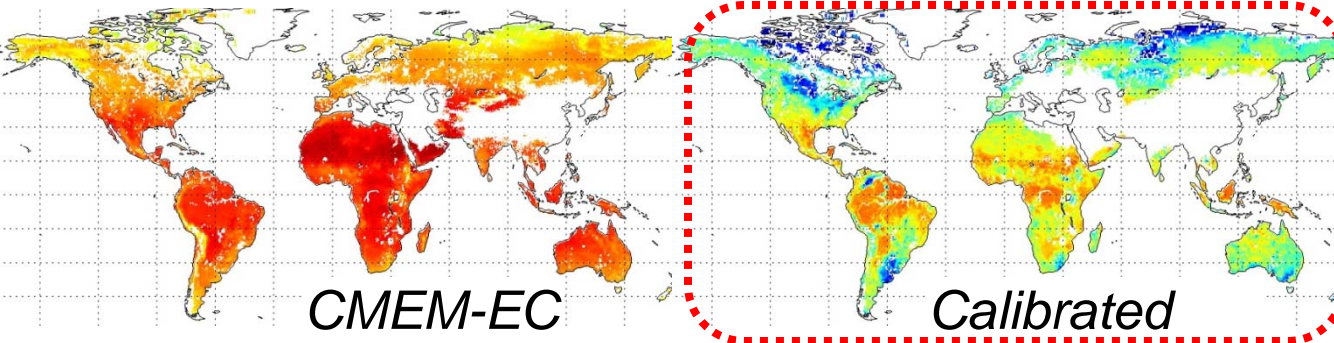


H-pol 42.5°
1/1/2011 – 1/1/2012
(validation period)

Calibrated parameters
yield mostly unbiased
long-term mean Tb.



Model



RTM parameters

Prescribed:
SMAP Level2 ATBD
LMEB literature
SMOS-monitoring at
ECMWF (CMEM-EC)

Calibrated:
From multi-angular
calibration during
1/1/2010 – 1/1/2011



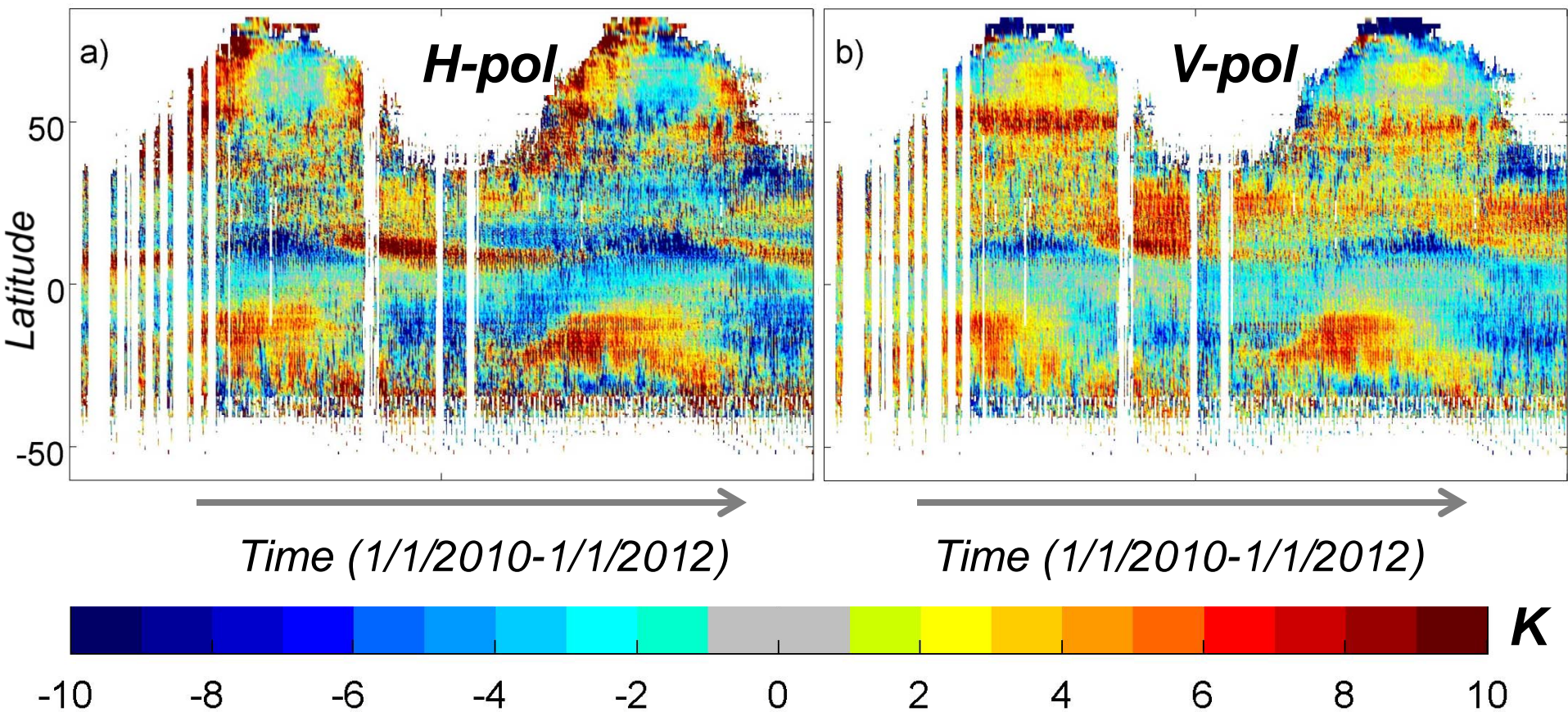
L-band brightness temp.: SMOS vs. Catchment/RTM

