

Forecasting seasonal agricultural droughts in East Africa using satellite based observations, land surface models and dynamical weather/climate forecasts

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Seasonal agricultural drought

Moisture deficit forecasts

Initial
moisture state

Land surface model

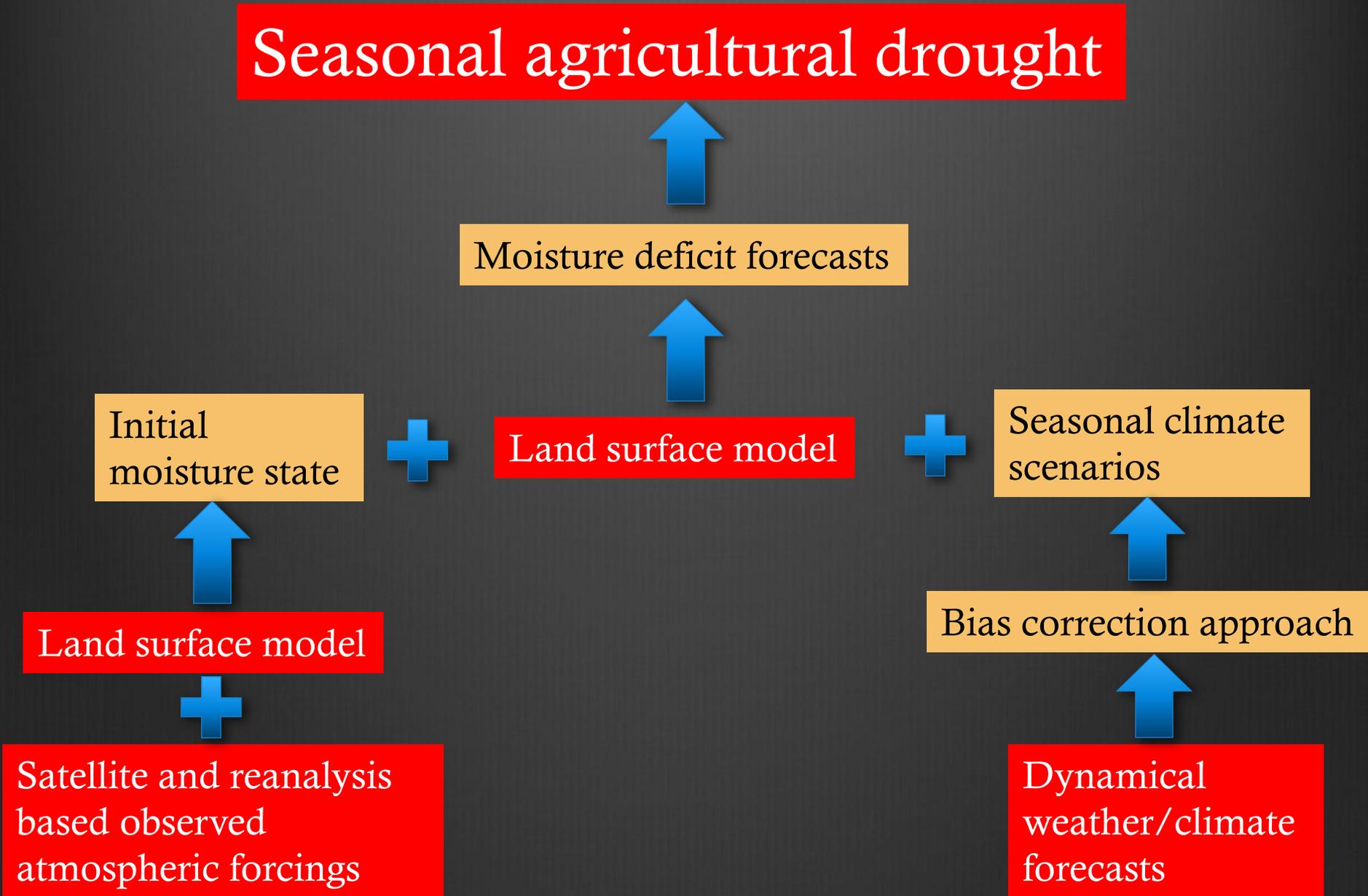
Seasonal climate
scenarios

Land surface model

Bias correction approach

Satellite and reanalysis
based observed
atmospheric forcings

Dynamical
weather/climate
forecasts



Outline

- ⊗ Evaluation of seasonal climate forecasts
- ⊗ Hybrid approaches to improve precipitation forecast skill in the region
- ⊗ Implementation and evaluation of a seasonal agricultural drought monitoring and forecasting system
- ⊗ Concluding remarks

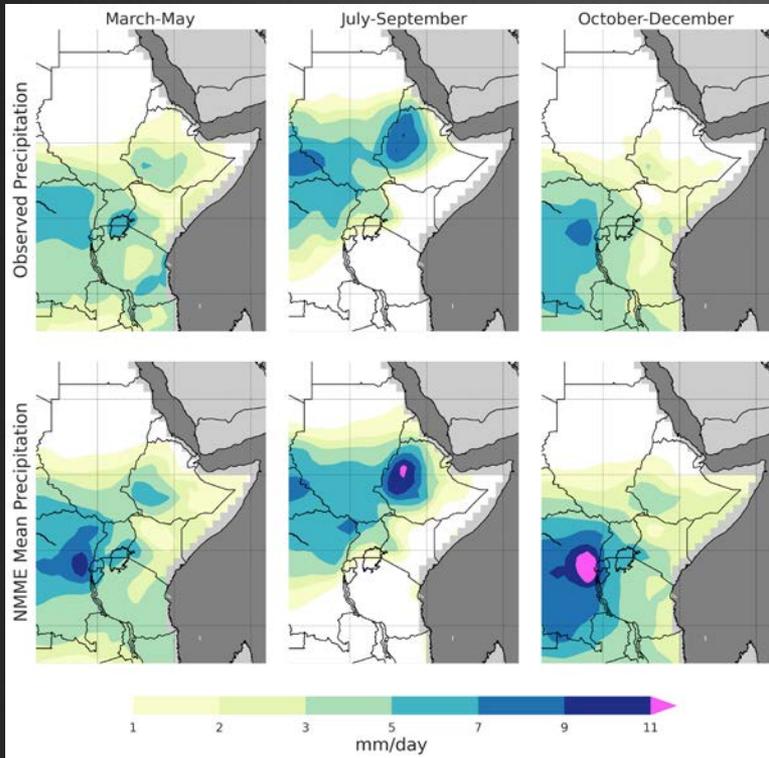
Evaluation of seasonal climate forecasts

North-American Multi-model Ensemble (NMME)

