

Implementation and Application of the Kalman Filter Data Assimilation Approaches in NASA's Land Information System Infrastructure

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Motivation

Many land surface **satellite data products** have been or will be available from NASA, NOAA, ESA, & JAXA satellites (e.g., NDVI/EVI, LAI, LST, Snow, & SM from MODIS, AMSR etc);

Many land surface data assimilation algorithms such as the **Kalman filters** have been published in the literature;

Based the successful development of NLDAS & GLDAS, the **Land Information System (LIS)** could become the best test bed for applying the DA algorithms to assimilate the available satellite data products into land surface models for operational numerical weather predictions in addition to its other applications;

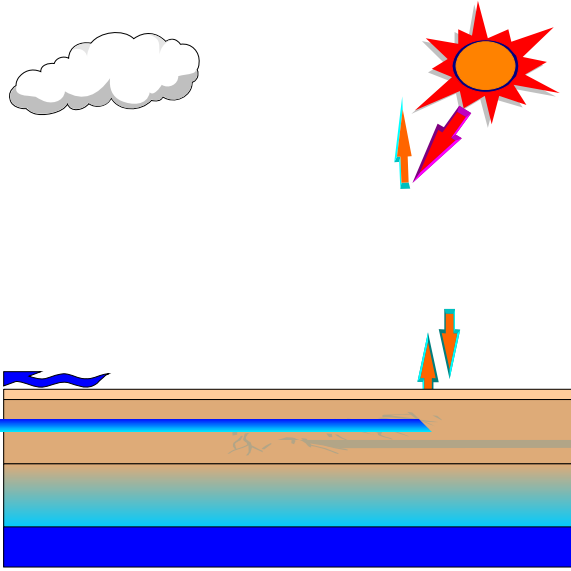
Therefore, **implementing and evaluating** the Kalman filter DA assimilation algorithms in LIS be

**Topography,
Soils**

**Land Cover,
Vegetation
Properties**

Meteorology

**Snow
Soil Moisture
Temperature**



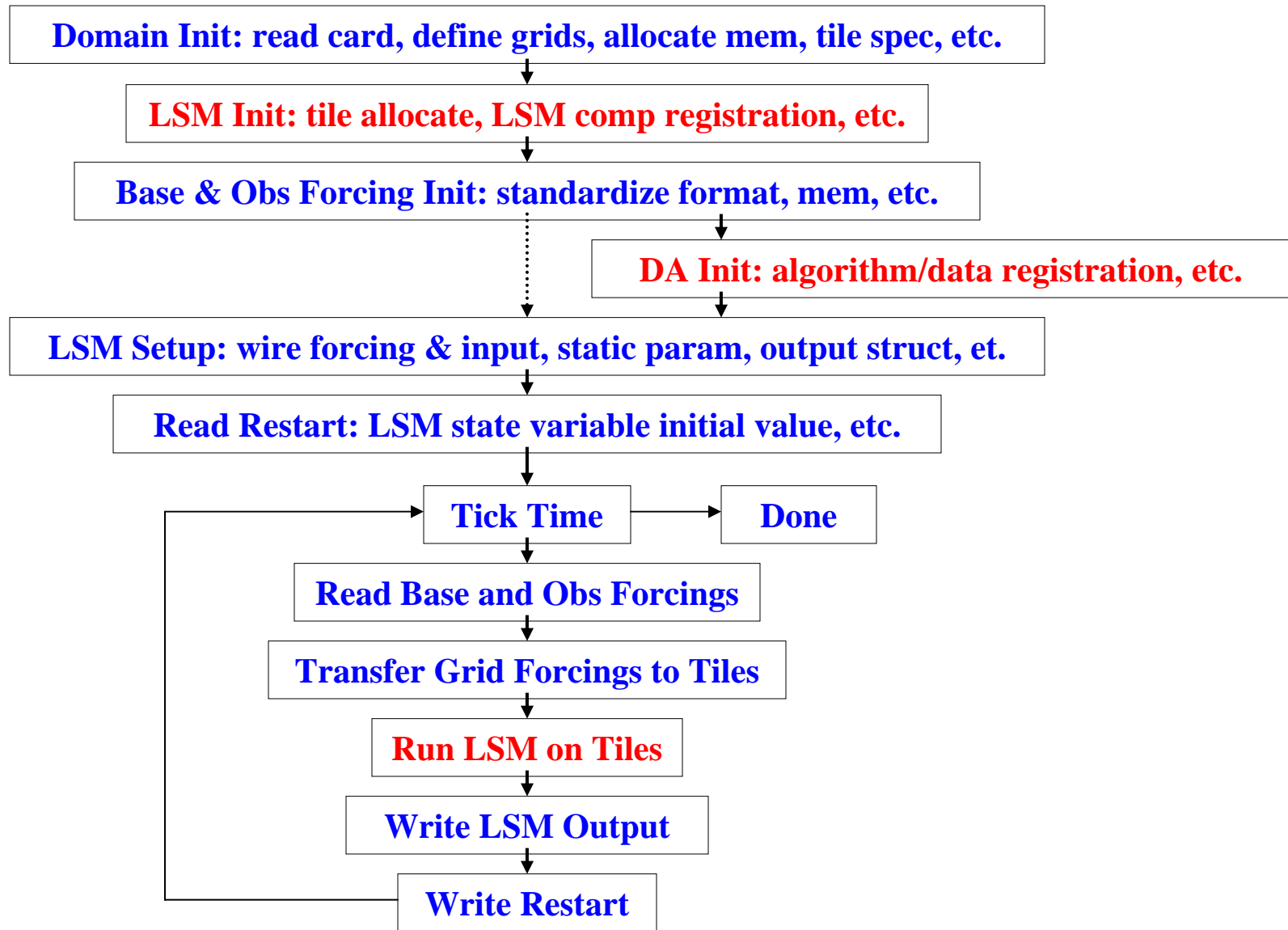
Land Information System (LIS)

LIS is a NASA “software of year” **award winning software** system to enhance the computational capability of the Land Data Assimilation Systems (NLDAS and GLDAS);

LIS coding is based on the Fortran subroutines and functions of the original LDAS Systems, but is restructured using new technologies such as **object-oriented programming**, parallel and low cost cluster computing so that LIS can run land surface models for any size of domain (global or any region), any spatial resolution (0.25 degree, 1km, 30m or any other);

The O-O programming technology allows **adding new feathers** (such as **data assimilation**) to LIS as new plug-ins without major modifications to the original LIS structure and code.

