

Monitoring and Forecasting Water Availability in Food Insecure Regions

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Jossy Jacob, and NASA GSFC Land Information
System Team



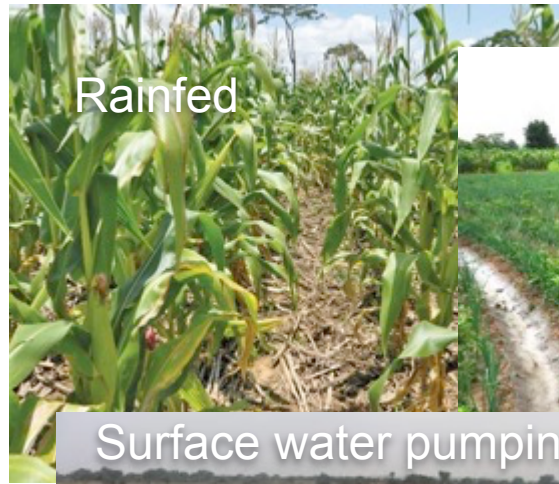
Water availability, for what?

Soil water storage: agriculture & pasture (i.e. plant water availability), water points for people and animals

Groundwater storage: irrigation, domestic water supply

Surface water storage: reservoirs, lakes, streams used for domestic, agricultural and industrial needs

Snow water storage: irrigated agriculture and domestic water supplies



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Snow water storage: irrigated agriculture and domestic water supplies

The Famine Early Warning Systems Network (FEWS NET) has long history of monitoring plant available water for agriculture and pasture with NDVI and rainfall statistics.

NASA LIS modeling capabilities let us model different storage reservoirs to monitor conditions that impact food security, beyond rainfed agriculture.

FEWS NET LDAS (FLDAS) Africa

Parameters:

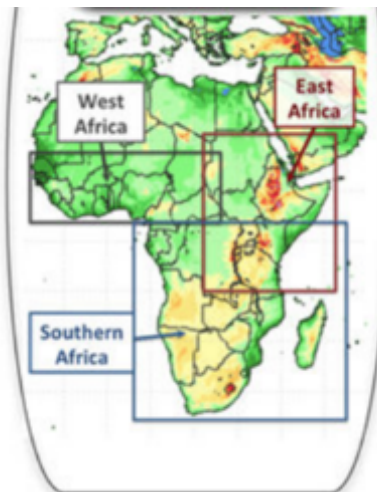
IGBP MODIS landcover,
FAO soils, GTOPO30
elevation, NCEP albedo
(quarterly/seasonal)

Meteorological Inputs:

MERRA-2, GDAS
CHIRPS, CHIRPS-
prelim, RFE2

LIS-based LSMs:

Noah3.3
10km x 10km resolution,
VIC 4.1.2
25km x 25 km resolution



Daily & Monthly outputs:

Full water balance e.g. ET,
SM, runoff
Full energy balance e.g
sensible heat flux, ground
heat flux, net radiation

Evaluation

remote sensing
(e.g. MODIS NDVI, SSEBop
ET, ESA CCI SM), GRDC
streamflow, GRACE TWS

Data delivery:

LIS Data
Portal,
USGS Early
Warning
Data Portal,
NASA GES
DISC &
GIOVANNI

McNally, Amy, Kristi Arsenault, Sujay Kumar, Shraddhanand Shukla, Pete Peterson, Shugong Wang, Chris Funk, Christa D. Peters-Lidard, and James P. Verdin, 2017: "A land data assimilation system for sub-Saharan Africa food and water security applications." Scientific Data 4. doi:

[10.1038/sdata.2017.12](https://doi.org/10.1038/sdata.2017.12)

Detailed specifications available at <https://ldas.gsfc.nasa.gov/FLDAS/FLDASspecs.php>