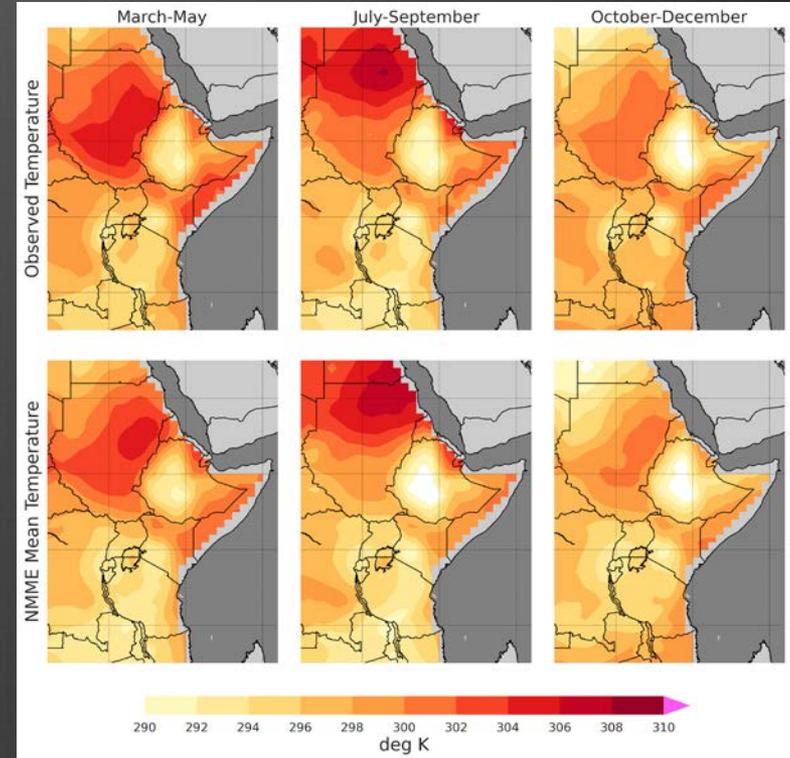
- ❑ Total 99 ensemble members
- ❑ Models used:
 - CMC1-CanCM3
 - CMC2-CanCM4
 - CCSM4
 - GFDL-CM2p1-aer04
 - GFDL-CM2p5-FLOR-A06
 - GFDL-CM2p5-FLOR-B01
 - GMAO
 - CFSv2
- Hindcast available for 1982-2010 at 1 deg X 1 deg spatial resolution.
- Evaluation performed for 1982-2010 period with respect to **The Climate Hazards Group Infrared Precipitation with Stations (CHIRPS)** precipitation and the **Climate Research Unit (CRU's)** temperature data.

Evaluation of monthly mean climatology

Precipitation



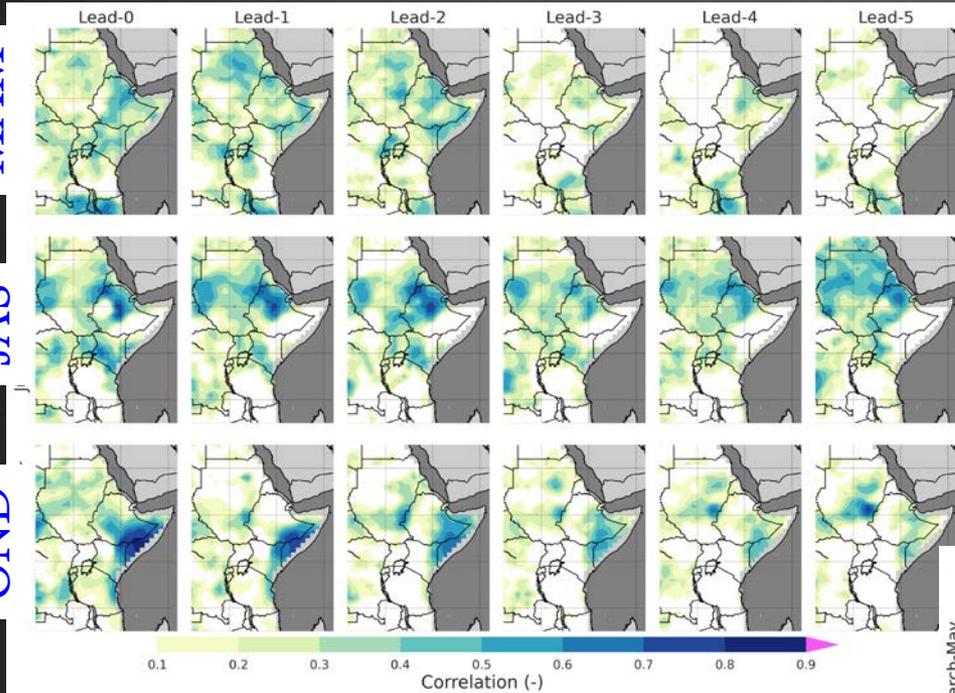
Temperature



The NMME mean in general captures the spatial differences in the climatological mean of both precipitation and temperature. However in general the NMME mean is wetter and colder than the observations.

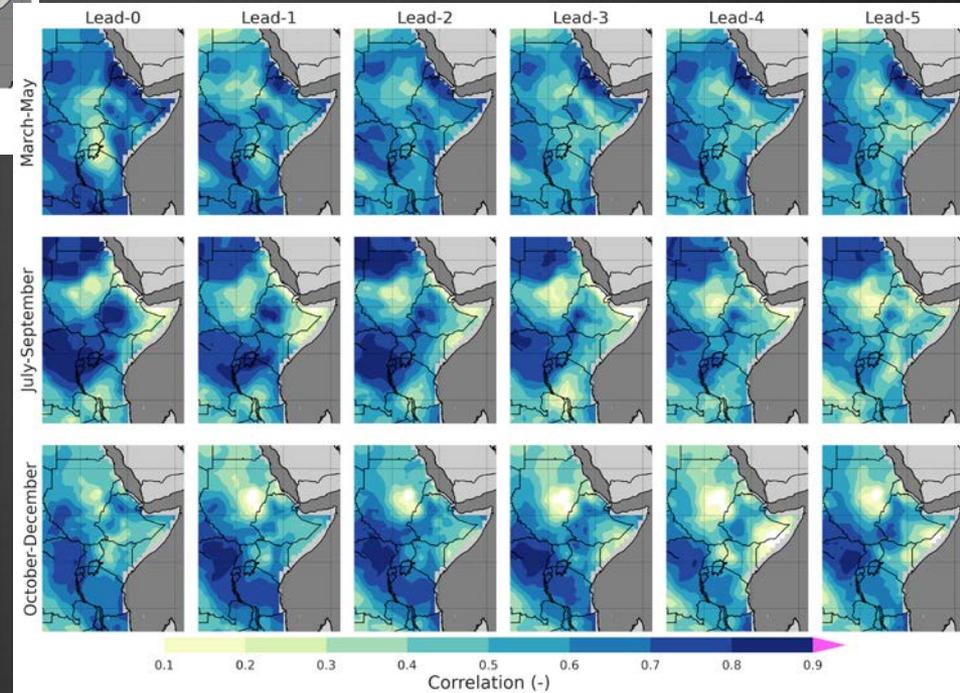
Evaluation of interannual variability

Precipitation



- Precipitation forecast skill in this region is generally limited.
- The temperature forecast skill is generally greater than the precipitation forecast skill and correlation score greater than 0.50 is apparent over a large part of the domain in case of all seasons.