LIS Flow Chart and Data Assimilation Implementation



```
LIS main core/lisdrv.F90:
```

```
call LIS_config
call LIS_domain_init
call LIS_lsm_init
call LIS_baseforcing_init
call LIS_obsforcing_init
call LIS_dataassim_init
call LIS_setuplsm
call LIS_readrestart
do while (.NOT. LIS_endofrun())
  call LIS_ticktime
  call LIS_setDynlsm
  call LIS_get_base_forcing
  call LIS_get_obs_forcing
  call LIS_force2tile
  call LIS_lsm_main
  call LIS_lsm_output
  call LIS_writerestart
enddo
```

```
LIS_lsm_init in plugins/lsm_pluginMod.F90:
```

```
...
call registergetstatevar(1,1,noah_getSMStVar)
call registersetstatevar(1,1,noah_setSMStVar)
call registerobstransform(1,1,noah_obs2st_vsm)
call registerlsmreset(1,noah_reset)
call registerlsmptbtf(1,noah_ptbtf)
call registerlsmptbts(1,noah_ptbts)
call registerqa_check(1,1,noah_qa_sm)
...
call registergetstatevar(4,1,mos_getSMStVar)
call registersetstatevar(4,1,mos_setSMStVar)
call registerobstransform(4,1,mos_obs2st_vsm)
call registerreset(4,mos_reset)
call registerlsmptbtf(4,mos_ptbtf)
call registerlsmptbts(4,mos_ptbts)
call registerlsmqa_check(4,1,mos_qa_sm)
...
```

LIS-DA Code for LSM-Specific Subroutine Registration

These calls are to register the subroutines for the specific LSM and observed state variable. The code for these subroutines are deposited in the subdirectory corresponding to the LSM. For example, the subroutine [noah_ptbtf](#) is [noah_ptbtf.F90](#) in [lsms/noah.2.6](#).

The called registerxxx functions are defined in [core/LIS_lsm_FTable.c](#)

```
LIS main core/lisdrv.F90:
```

```
call LIS_config
call LIS_domain_init
call LIS_lsm_init
call LIS_baseforcing_init
call LIS_obsforcing_init
call LIS_dataassim_init
call LIS_setuplsm
call LIS_readrestart
do while (.NOT. LIS_endofrun())
  call LIS_ticktime
  call LIS_setDynlsm
  call LIS_get_base_forcing
  call LIS_get_obs_forcing
  call LIS_force2tile
  call LIS_lsm_main
  call LIS_lsm_output
  call LIS_writerestart
enddo
```

```
LIS_dataassim_init in core/dataassim_module.F90:
```

```
call dataassim_plugin
call daobs_plugin
call dasetup(lis%a%daalg,lis%a%davar)
call daobssetup(lis%a%daobs)
```

```
DA plugin in plugins/dataassim_pluginMod.F90:
```

```
call registerdasetup(1,1,sm_di_init)
call registerfrcst(1,1,sm_di_frcst)
call registerupdate(1,1,sm_di_update)
call registerdasetup(2,1,sm_ekf_init)
call registerfrcst(2,1,sm_ekf_frcst)
call registerupdate(2,1,sm_ekf_update)
call registerdasetup(3,1,sm_enkf_init)
call registerfrcst(3,1,sm_enkf_frcst)
call registerupdate(3,1,sm_enkf_update)
```

```
OBS plugin in plugins/dataobs_pluginMod.F90:
```

```
call registerdaobssetup(1,smobs_setup)
call registerdaobsread(1,read_tmism)
call registerdaobssetup(2,smobs_setup)
call registerdaobsread(2,read_synsm)
```

LIS-DA Code for DA Method-Specific Subroutine Registration

The method-specific subs such as `sm_enkf_assim` is in `dataassim/enkf/sm/smenkfdrv_module.F90`. The `dataassim_plugin` and `daobs_plugin` subs are to register the DA and OBS subs. The called `registerxxx` functions are in `core/LIS_dataassim_FTable.c`

```
LIS main core/lisdrv.F90:
```

```
call LIS_config
call LIS_domain_init
call LIS_lsm_init
call LIS_baseforcing_init
call LIS_obsforcing_init
call LIS_dataassim_init
call LIS_setuplsm
call LIS_readrestart
do while (.NOT. LIS_endofrun())
  call LIS_ticktime
  call LIS_setDynlsm
  call LIS_get_base_forcing
  call LIS_get_obs_forcing
  call LIS_force2tile
  call LIS_lsm_main
  call LIS_lsm_output
  call LIS_writerestart
enddo
```

LIS-DA Code for DA Execution Procedure at Each Time Step

```
LIS_lsm_main in core/lsm_module.F90:
```

```
if(lis%daalg.gt.0) then
  call daobsread(lis%daobs)
  call forecast(lis%daalg,lis%davar)
  call update(lis%daalg,lis%davar)
else
  call lsmrun(lis%d%lsm)
endif
```