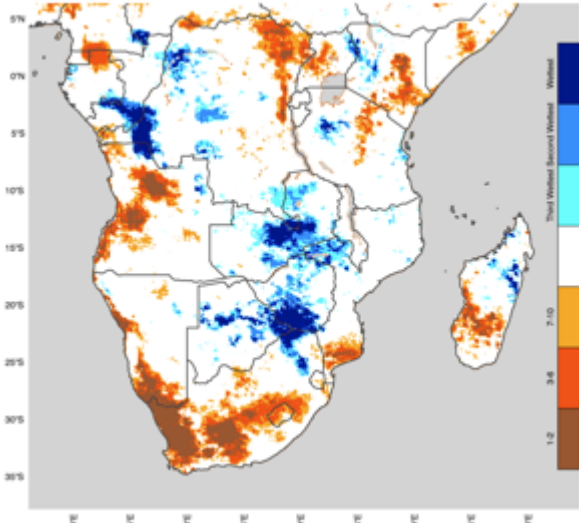
FLDAS Drought Monitoring



Plant water availability 2017/18

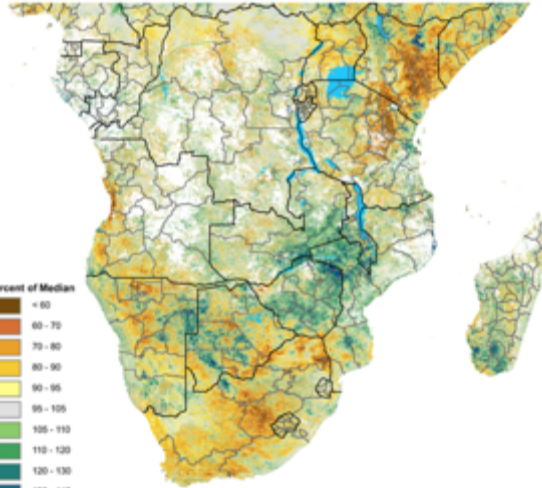
Subsurface soil moisture rank

Oct_Dec Soil Moisture (10-40cm) Rank 2017

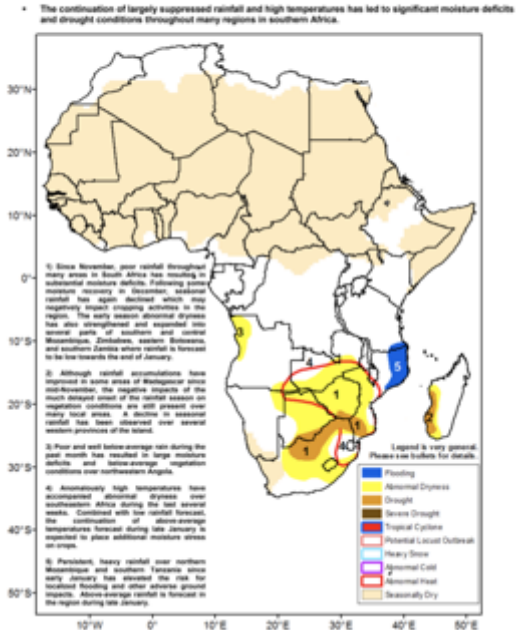


Southern Africa
eMODIS 250m Percent of Median NDVI

Period 36 / Dec 21 - 31, 2017



Climate Prediction Center's Africa Hazards Outlook
January 25 - January 31, 2018

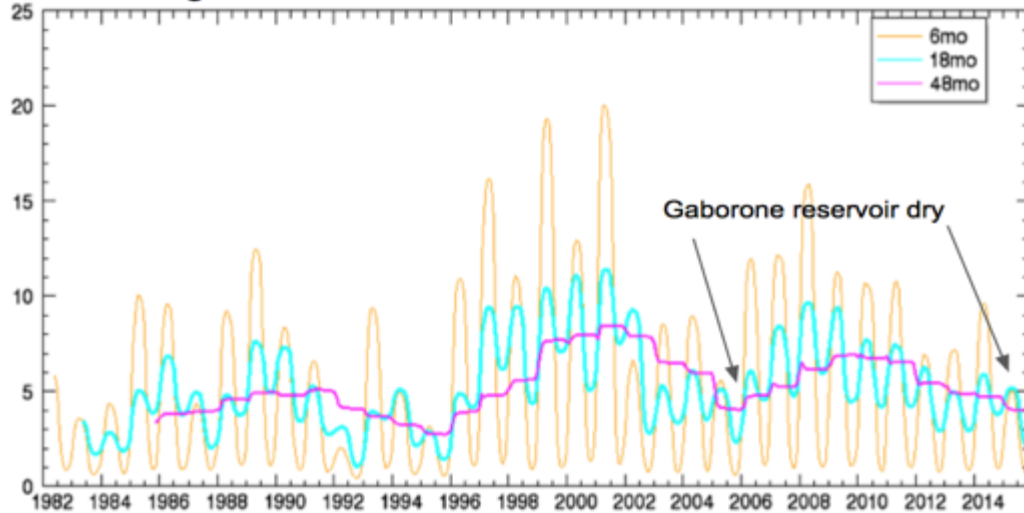


Southern Africa was facing abnormally dry, abnormally hot and drought conditions as classified by NOAA CPC. We find that the Oct-Dec average surface and subsurface soil moisture is some of the driest since 1982.