Temperature



Forecast lead 0 months to 5 months

- In case of precipitation the correlation value is generally the lowest for MAM season and highest for the OND season.

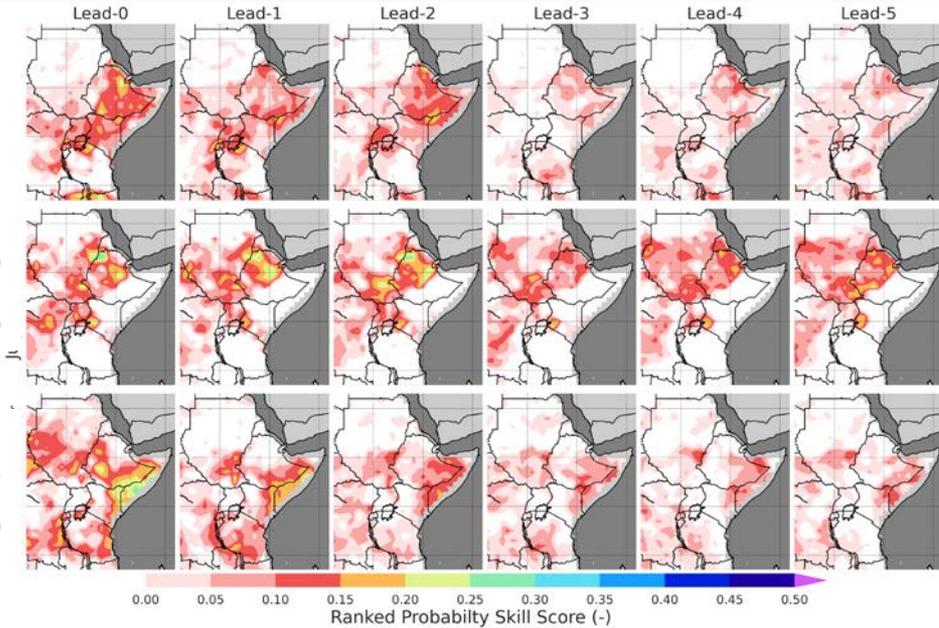
Evaluation of probabilistic forecast

Precipitation

MAM

JAS

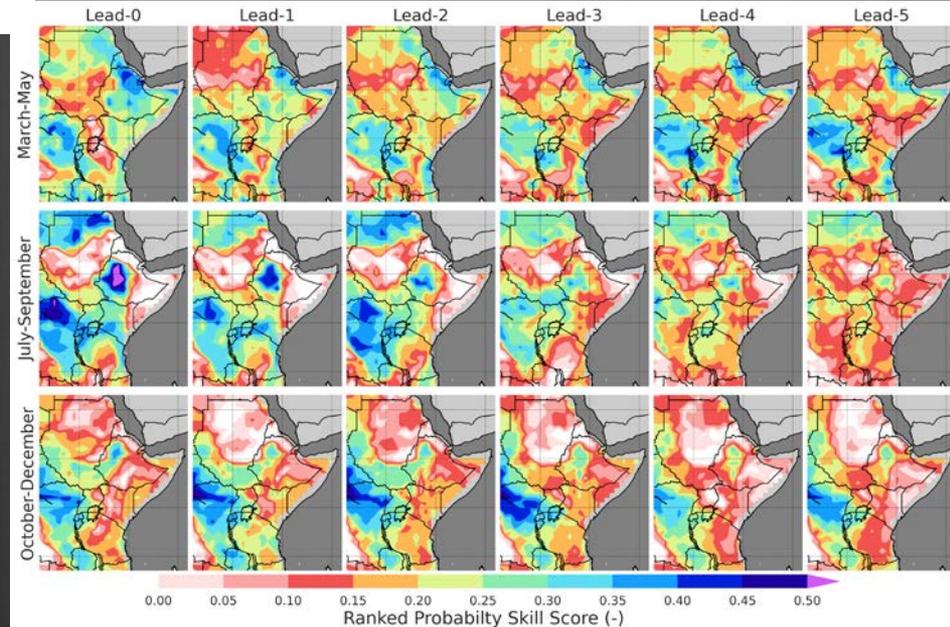
OND



- The Ranked Probability Skill Score (RPSS) presented here, evaluates the distance between the forecast probability and observed probability.
- The RPSS for temperature is much higher than the precipitation and greater than 0.30 RPSS is apparent over a much larger part of the domain.

Temperature

Forecast lead 0 months to 5 months



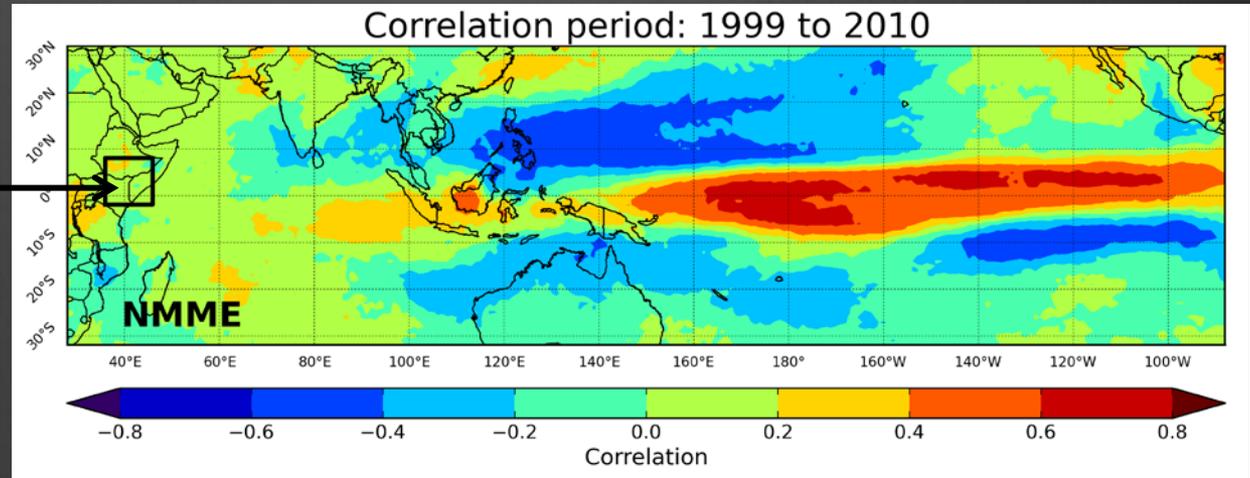
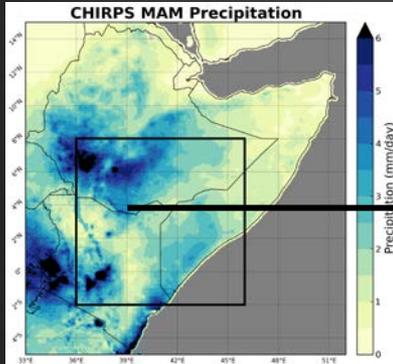
Hybrid approaches to improve precipitation forecast skill in the region

What is a hybrid approach?

