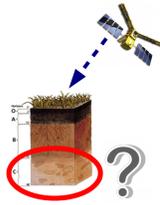


Motivation

As the ocean, the long-term memory of the (deep) soil exhibits potential seasonal-to-decadal predictability.^[1]
→ Soil temperature & moisture should be initialized as realistic as possible.
Challenge: *There are no observations for the deep soil...*



Approach: Ensemble-based data assimilation

- Use offline soil model (TERRA-ML from COSMO-CLM)
- Assimilate atmospheric forcing data and satellite observations
- TERRA-ML transports observational information downwards
- Achieve optimized soil initial state for predictions

Observation Simulation System Experiments

TERRA-ML

- Multilayer land surface scheme, default in COSMO-CLM
- Prognoses soil temperature & moisture
- Offline model, driven by atmospheric forcing

Data assimilation experiment

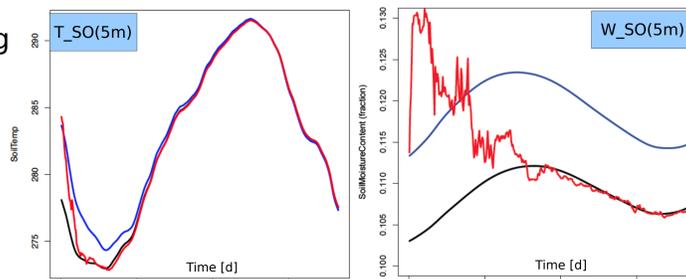
- One vertical soil column (Lindenberg)
- Use Ensemble Kalman Filter
- Generate truth run, but initialize filter with strong bias (initial uncertainty)

Observations

Soil temperature & moisture in first layer only

Results

Compare **filter analysis** with **spin-up** (forcing only) and **truth** over one year:



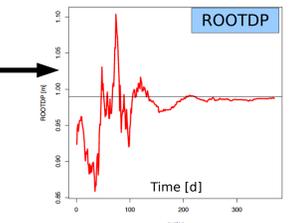
Evolution of truth, analysis and spin-up in 5m depth for soil temperature (T_SO) and moisture (W_SO) for one year.

State estimation

One observation per day already valuable:
→ Soil temperature & moisture are estimated more reliably and quicker than with spin-up

Parameter estimation

True root depth can be retrieved
→ Estimation of uncertain parameters works in principle



→ Deep soil estimation by ensemble data assimilation is feasible.

Regional Data Assimilation with Real Observations

Technical setup

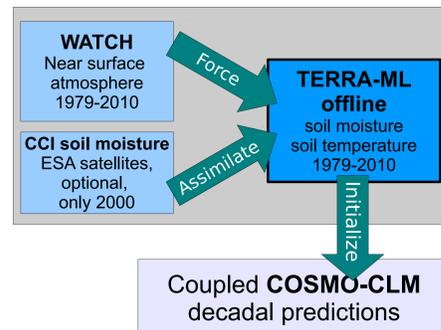
- Domain: Africa, decade 2001-2010, within the *Miklip-DEPARTURE* project
- TERRA-ML coupled with **Parallel Data Assimilation Framework** (PDAF^[2])
- Use **Local Ensemble Transform Kalman filter** (LETKF^[3])

Methodology

- Create land surface re-analysis 1979-2010 using WATCH forcing data^[4]
- Additionally assimilate satellite soil moisture in 2000

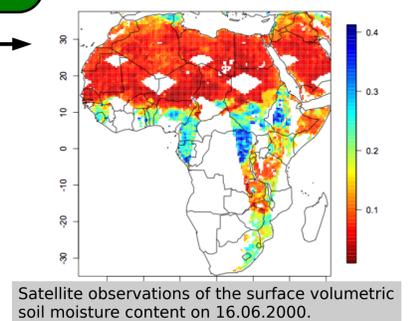
Data assimilation system

Generates soil initial conditions



Satellite observations

- Daily ESA CCI surface soil moisture^[5] in year 2000
- Includes uncertainty information: used as observation error in DA system
- Soil temperature unobserved, but is updated through correlations with soil moisture (e.g., heat capacity and transfer coefficient depend on soil moisture)



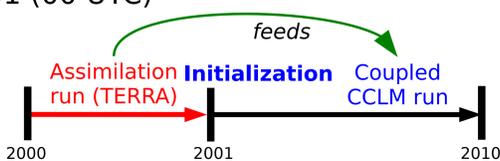
Satellite observations of the surface volumetric soil moisture content on 16.06.2000.