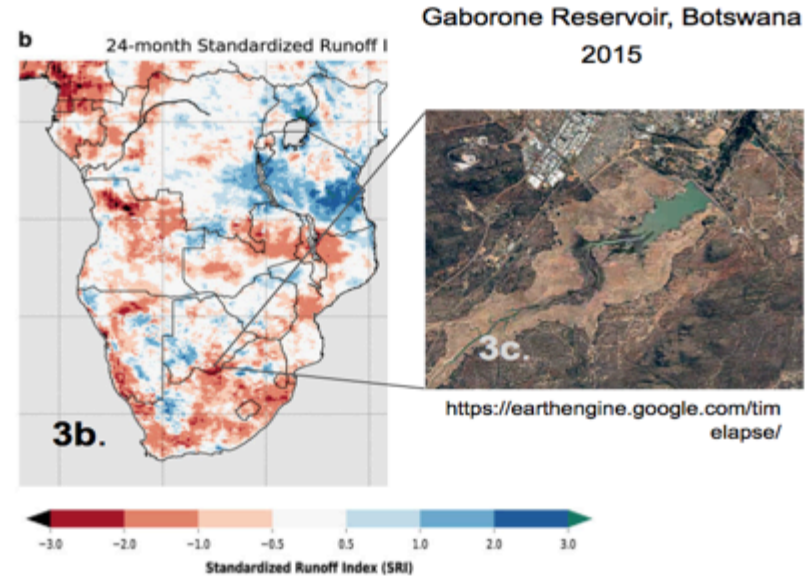
FLDAS Reservoir Monitoring



regional mean runoff at different time scales

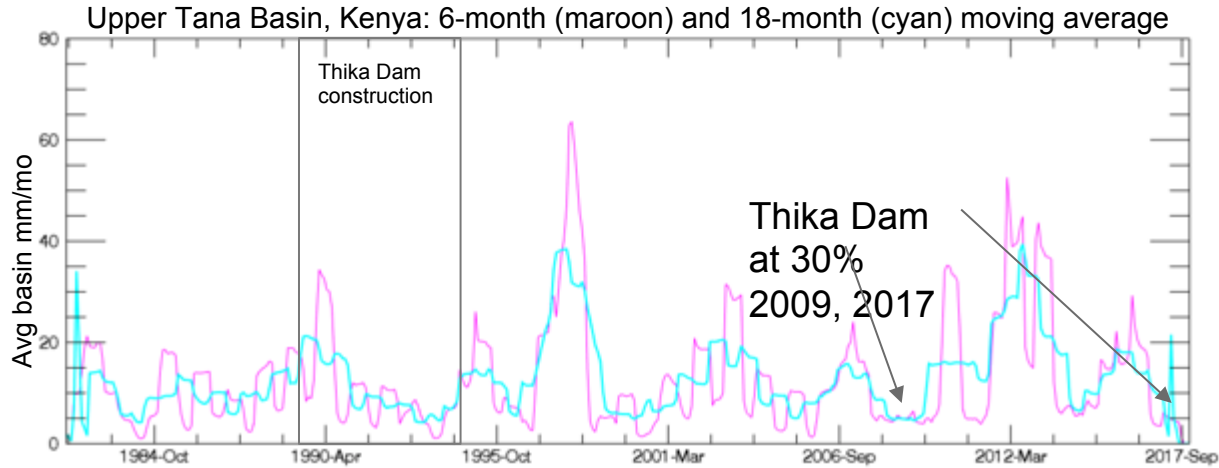


Gaborone Dam, Botswana 2015-16 drought

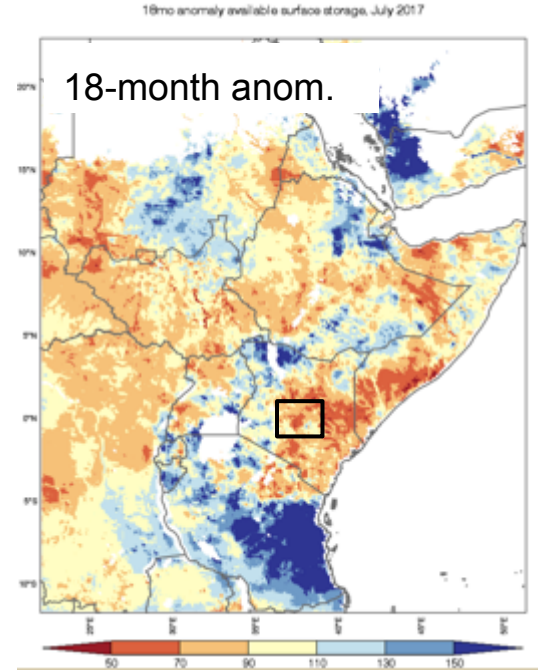


Time series analysis and Standardized Runoff Index (SRI-1, SRI-24)

FLDAS Reservoir Monitoring

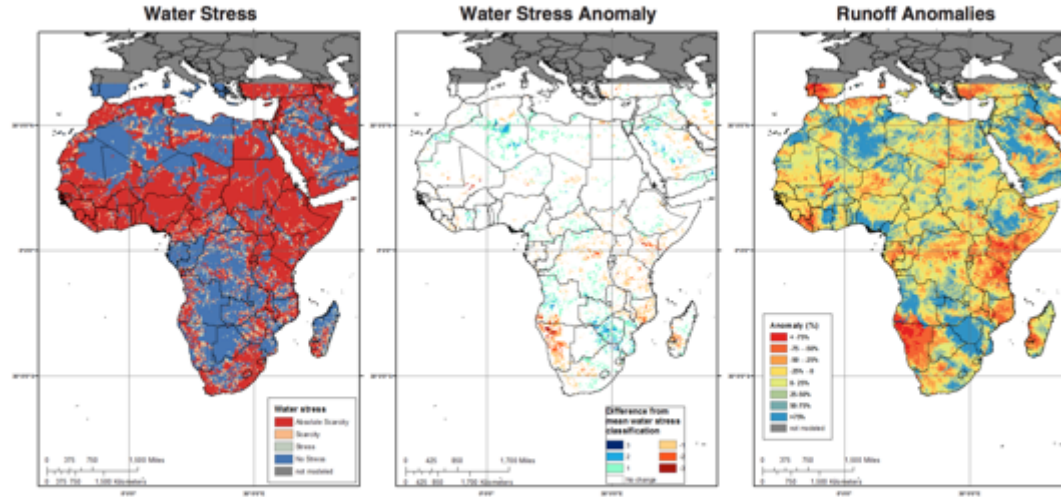


Thika Reservoir, Kenya Sept-2017



Time series analysis and Multi-month anomalies

Water Stress per Falkenmark Index (runoff per capita)



Water Stress maps highlight locations experiencing water stress based on current runoff and 2015 population.

Water Stress Anomaly maps highlight departure from average (1982-2017).

Runoff Anomaly maps highlight water supply departures from average conditions (1982-2017).