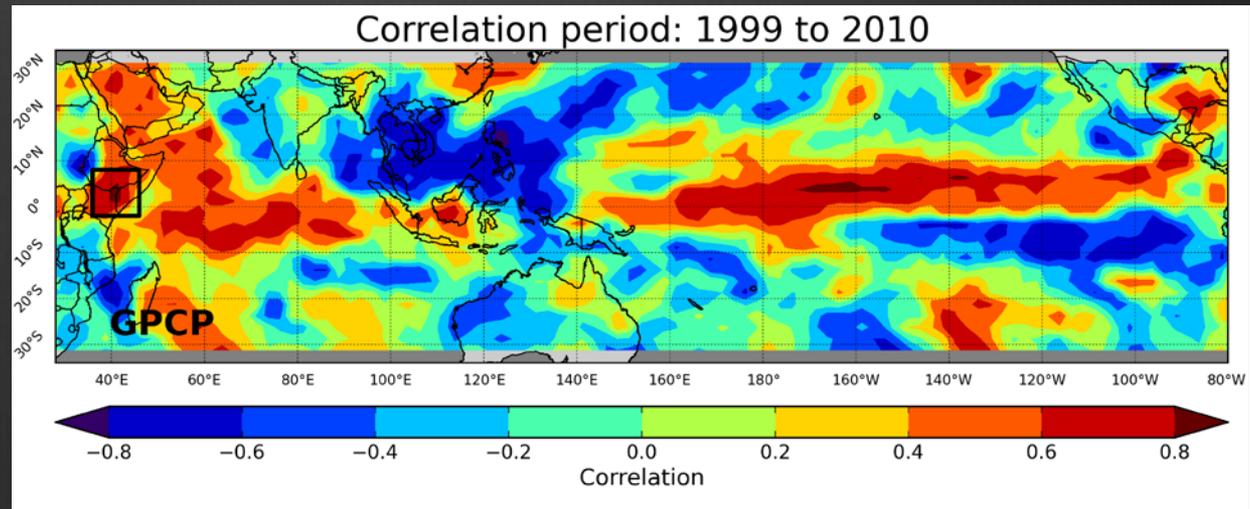
- ⊗ A combination of statistical method and dynamical climate forecasts that harnesses recent teleconnection between EA MAM rainfall over the focus domain and Indo-Pacific Precipitation. **Shukla et al., 2014 (HESS) and Shukla et al., 2014 (ERL)**
- ⊗ Thus far only implemented and evaluated for the MAM season and over equatorial East Africa.
- ⊗ Similar approaches for other seasons and regions are to be explored in future.

A Hybrid approach

Median of the correlation with each of NMME precipitation forecast superensemble

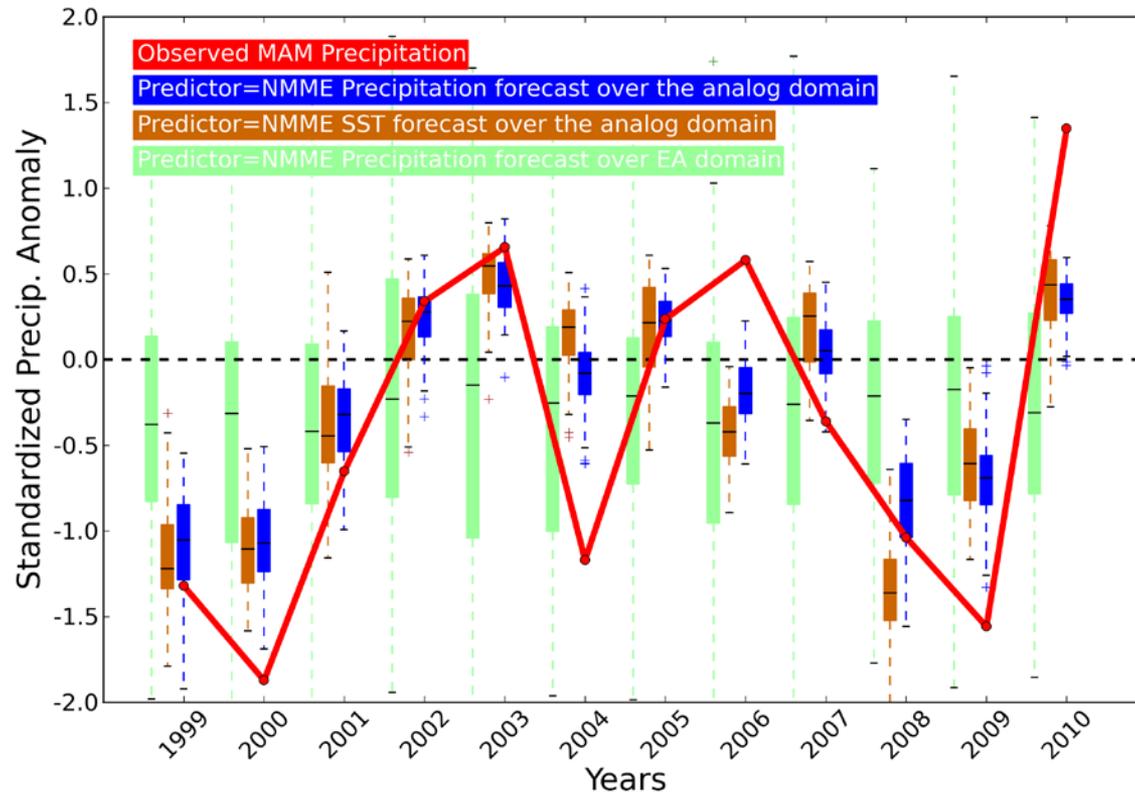


Correlation with GPCP



Shukla et al., 2014
(ERL)