Seasonally varying, residual biases after calibration.

Need to address in assimilation system.

RFI in V-pol from Distant-Early-Warning (DEW) Line?

Suppressed in H-pol through calibration?



Bias \equiv SMOS minus Model, average over 6 angles, ascending only



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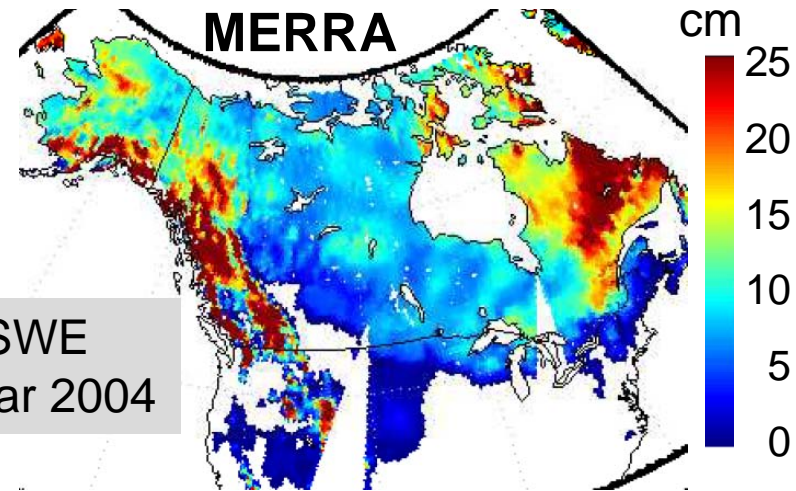
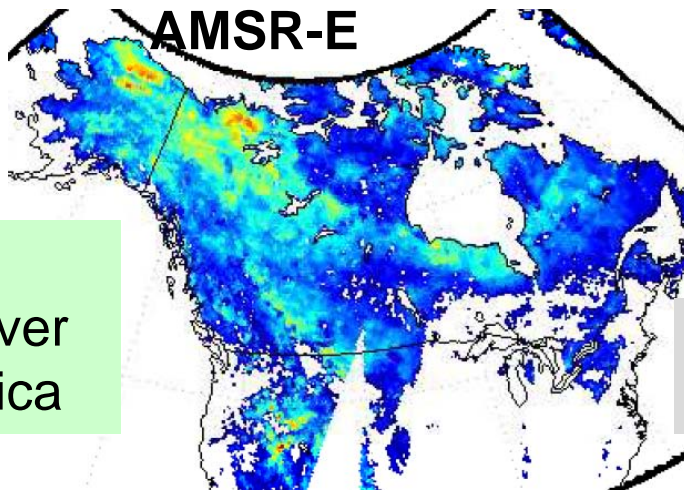
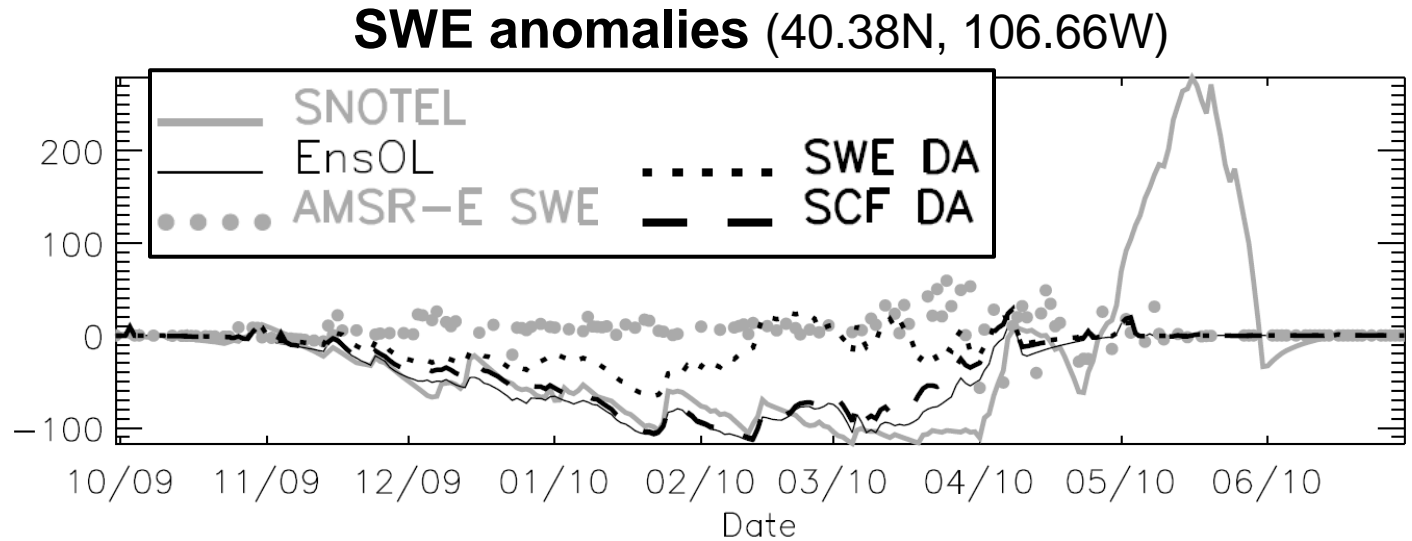
- Soil moisture
- Snow