The variables are read from namelists in lis.crd file:

```
&data_assimilation
LIS%DAALG = 3 ! 0-no,1-di,2-ekf,3-enkf da method
LIS%DAOBS = 2 ! 0-no,1-tmi_sm,2-syn_sm,3-amsre_sm, ...
LIS%DAVAR = 1 ! 0-no,1-sm,2-ts,
LIS%NSTV = 16 ! # of LSM state variables
LIS%NDAV = 4 ! # of prog model vars to be assimilated
LIS%NENS = 5 ! number of ensemble members for EnKF
LIS%RENSEM= 0 ! write ensembler data into files (0-n,1-y)
LIS%fvlfm = "./data/perturb/f_mos.dat" ! forc perturb'ns
LIS%svlfm = "./data/perturb/s_mos.dat" ! stat perturb'ns
/

&soil_moisture_da
smobsdir = "./data/NOAHSoilMoist1_NLDAS" ! data location
nob = 1 ! # of obs vars (e.g. surface sm)
oer = 0.01 ! obs error rate
Mer = 0.03 ! Mdl error rate
/
```


Data Assimilation Algorithms Implemented

Direct Insertion (DI): replace LSM states when corresponding observation data are available:

$$X_a = Z$$

Extended Kalman Filter (EKF): correct LSM states by weighing model forecasts and observations with their error covariance:

$$K = P * H^T / [H * P * H^T + R]$$

$$P = (I - K * H) P$$

$$X_a = X_b + K * [Z - h(X_b)]$$

Ensemble Kalman Filter (EKF): correct LSM states by weighing model forecasts and observations with their error covariance:

$$P_{ym} = \Sigma (X_y - \mu_y) (X_m - \mu_m) / n$$

$$K = P_{ym} / (P_{mm} + R)$$

$$X_a = \Sigma [X_b + K * (Z - X_b)] / n$$

Run Data Assimilation In LIS

&data_assimilation

```
LIS%a%DAALG = 3    ! 0-no,1-di,2-ekf,3-enkf da mehod
LIS%a%DAOBS = 2    ! 0-no,1-tmi_sm,2-syn_sm,3-amsre_sm, ...
LIS%a%DAVAR = 1    ! 0-no,1-sm,2-ts,
LIS%a%NSTV  = 16   ! # of LSM state variables
LIS%a%NDAV  = 4    ! # of prog model vars to be assimilated
LIS%a%NENS  = 5    ! number of ensemble members for EnKF
LIS%a%RENSEM= 0    ! write ensembler data into files (0-n,1-y)
LIS%a%fvlfm = "./data/perturb/f_mos.dat"  ! forc perturb'ns
LIS%a%svlfm = "./data/perturb/s_mos.dat"  ! stat perturb'ns
/
```

&soil_moisture_da

```
smobsdir    = "./data/NOAHSoilMoist1_NLDAS" ! data location
nob         = 1          ! # of obs vars (e.g. surface sm)
oer        = 0.01       ! obs error rate
Mer        = 0.03       ! Mdl error rate
/
```