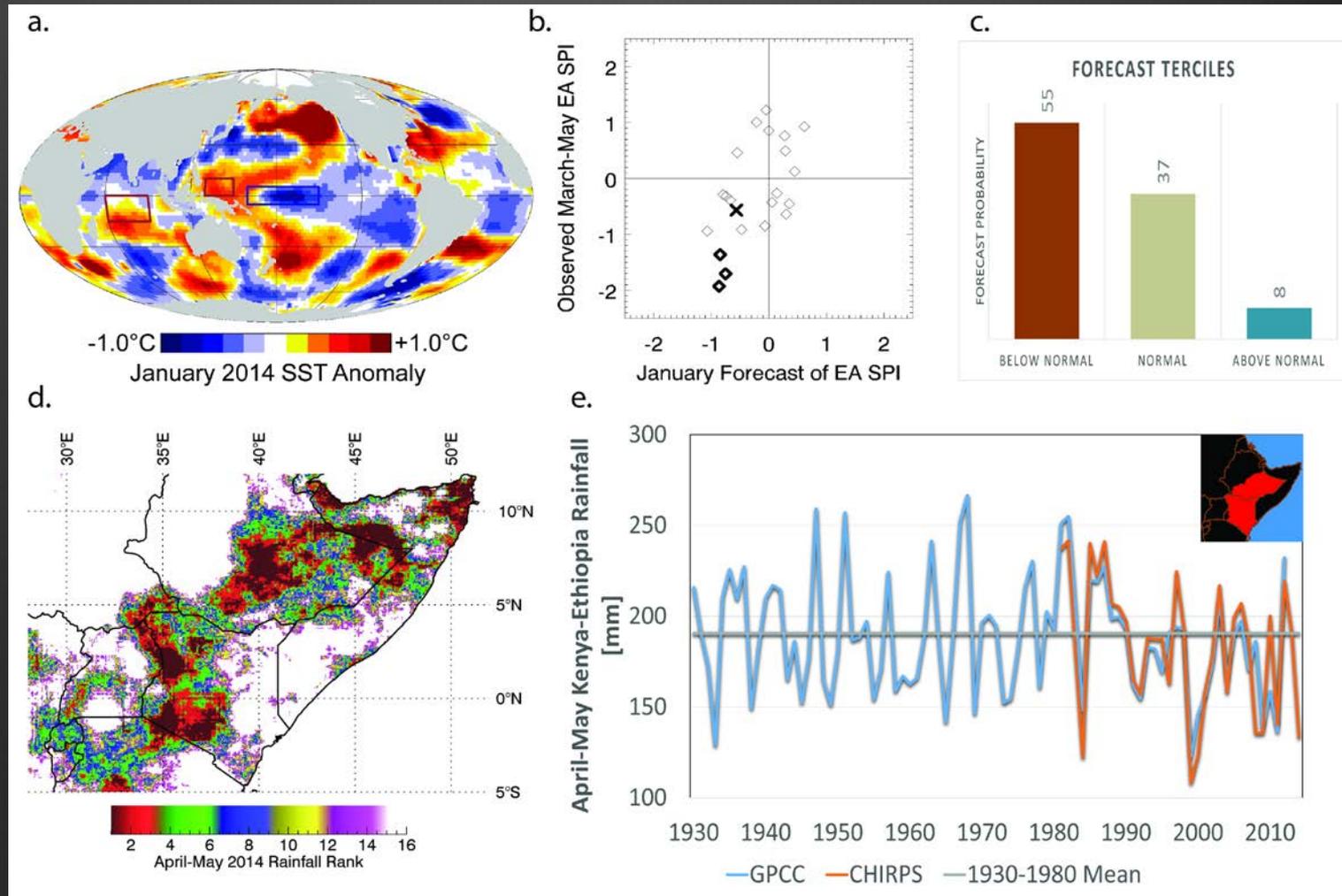
A Hybrid approach



- Large scale predictors (e.g. Indo-Pacific precipitation and SST) result in greater equatorial EA MAM rainfall skill than the local scale predictor.

Another example of a Hybrid Approach

Funk et al., 2014



Implementation and evaluation
of a seasonal agricultural
drought monitoring and
forecasting system