Assimilate AMSR-E snow water equivalent (SWE)?

Example 1:
Anomaly time series at SNOTEL site.

De Lannoy et al. WRR (2012)



Example 2:
Snapshot over North America

Satellite SWE retrievals not (yet) suitable for assimilation.
Radiance assimilation?



GEOS-5 vs. AMSR-E brightness temp. (18 GHz, V-pol)

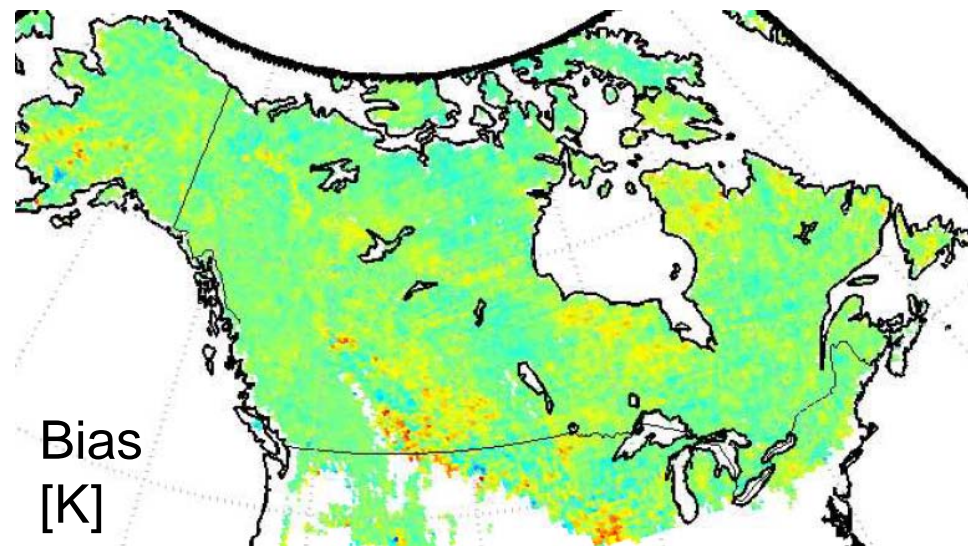
*Artificial Neural Network for
GEOS-5 global snow model*

Input: Snow water equivalent, density,
liquid water content, snow/air/soil
temperature)

Output: T_b (H/V-pol, 10/18/36 GHz)

*Robust forward modeling of
AMSR-E T_b using GEOS-5.*

Forman et al. (2012), IEEE/TGARS, submitted.