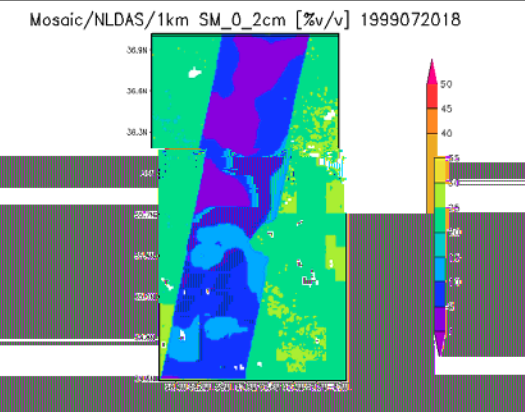
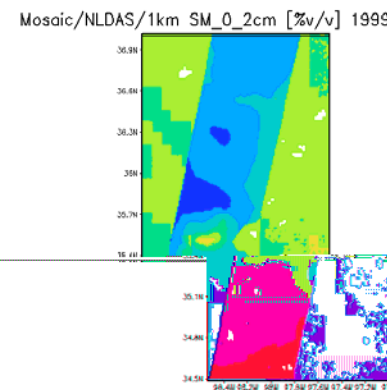
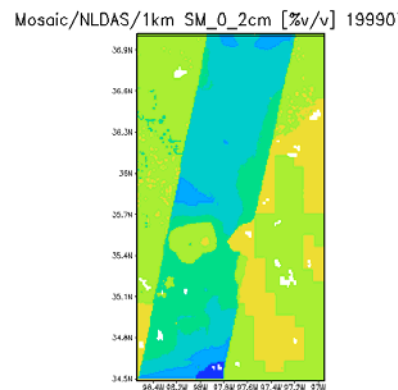
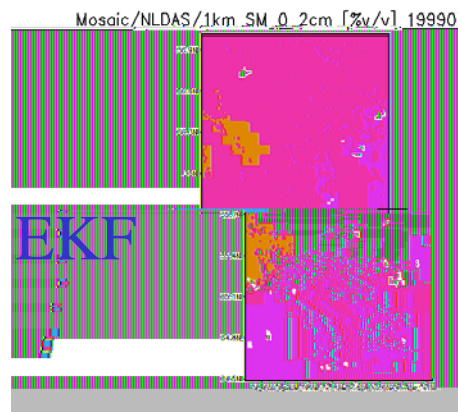
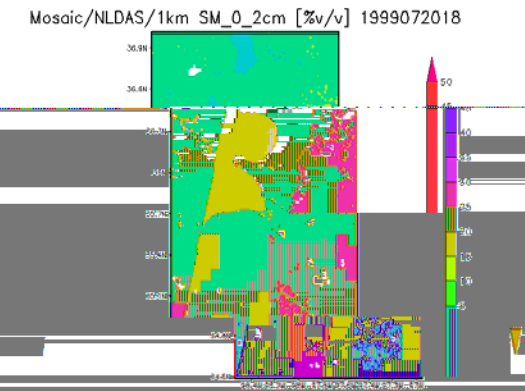
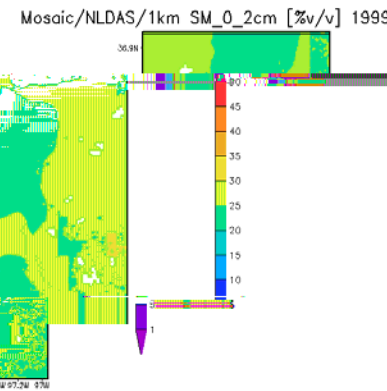
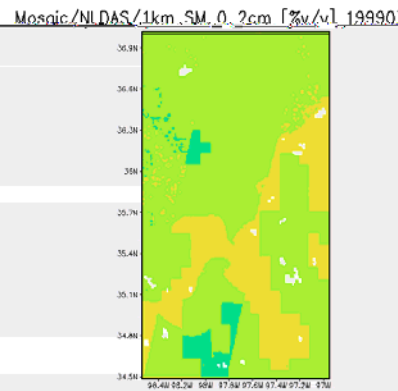
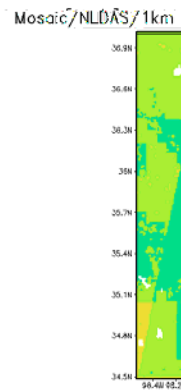
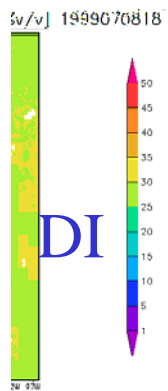
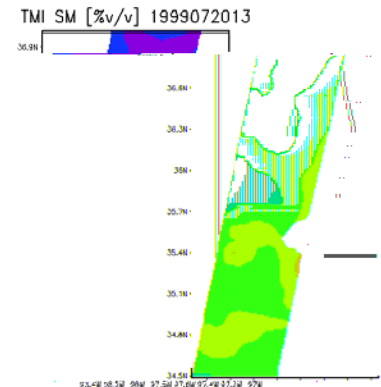
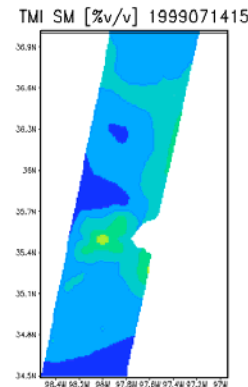
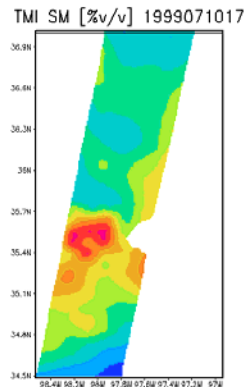
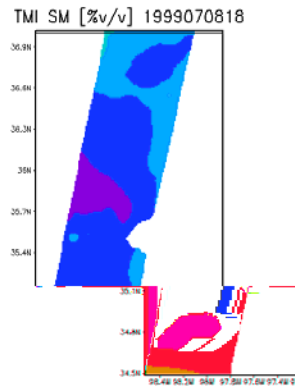
Data Sets for Testing the DA Features in LIS

- 1. SGP'99 TMI observations:** Resampled and calibrated soil moisture retrievals from TRMM Microwave Imager (TMI) for the SGP'99 domain for the days from 8 to 20th of July, 1999.
- 2. Synthetic 0-2cm SM from Mosaic or Noah LSMs:** simulated soil moisture Mosaic and Noah LSM in LIS and the NLDAS atmospheric forcing data set for the days from 18 June to 20 July, 2002;
- 3. NASA's AMSR-E soil moisture data product:** Level 3 soil moisture retrievals since 18 June, 2002