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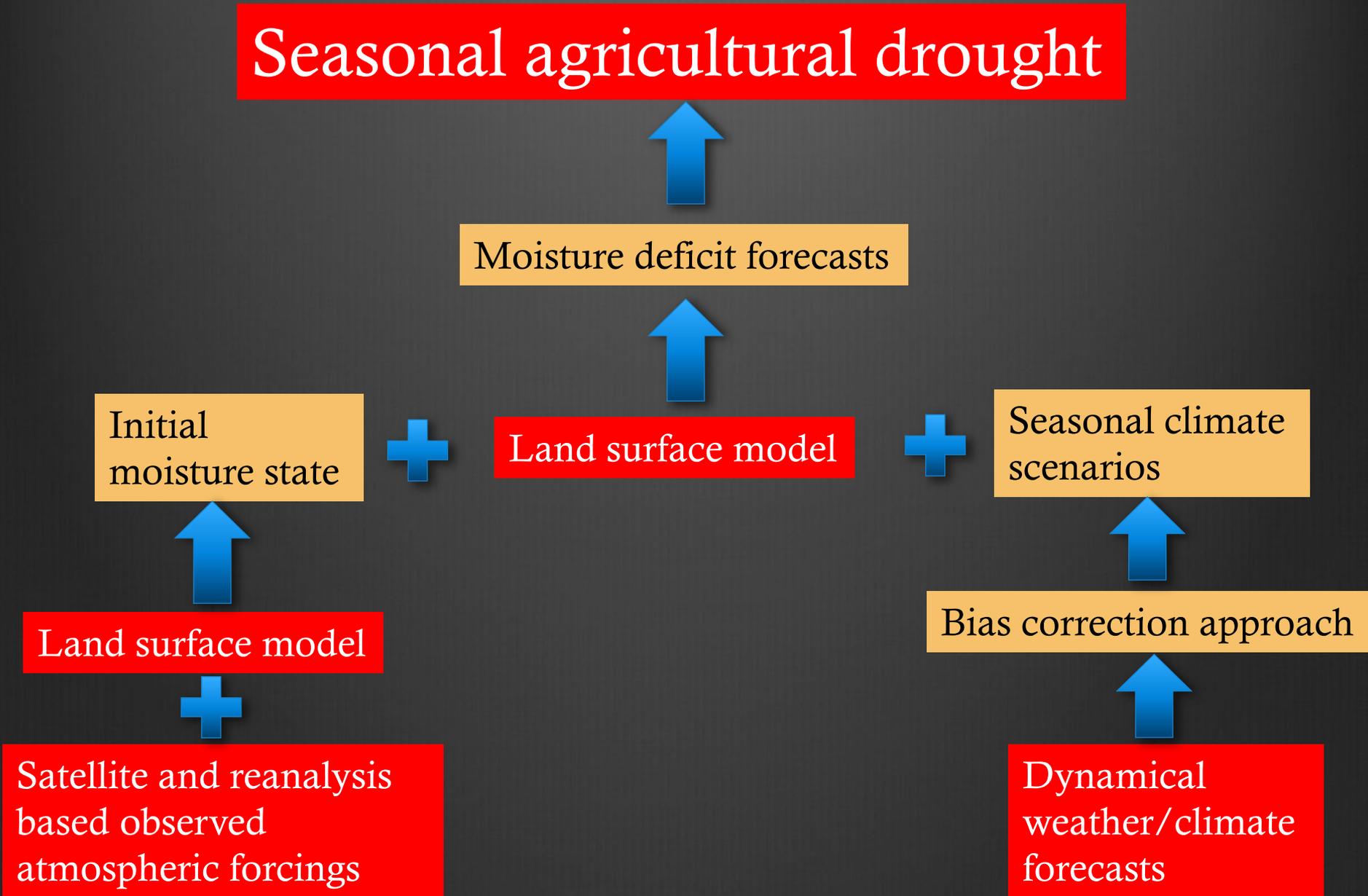
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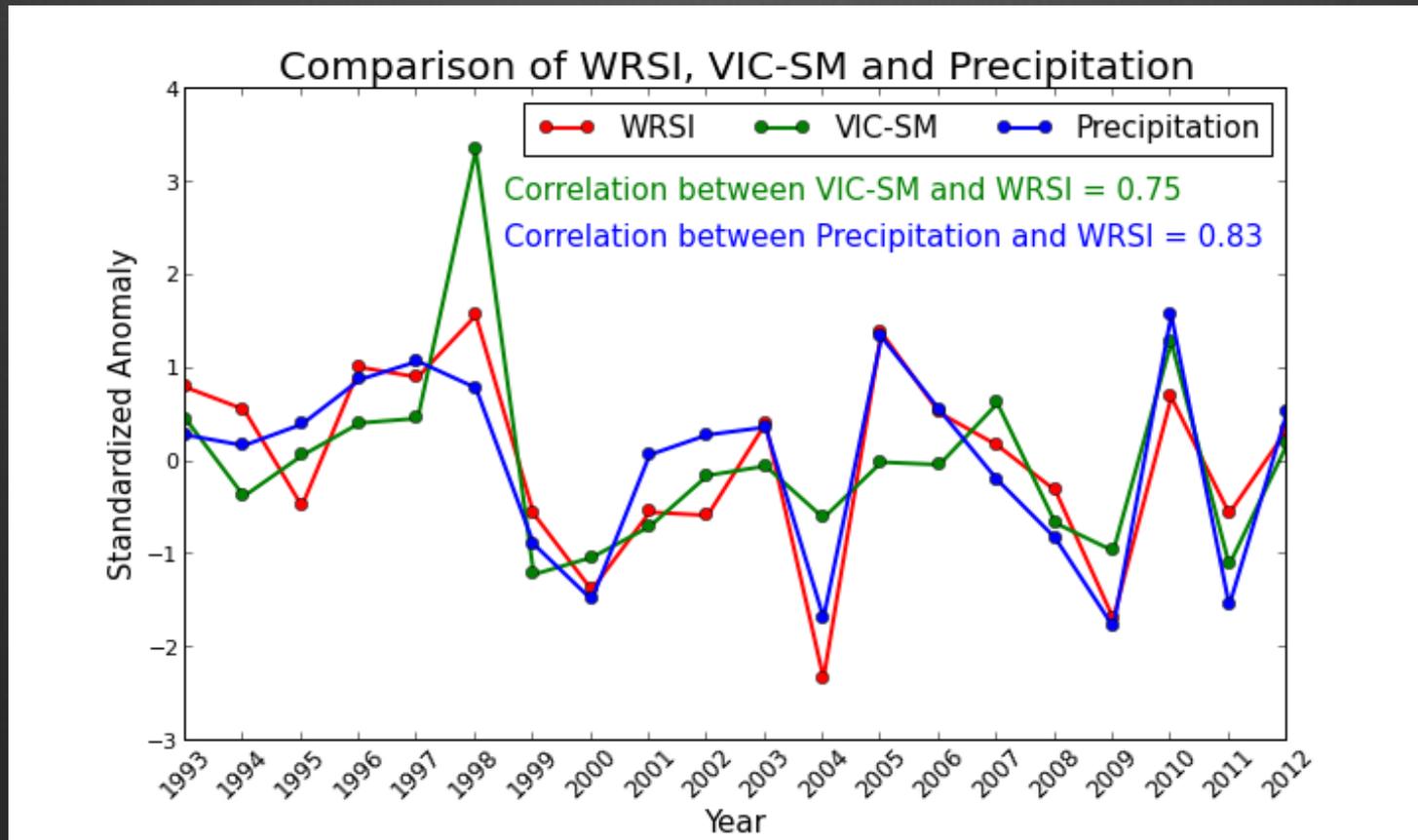
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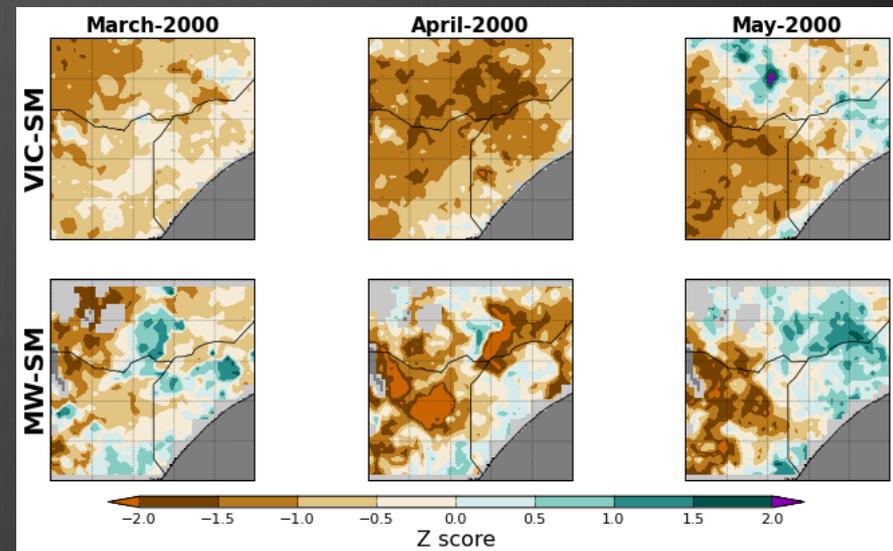
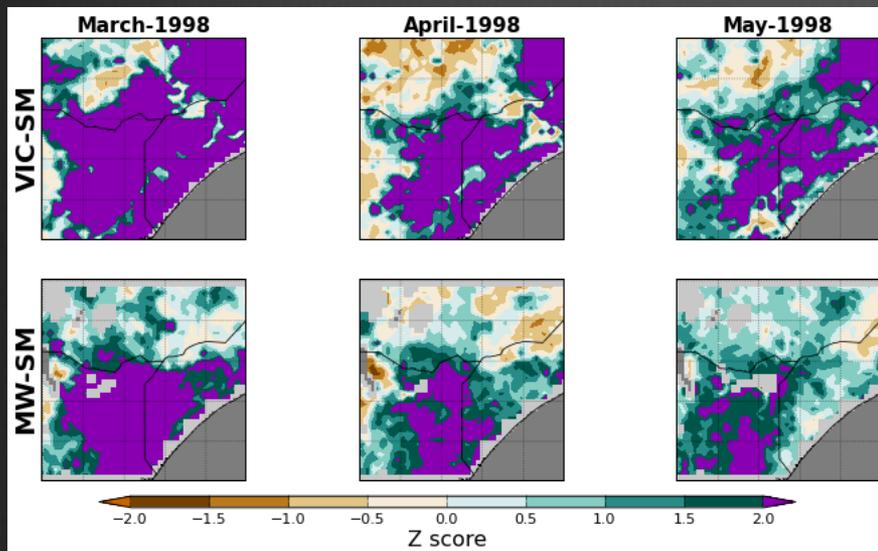


Monitoring



Comparison of MAM precipitation, VIC-soil moisture (VIC-SM) and End-of-Season Water Requirement Satisfaction Index (WRSI) for crop zones in the focus domain for each year between 1993-2012.