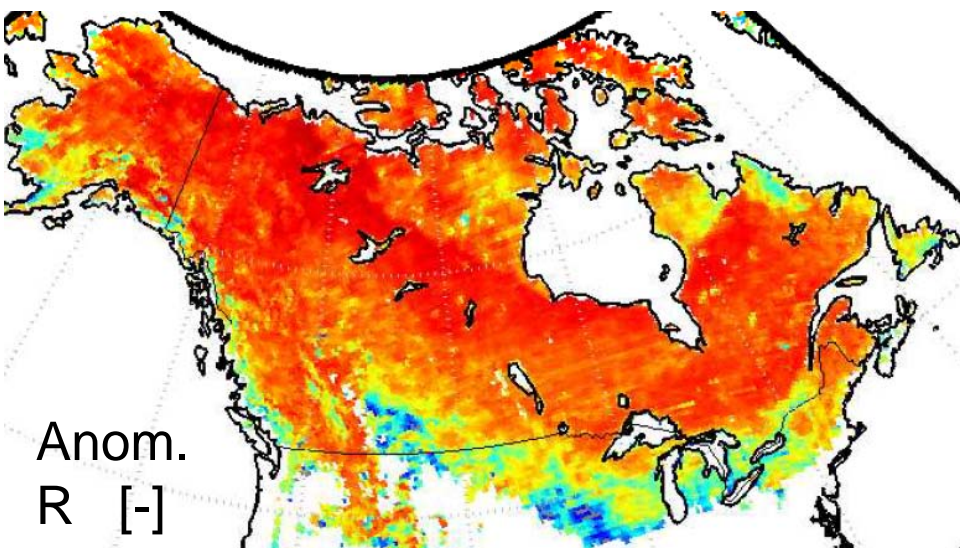
Bias
[K]



-5

0

5



Anom.
R [-]



0

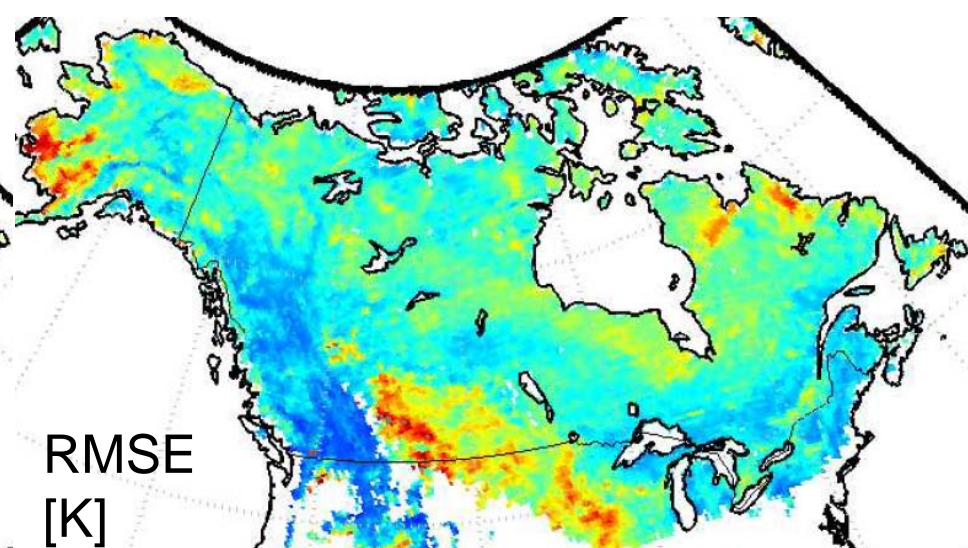
0.2

0.4

0.6

0.8

1



RMSE
[K]



0

5

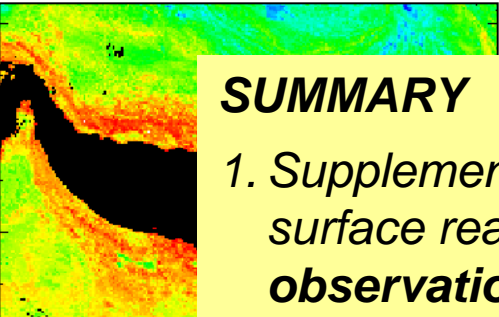
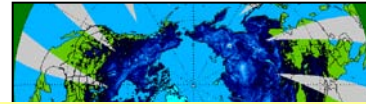
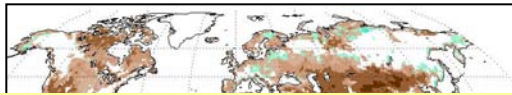
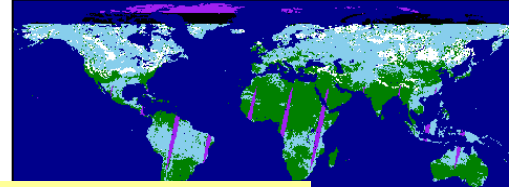
10

15

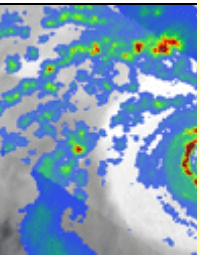
20



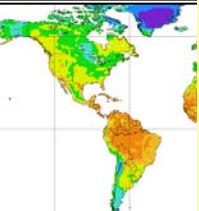
Satellite observations



Land surface temperature (MODIS, AVHRR)



Precipitation (TRMM)



Radiation flux (CERES)

SUMMARY

1. Supplemental **MERRA-Land** data product provides enhanced land surface reanalysis estimates (through use of **precipitation observations** and **land model improvements**).
2. Retrieval-based **soil moisture assimilation** can further improve reanalysis. Radiance-based soil moisture analysis for SMOS and SMAP requires careful calibration of radiative transfer model.
3. **Snow water equivalent** retrievals not (yet) suitable for assimilation. Developed neural network-based forward operator for radiance assimilation.
4. Other topics: Assimilation of **skin temperature**, **snow cover**, and **terrestrial water storage** not discussed in this presentation.