Water Stress Thresholds from Falkenmark

Water Stress

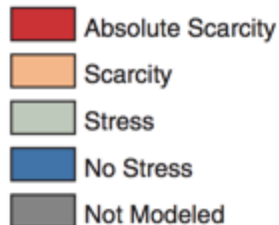
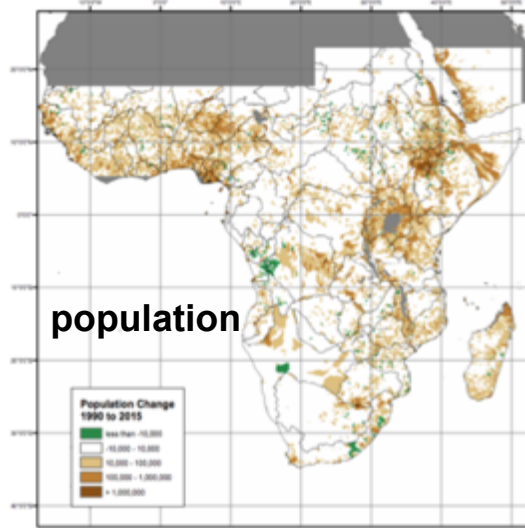
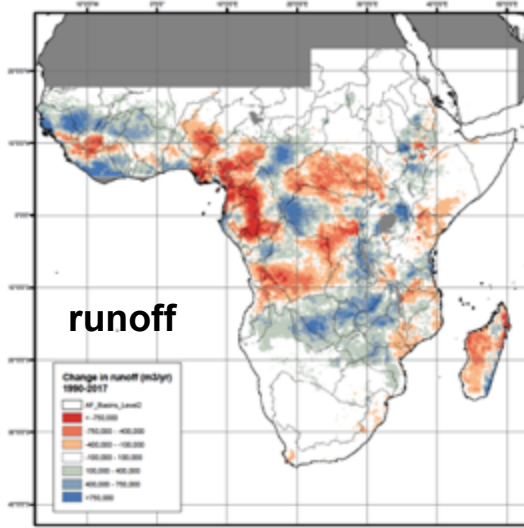


Table 1. Annual and Monthly Falkenmark Categories

category	m3/yr/cap	m3/mo/cap
no stress	>1700	>142
stress	1000–1700	83–142
scarcity	500–1000	41–82
absolute scarcity	<500	<41

Maps updated twice a month at <https://lis.gsfc.nasa.gov/projects/fewsnet>

Trends in Runoff & Population

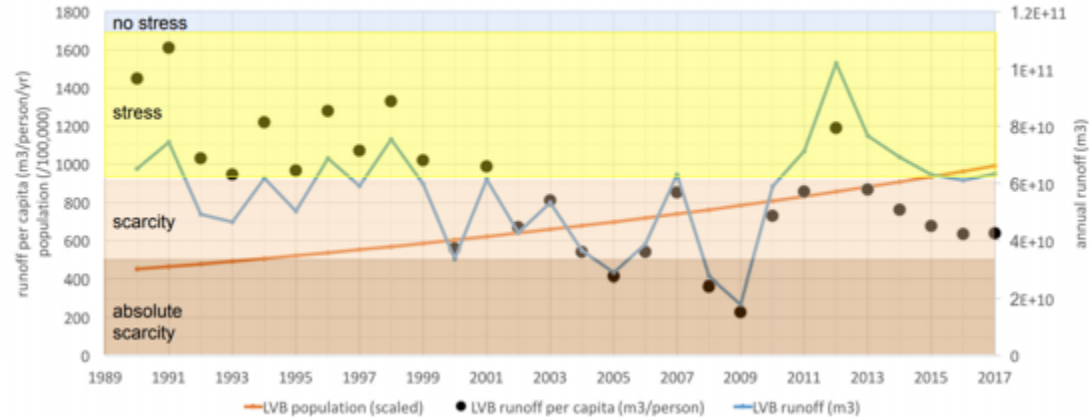


Maps of annual runoff and population change, and time series of the Lake Victoria Basin show the relationship between runoff and water stress.