→ Regional, ensemble-based land surface data assimilation system fully working.

Impact on Decadal Predictions

Setup

- Decadal COSMO-CLM hindcast (2001-2010) for Africa, downscales global MPI-ESM
- Initialize with soil fields from DA system at 01.01.2001 (00 UTC)

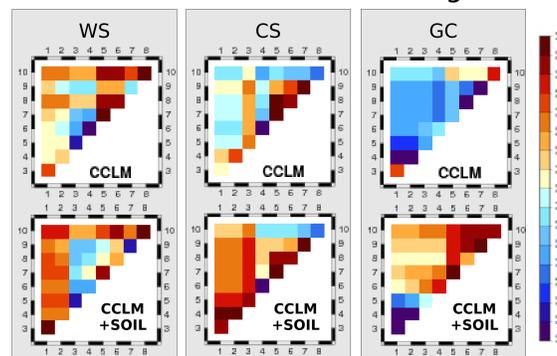


Evaluation

- West African Monsoon (WAM) precipitation in three regions: *Guinea Coast* (GC), *Western & Central Sahel* (WS, CS)
- Compare new soil initialization (**CCLM-SOIL**) with standard initialization (= soil taken from ERA-driven historical COSMO-CLM run)

Results

Correlations with observations for all sub-periods of the decade in the 3 regions:

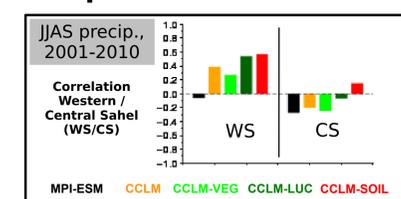


Correlations between observed & simulated JJAS precipitation for all sub-periods in 2001-2010. The upper panel shows the results for the standard CCLM hindcasts, while the lower panel uses new soil initial conditions. X-axis = start year; Y-axis = end year

→ Soil initialization consistently improves the predictability of sub-periods in all regions:

- WS/CS: particularly for longer periods
- GC: particularly at end of decade

Comparison with other boundary conditions



MPI-ESM: global model
CCLM: standard COSMO-CLM
CCLM-VEG: with VEG3D instead of TERRA_ML
CCLM-LUC: with new land use data to generate the BC
CCLM-SOIL: with new soil initial fields

- RCMs outperform the GCM and land surface BCs influence the decadal predictability
- Improved soil initialization increases skill in both regions

Conclusions and Outlook

Conclusions

- Soil initialization is relevant for medium-range climate
- Predictive skill can be enhanced by soil initialization with ensemble data assimilation due to long-term memory
- Limited by model & observations errors, but high potential due to improved satellite observations within past years



Outlook

- ▶ Further evaluation
- ▶ Use new satellite products
- ▶ Application to other predictions/models
- ▶ Exploration of parameter estimation

References

- [1] Bellucci et al. (2015): Advancements in decadal climate predictability: the role of non-oceanic drivers. *Reviews of Geophysics*, 53, 165-202.
- [2] Nerger and Hiller (2013). Software for Ensemble-based Data Assimilation Systems - Implementation Strategies and Scalability. *Comp. & Geosci.*, 55, 110-118.
- [3] Hunt et al. (2007): Efficient data assimilation for spatiotemporal chaos: A local ensemble transform Kalman filter. *Physica D*, 230, 112-126.
- [4] Weedon et al. (2014): The WFDEI meteorological forcing data set: WATCH Forcing Data methodology applied to ERA-Interim reanalysis data. *Water Res. Res.*, 50, 7505-7514.
- [5] Dorigo et al. (2014): Evaluation of the ESA CCI soil moisture product using ground-based observations. *Remote Sensing of Environment*, 162, 380-395