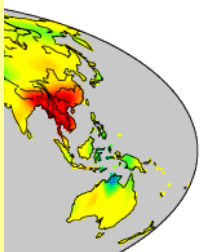
OUTLOOK

1. Focus has been on **univariate, off-line** assimilation. Need **multivariate** analysis of soil moisture, LST, snow cover, and snow water equivalent.
2. We are integrating land and atmospheric assimilation to allow feedbacks in **coupled** land-atmosphere analysis system.
3. Assimilate satellite-based **vegetation/carbon** observations.

Carbon (MIRS)



Sea level elevation (TOPEX/Poseidon)



Gravity field (GRACE)



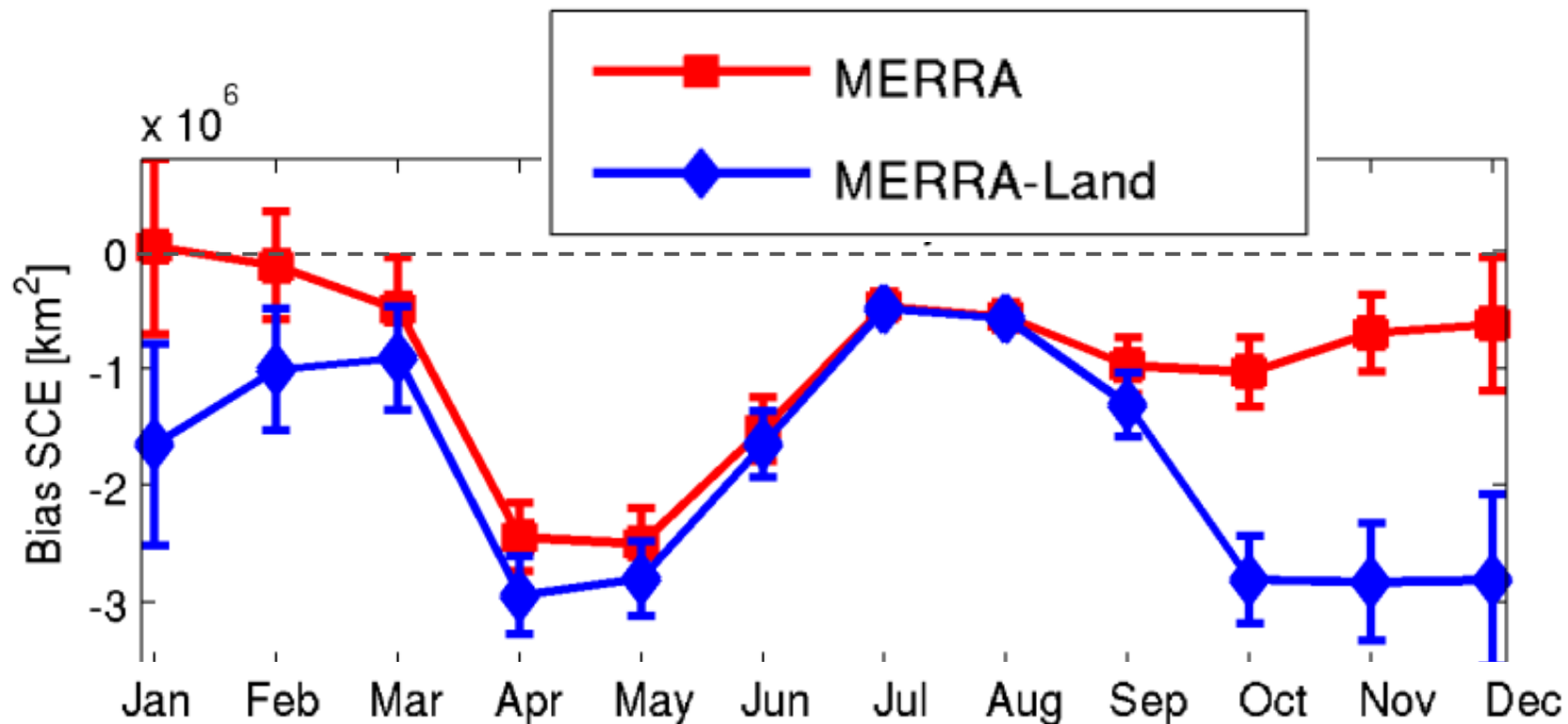
Carbon (MODIS, DESDynI, GPP, VIRI, LIST, etc.)