Water Stress Change Over Time: Lake Victoria Basin

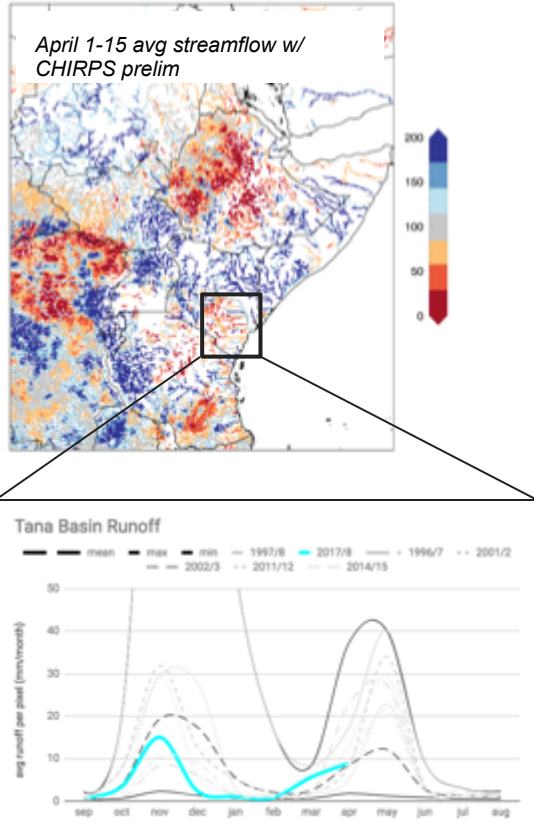
Lake Victoria Basin Runoff, Population and Runoff per capita

Annual runoff varies from year to year, while water availability trends downward with increasing population.

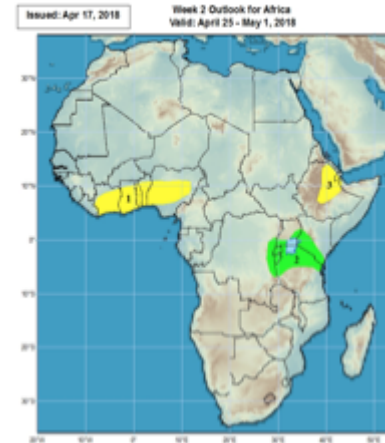


Water Availability Forecasts

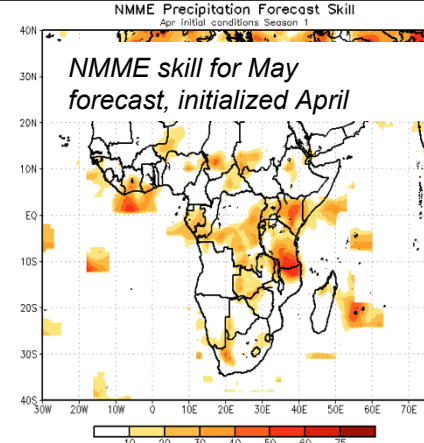
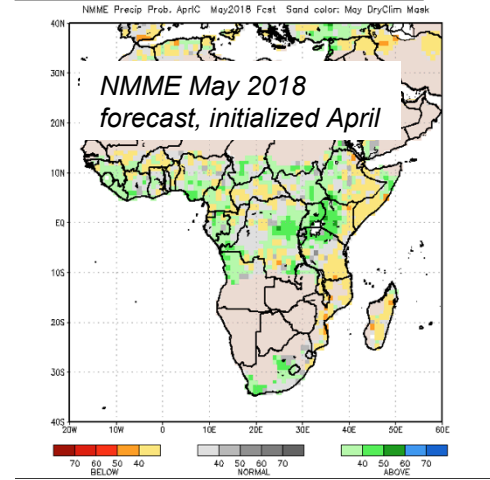
Current Conditions & Historical Context