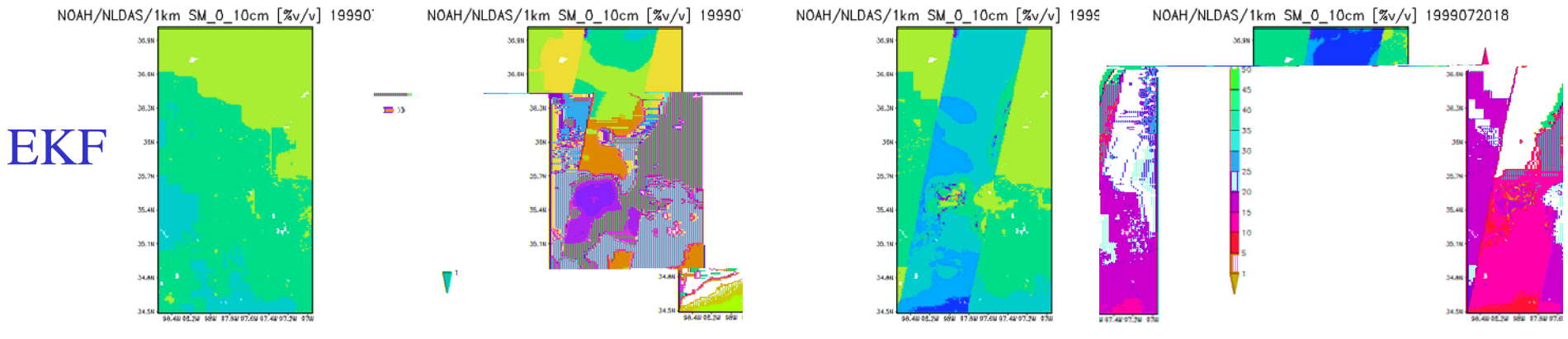
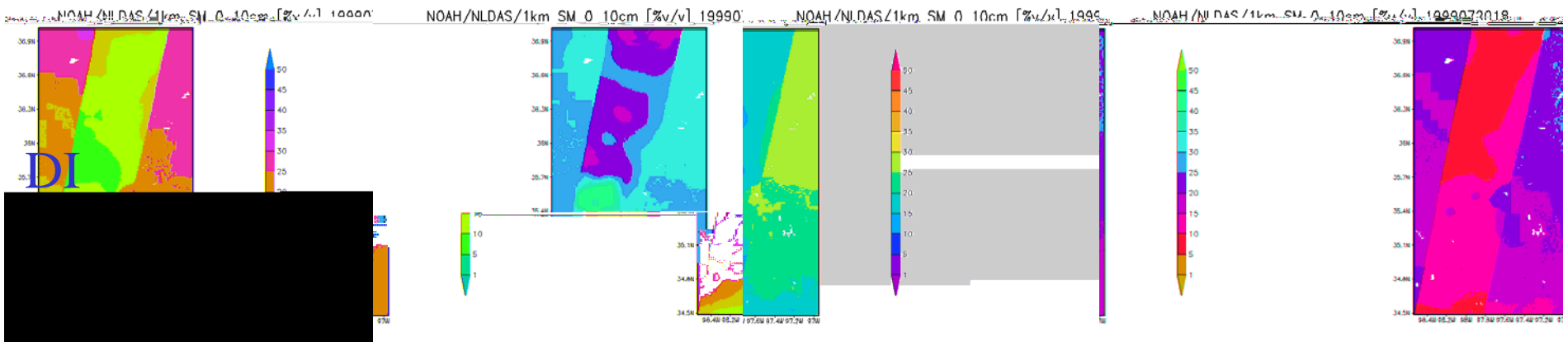
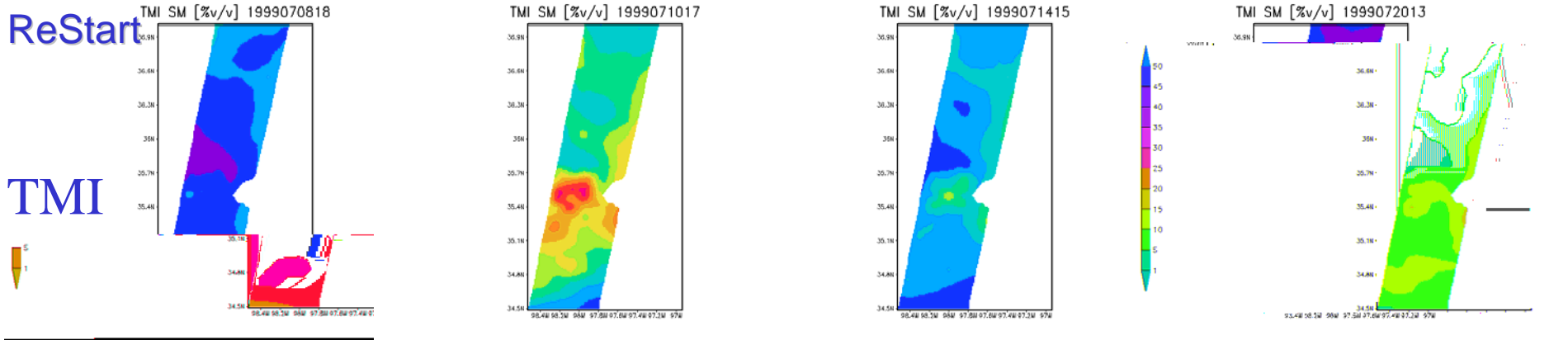
SGP'99 TMI SM Data Assimilation with Mosaic LSM