THANK YOU FOR YOUR ATTENTION!



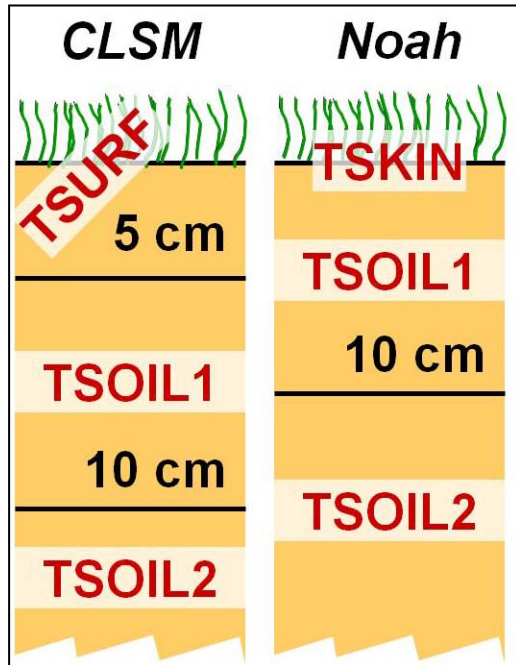
Snow cover extent (SCE) v. MODIS



Larger bias in MERRA-Land snow cover extent vs. MODIS (compared to same in MERRA) due to parameter change in snow model (WEMIN).



Land surface temperature (LST) assimilation



Assimilate ISCCP LST retrievals into two off-line land models.

Validate against in situ obs.

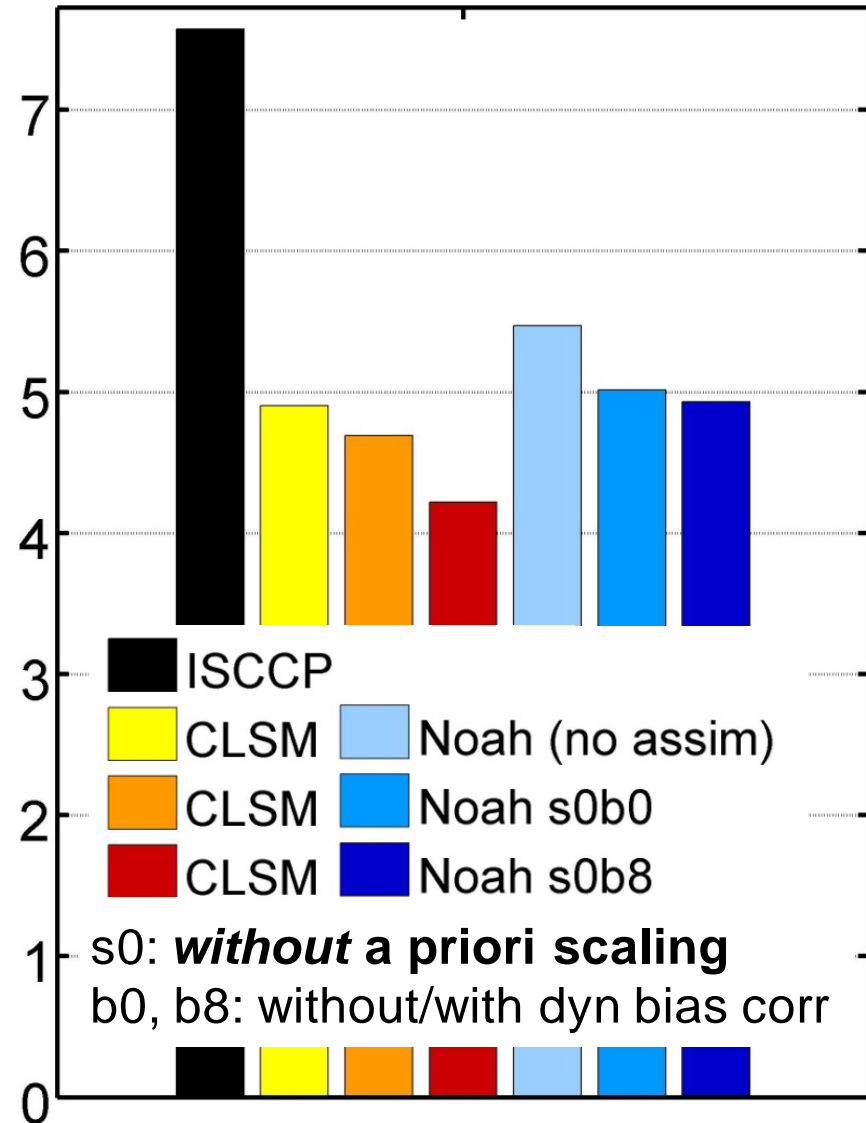
“Model” LST better than ISCCP.