1-week & 2-week weather forecasts



NMME Probability for next 1 - 6 months



FLDAS Model Evaluation-East and Southern Africa

Basin	SSEBop ET vs Noah33 (R)	GRACE TWS vs Noah33 (R)	Streamflow (R) (#stations)
Blue Nile	0.49	0.85	0.46-0.94 (11)
Awash	0.72	-	0.47-0.72 (3)
Juba-Shabelle	0.8	-	0.41-0.53 (4)
Upper Tana	0.64	-	-
Rufiji	0.56	-	0.41-0.61
Pangani	0.74	-	0.41-0.81
Orange	0.72	0.74	0.61-0.66 (3)
Zambezi	0.72	0.76	
Limpopo	0.78	0.66	
Wami-Ruvu	-	-	0.41-0.61

FLDAS Noah33 CHIRPS+MERRA-2 estimates are well correlated with independent observations in Southern and Eastern Africa basins.

Streamflow correlations are a function of human influence (e.g. irrigation, dams).

FEWS NET LDAS (FLDAS) Central Asia

Parameters:

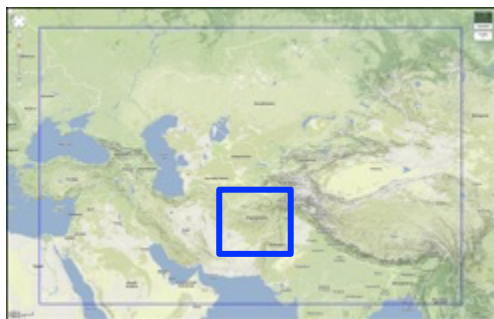
UMD-AVHRR
landcover, FAO soils,
GTOPO30 elevation,
NCEP albedo (quarterly/
seasonal)

Meteorological Inputs:

6 hourly GDAS - NCEP
Global Data
Assimilation System
Final (FNL) Operational
Global Analysis

LIS-based LSMs:

Noah3.6
1km x 1km resolution,
1 hr model timestep,
No snow assimilation



Daily - near real time outputs:

Snow water equivalent
Snow covered area
Snow depth
Air temperature

Evaluation

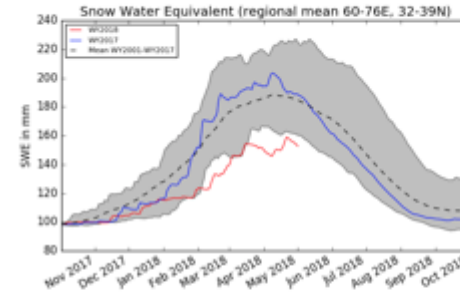
remote sensing
(e.g. MODIS & Landsat
Snow Cover Fraction, SSM/I
& AMSR-2 snow depth)

**Data
delivery:**
LIS Data
Portal,
USGS Early
Warning
Data Portal

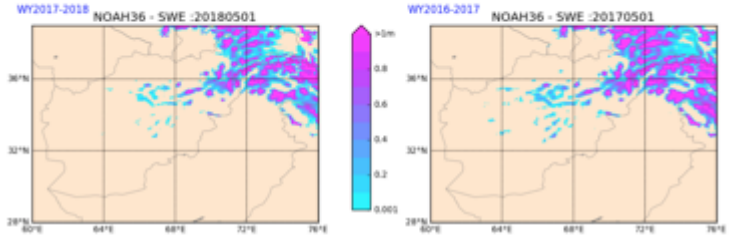
Central Asia Snow Modeling

Estimates of Snow Water Equivalent (SWE) and Snow Depth are provided daily to USGS Early Warning <https://earlywarning.usgs.gov>

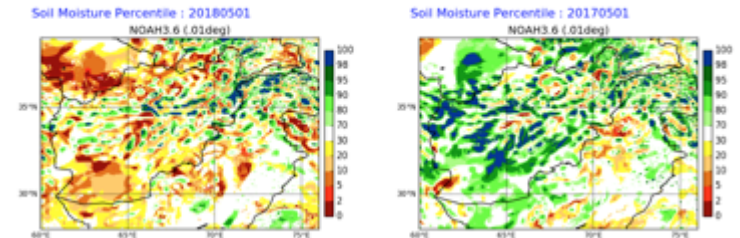
These estimates are then used by CPC NOAA and FEWS NET to track conditions in Afghanistan.