Wet
Start

TMI

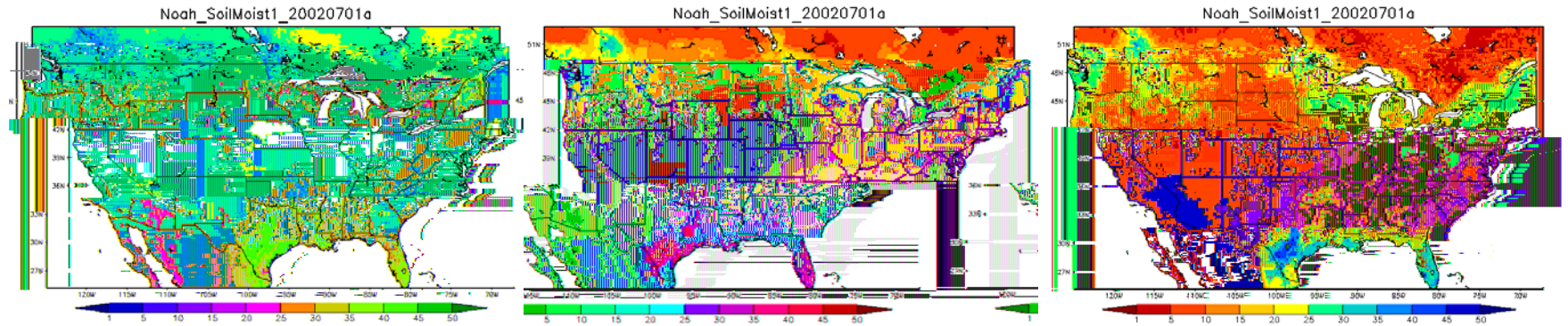


SGP'99 TMI SM Data Assimilation with Noah LSM

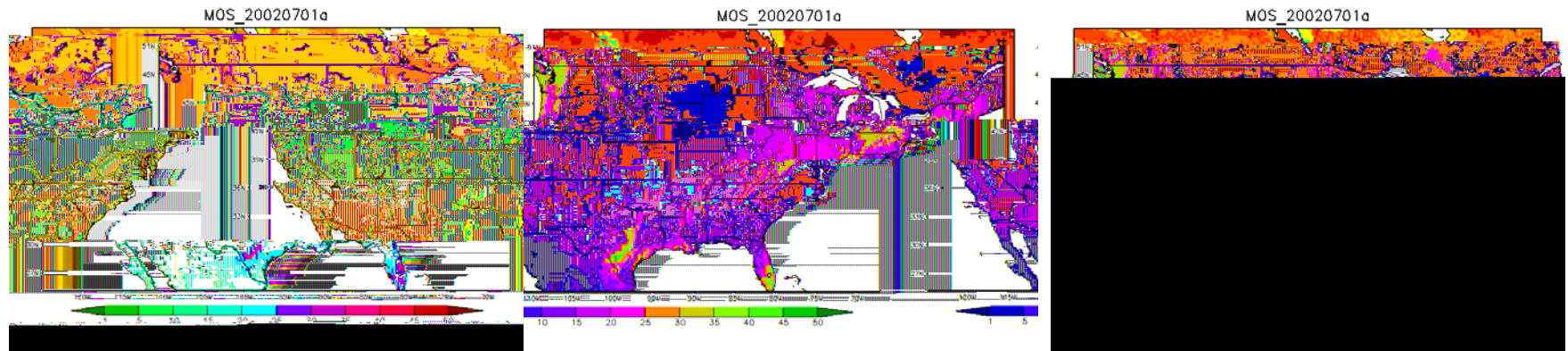


EnKF Assimilation of Synthetic SM Data

Noah



Mosaic



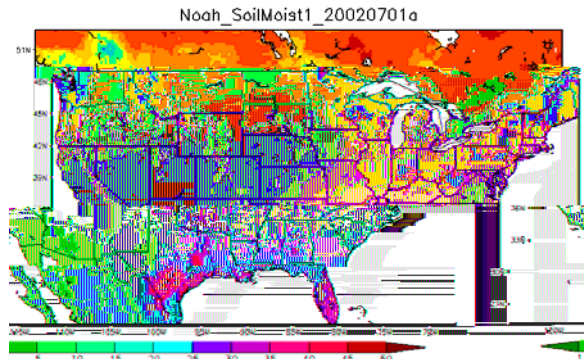
LSM Run

Synthetic Obs

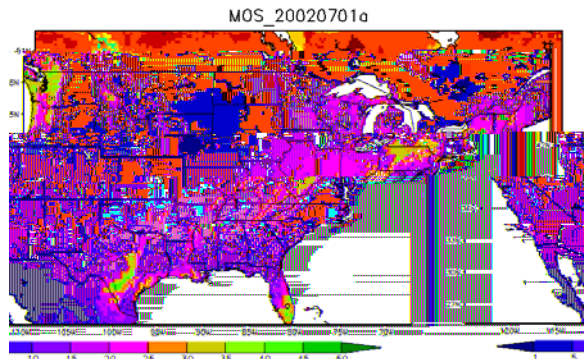
EnKF Assimilation

EnKF Assimilation of AMSR-E SM Retrievals

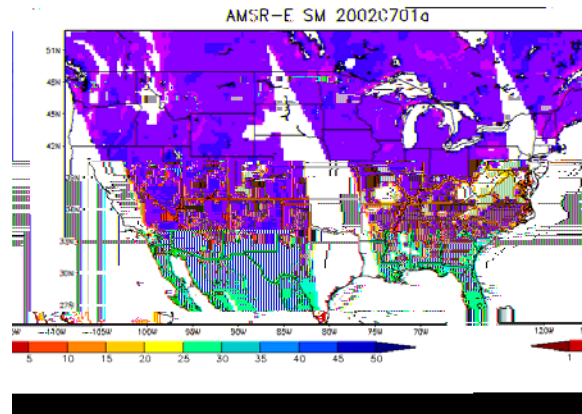
Noah



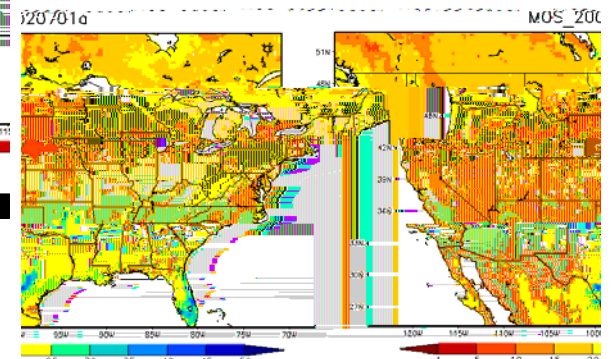
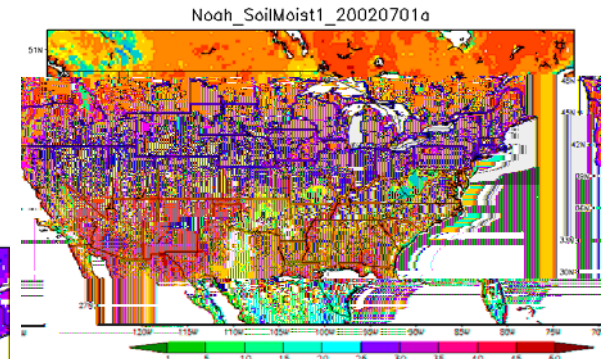
Mosaic