Assimilation reduces RMSE (by up to ~0.7 K), increases anomaly R (by up to 0.05).

Model formulation impacts assimilation: Dynamic bias correction key for CLSM.

Fluxes can be MUCH worse if bias is not addressed (not shown).

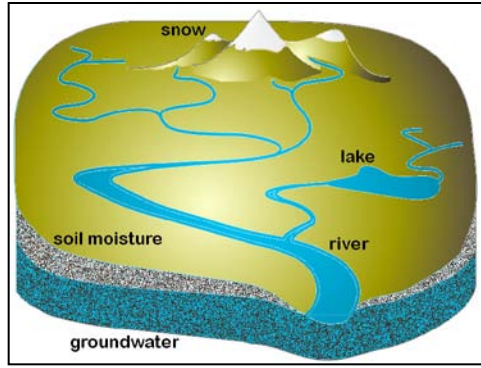
RMSE of raw LST [K]





Assimilation of GRACE terrestrial water storage (TWS)

GRACE measures
large-scale TWS
= groundwater
+ soil moisture
+ snow
+ surface water

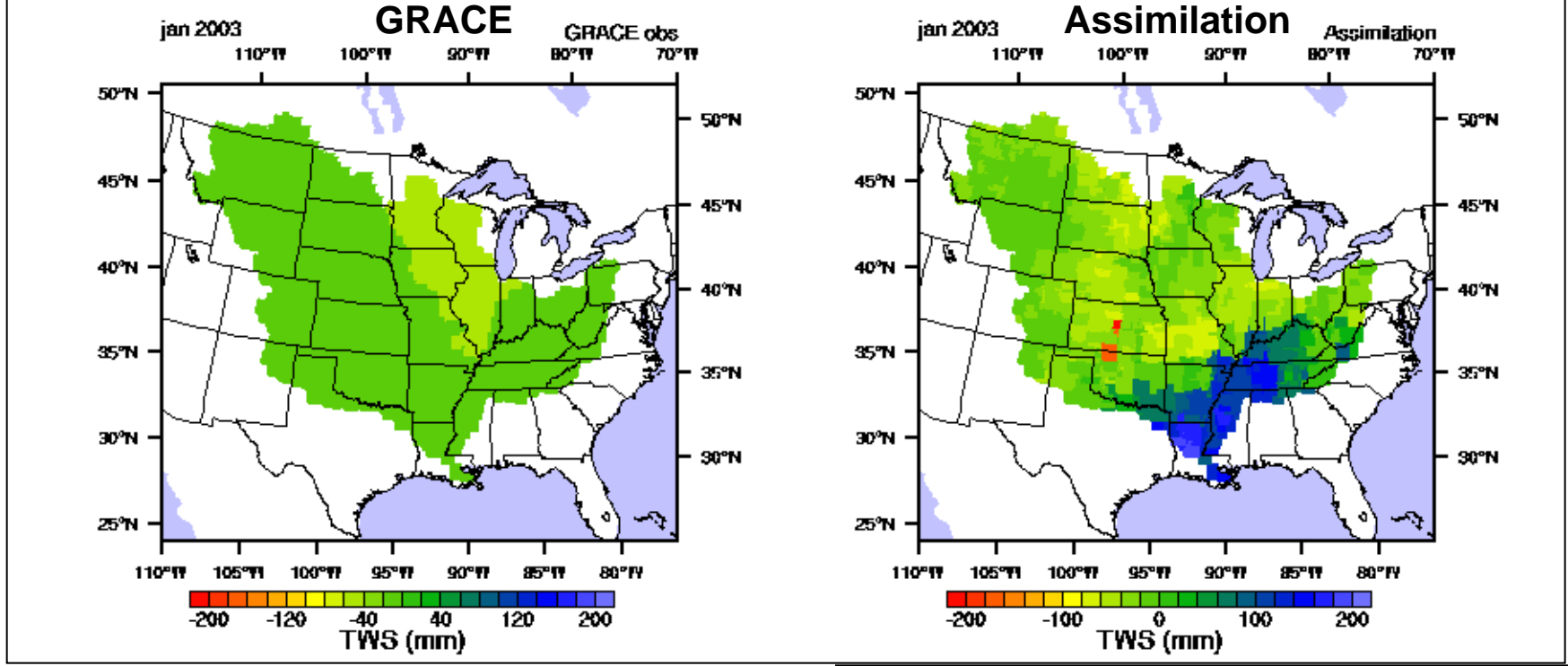


Assimilation yields:

- fine-scale information subject to GRACE basin-scale constraints
- better runoff than model (not shown).