Monitoring

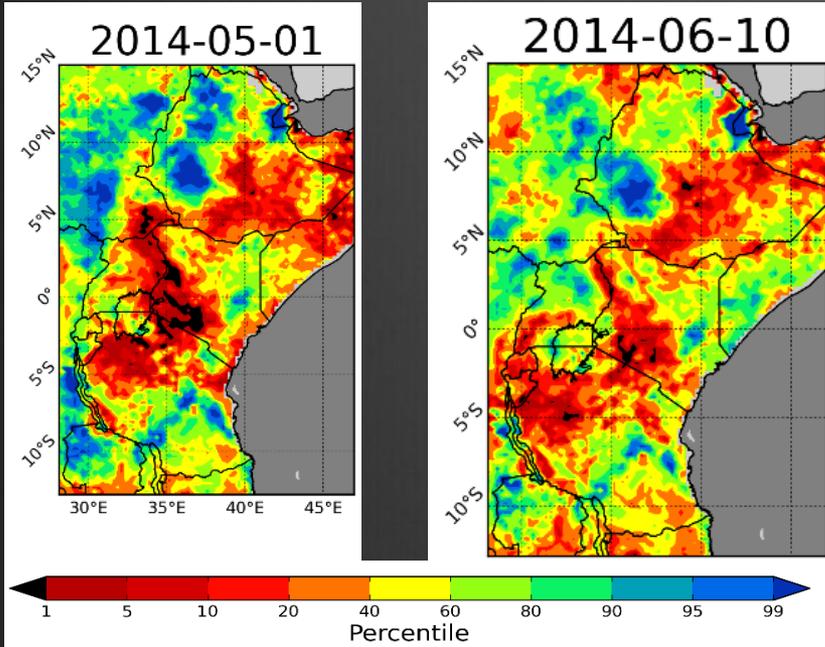


<http://www.esa-soilmoisture-cci.org/>

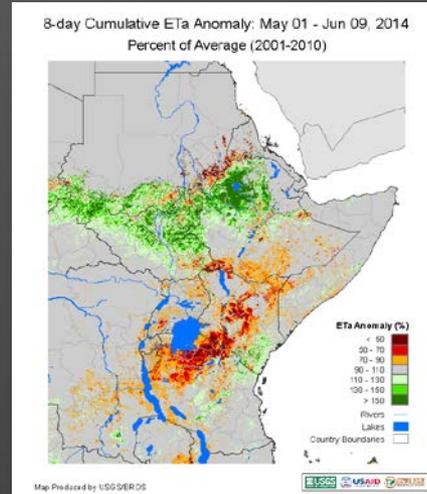
Comparison of VIC-SM with ESA's ECV soil moisture dataset for a wet and a dry years indicate reasonable agreement in both products. Longer terms comparisons have been documented in [Mcnally et al., 2015 \(in review\)](#)

March-June 2014

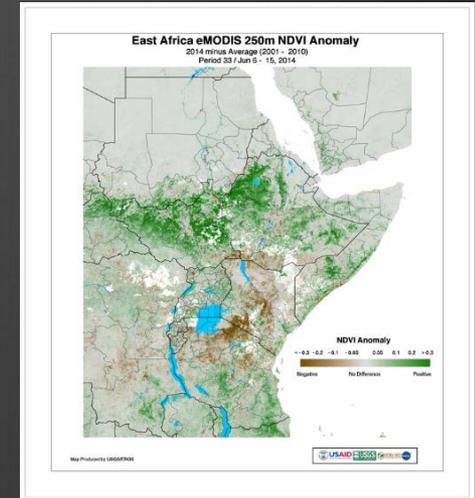
VIC SM percentile



Eta Anomaly

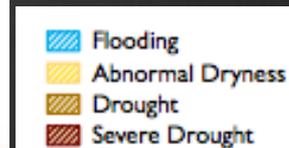
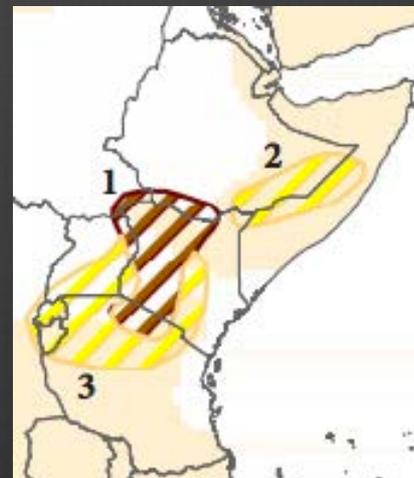


eModis NDVI



<http://earlywarning.usgs.gov/>

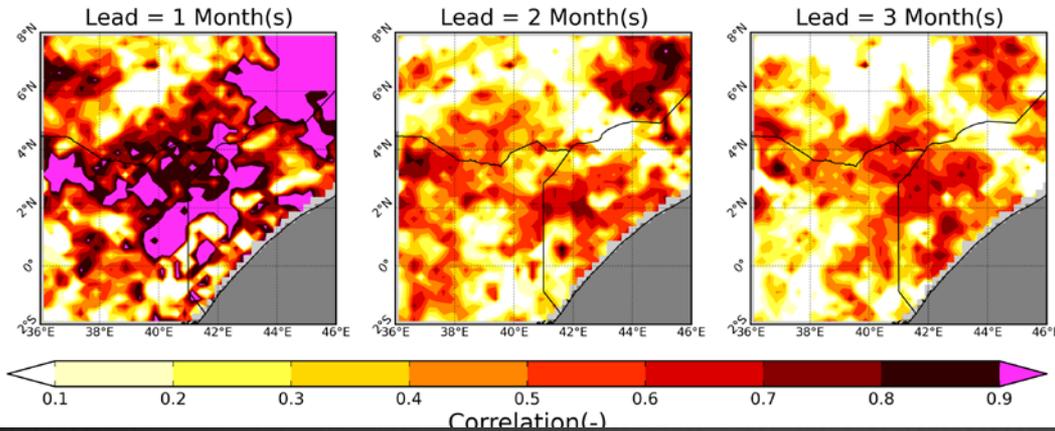
FEWS NET Weather Hazards
Summary, June 6-12th, 2014