LSM Run



AMSR-E



EnKF Assimilation

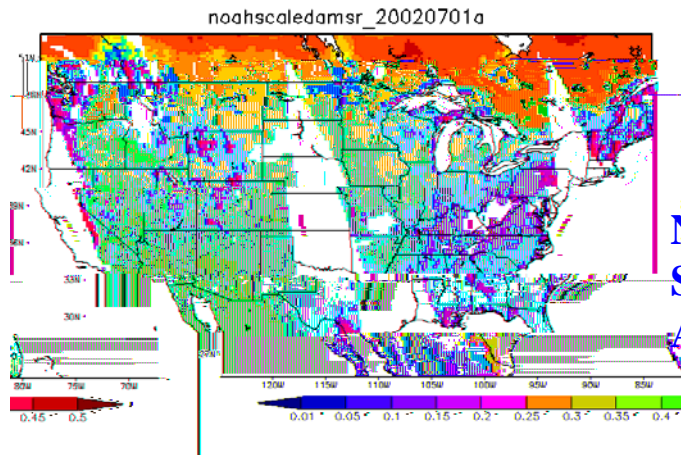
CDF Matching of AMSR-E Retrievals to Model Simulations

Purpose: LSM simulations can have full time and space coverage, but accuracy is uncertain; AMSR-E retrievals are patchy, biased and variations damped. Data assimilation can combine them, but their scales need to be matched for DA methods such as EnKF;

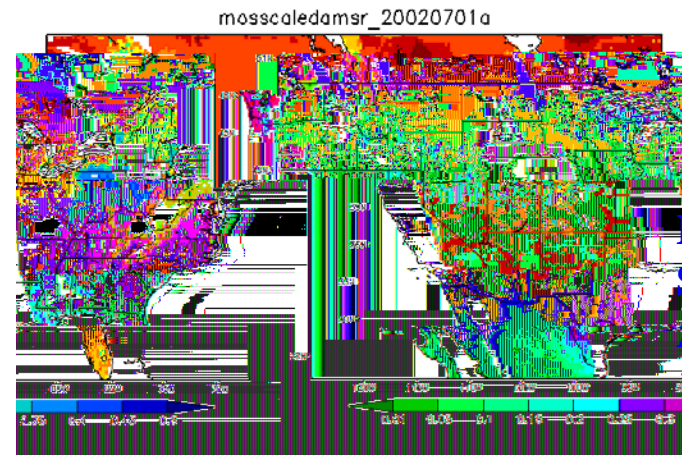
The cumulative distribution function matching method used in Reichle & Koster (2004) can scale the AMSR-E retrievals to the scales of the simulations of LSMs to be used for assimilation;

How will the scaled AMSR-E data be compared with the model simulations?

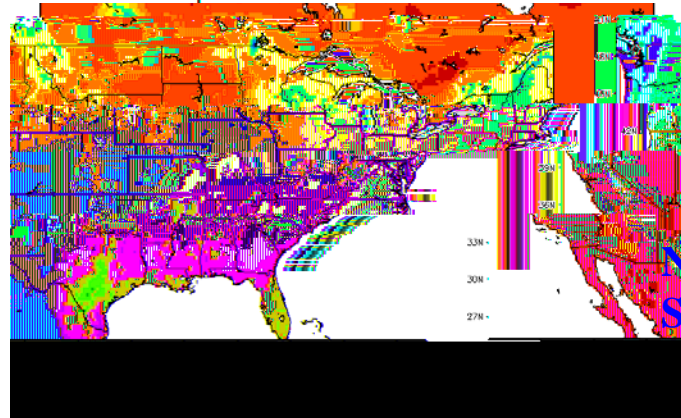
Effect of Scaling AMSR-E to Model Simulations