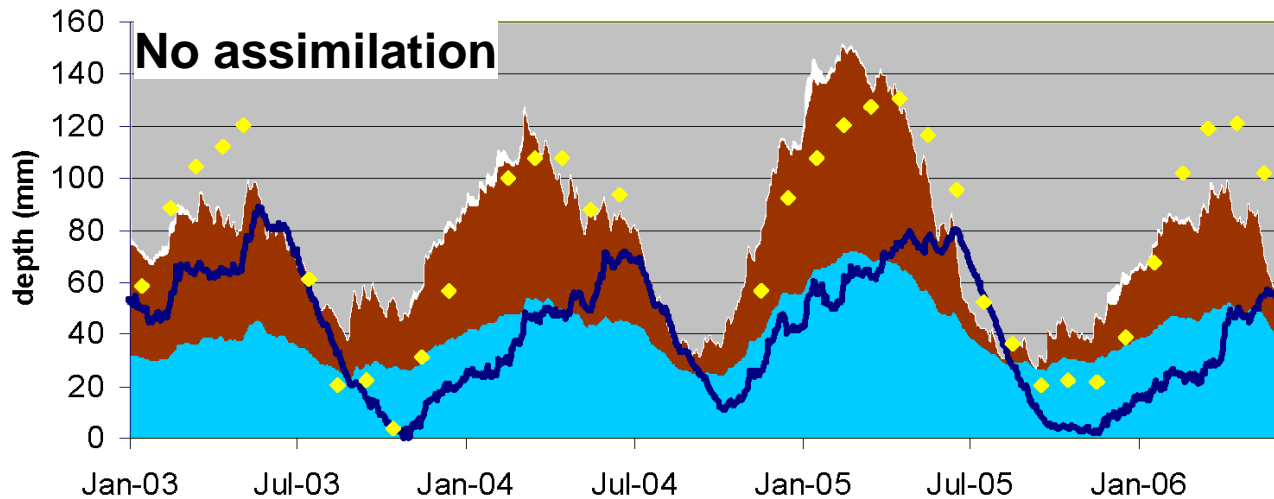


Terrestrial water storage anomaly (Jan. 2003 – Jun. 2006 loop)





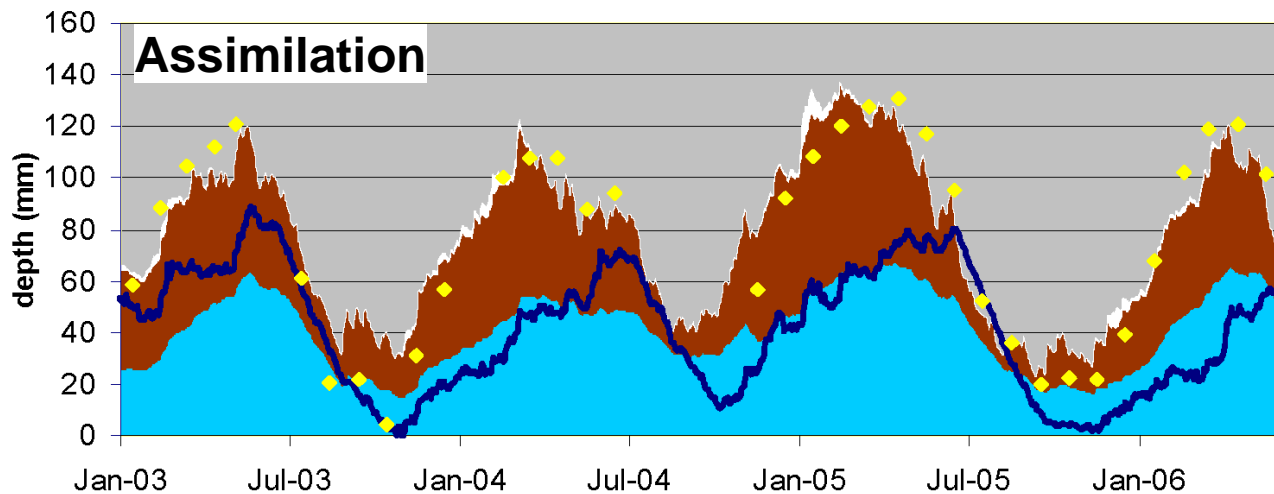
Assimilation of GRACE terrestrial water storage (TWS)



Validation against
observed
groundwater:

RMSE = 23.5 mm

$R^2 = 0.35$



RMSE = 18.5 mm

$R^2 = 0.49$

Groundwater Soil Moisture Snow Water Equivalent
GRACE Total Water Observed Groundwater

Assimilation disaggregates GRACE data into snow, soil moisture, and groundwater.
Assimilation estimates of groundwater better than model estimates.