www.fews.net

Forecasts

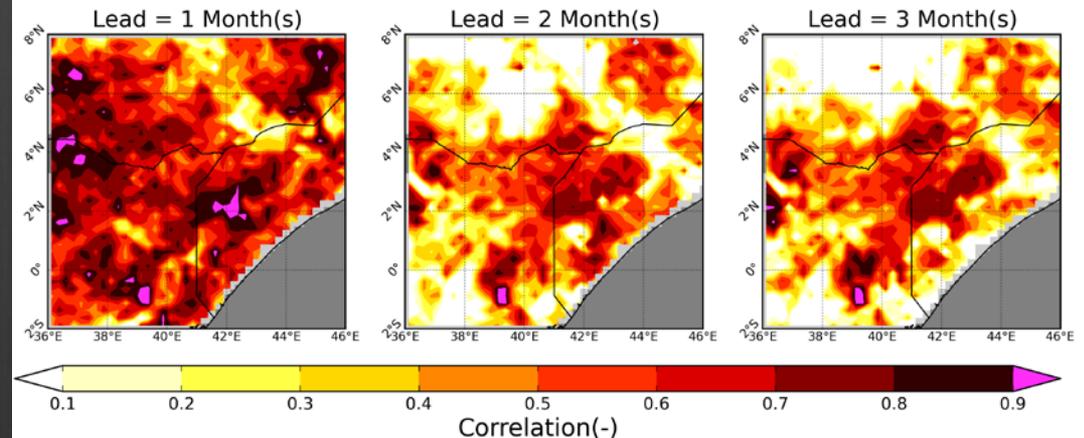
Forecast initialized on March 05



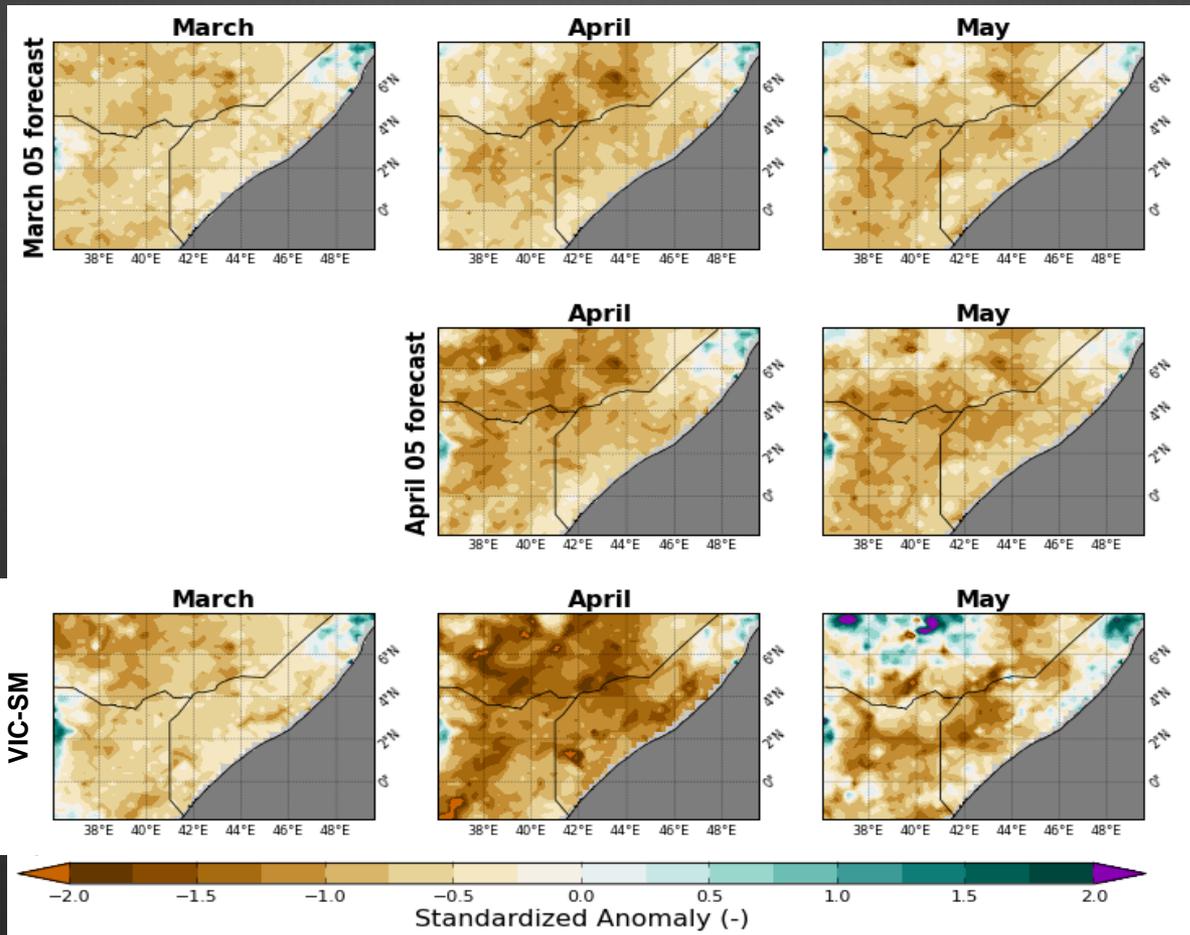
- Correlation between ensemble median SM forecasts (generated using the hybrid approach based climate scenarios) and VIC-SM

- The skill through the rest of the season, is the highest when the SM forecast was initialized in the middle of the season.

Forecast initialized on April 05



2011 MAM Forecast



- The system would have successfully forecasted the 2011 MAM drought as early on March 5th however the drought intensity would have been underestimated (likely due to the small analog sample used to generate climate scenarios)

Concluding Remarks

Concluding remarks

- ⊗ The NMME precipitation forecast skill in East Africa is limited at best. Temperature forecast is generally much more skillful across the region.
- ⊗ The hybrid approach proposed in this study results in improved MAM precipitation forecast skill in the region.
- ⊗ VIC model derived SM values are suitable for providing agricultural drought outlook assessment in the region.
- ⊗ SM forecasts initialized at the beginning of the season are skillful across the domain at 1-month lead and for some parts over 3-month lead.
- ⊗ Contribution from the antecedent SM state to SM forecast skill during rest of the season could be most useful when the forecast is initialized in the middle of the season.

Acknowledgement

- ⊕ Primary support for this was provided by the USAID's FEWS NET (USGS award #G09AC00001) and NOAA Technical Transitions grant NA11OAR4310151.

For further details see:

Shukla, S., McNally, A., Husak, G., and Funk, C.: A seasonal agricultural drought forecast system for food-insecure regions of East Africa, *Hydrol. Earth Syst. Sci. Discuss.*, 11, 3049-3081, doi:10.5194/hessd-11-3049-2014, 2014.

Shukla S., Funk, C., and Hoell, A.: Using constructed analogs to improve the skill of National Multi-Model Ensemble March-April-May precipitation forecasts in equatorial East Africa. *Environ. Res. Lett.* 9, 094009. 2014

Funk, C., Hoell, A., Shukla, S., Bladé, I., Liebmann, B., Roberts, J. B., Robertson, F. R., and Husak, G.: Predicting East African spring droughts using Pacific and Indian Ocean sea surface temperature indices, *Hydrol. Earth Syst. Sci.*, 18, 4965-4978, doi:10.5194/hess-18-4965-2014, 2014.

Shukla, S., Roberts, J., Hoell, A., Funk, C., and Robertson F.: Assessing North American Multi-model Ensemble (NMME) Seasonal Forecast Skill to Assist in the Early Warning of Hydrometeorological Extremes over East Africa. *BAMS* (to be submitted)

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Thank you!