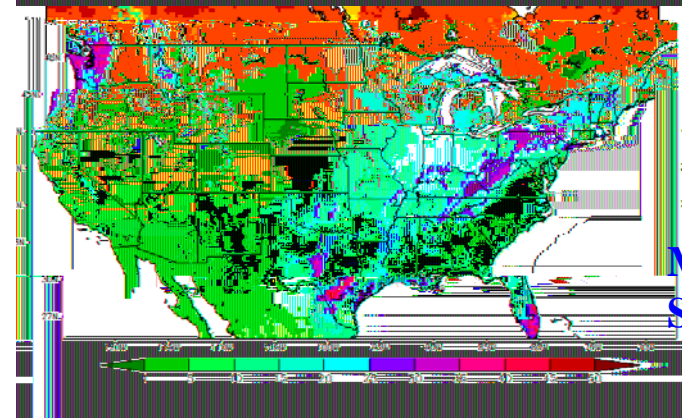
Noah-Scaled
AMSR-E



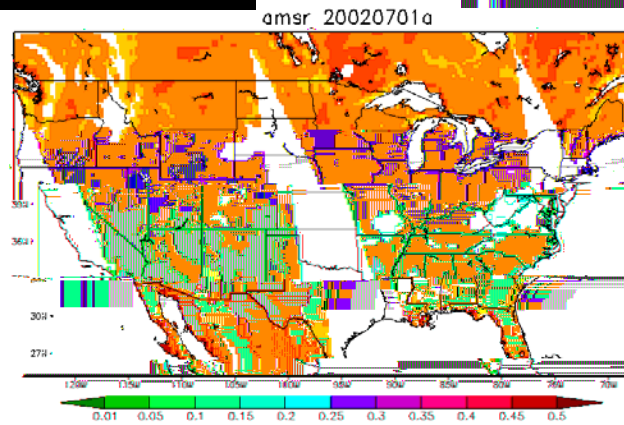
Mosaic-Scaled
AMSR-E



Noah
Simulation

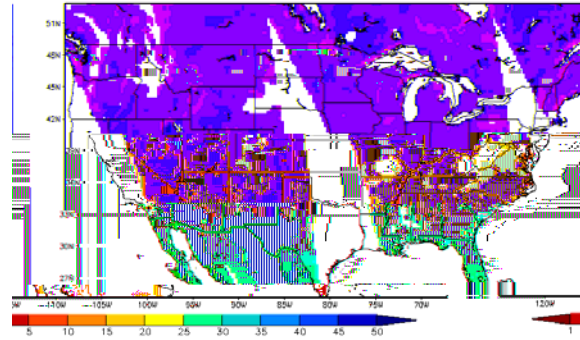


Mosaic
Simulation



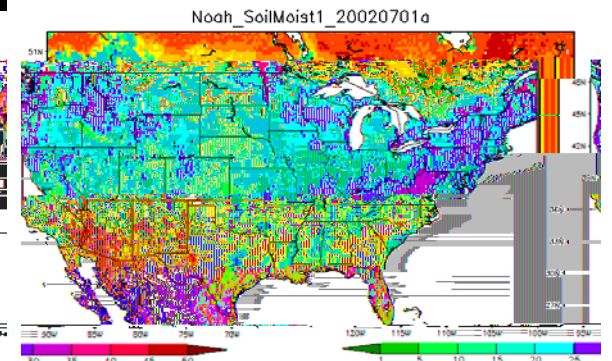
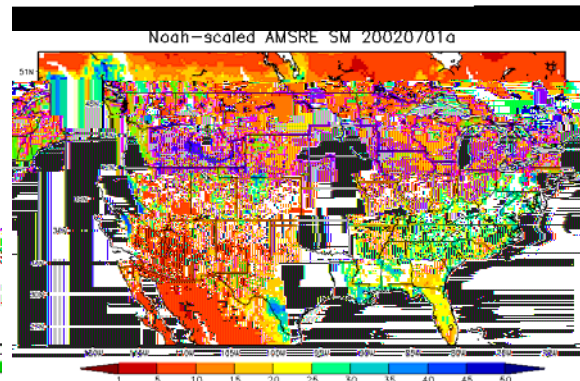
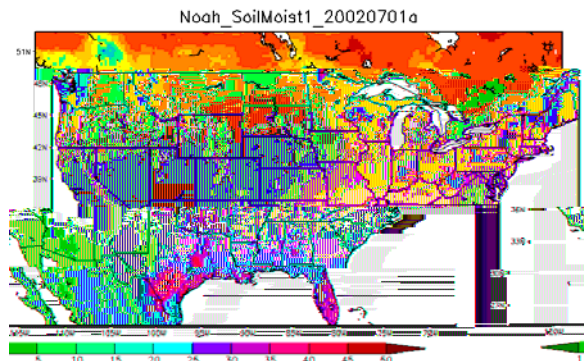
AMSR-E
Retrieval

EnKF Assimilation of Scaled AMSR-E SM Retrievals

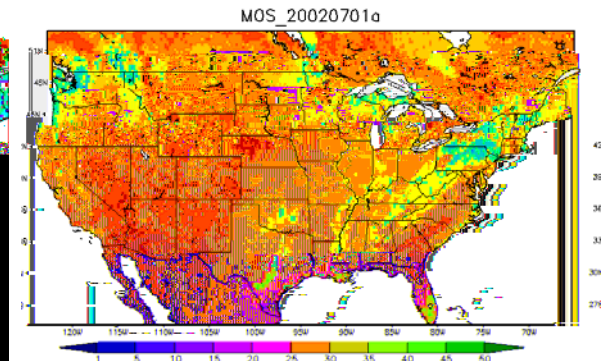
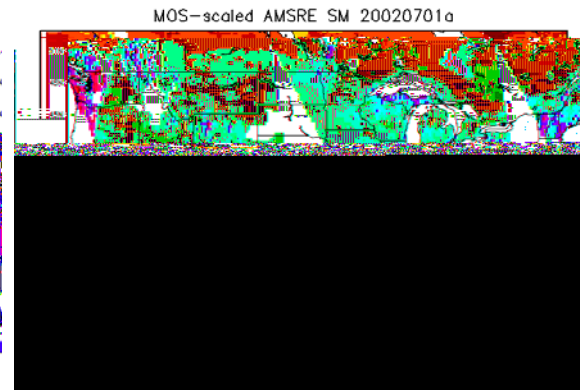
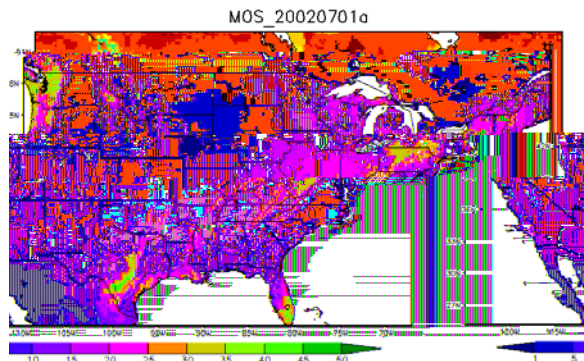


AMSR-E SM

Noah



Mosaic



LSM Run

Scaled AMSR-E

EnKF Assimilation

Ensemble Generation & Model Error Covariance

Model error covariance depends on the magnitude of perturbations added to the ensemble generation. Therefore, the EnKF assimilation result is subject to the perturbation setup;

Dee (1995) and Crow et al. (2005): Standardized innovation

$$[Z-X_b]^T[Z-X_b]/[P_{mm}+R] \sim N(\mu, 1)$$

for the assimilated states to be optimal.

A post processing tool to compute the standardized innovation variance is created for adjusting the perturbation setup manually.

Summary

Three data assimilation algorithms (DI, EKF, EnKF) have been implemented in the Land Information System (LIS) to assimilate soil moisture observations into Mosaic or Noah LSM simulations;

The LIS DA capability has been tested with difference soil moisture observation data sets;

A tool for generating the ensembles of EnKF is created to ensure the assimilated result is optimal;

Further quantitative evaluation is needed and will be made before we can employ the LIS DA function for large scale applications.