Snow Water Equivalent (SWE) for 2 Water Years - WY2018 and WY2017



Soil Moisture for 2 Water Years - WY2018 and WY2017



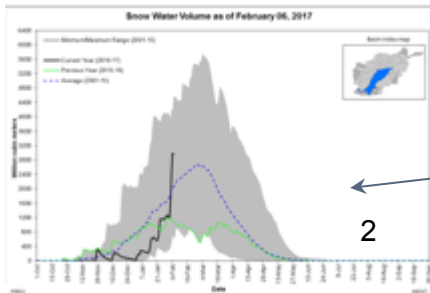
Afghanistan Snow Water Equivalent

Summarized by Basin

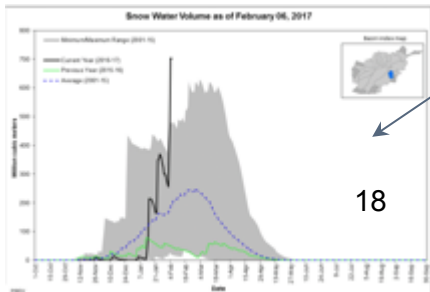
Click on the ID number to view the chart for that basin.



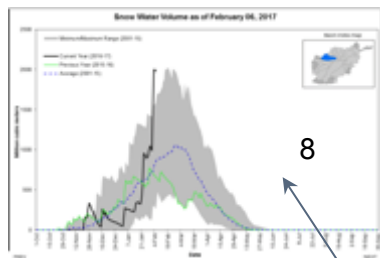
Click on a basin to display snow cover volume charts



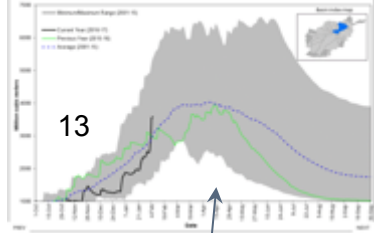
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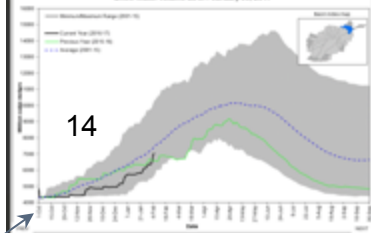
18



8



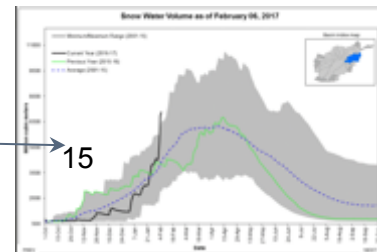
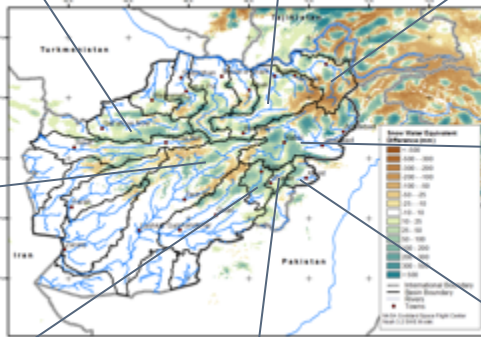
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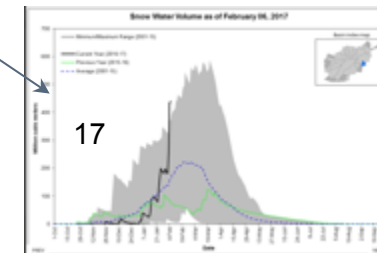
14

Daily Snow Water Equivalent Difference Anomaly

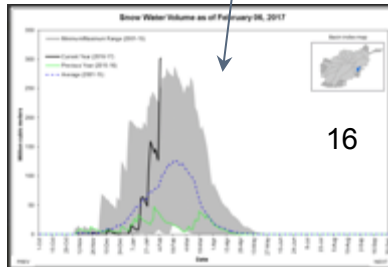
February 06, 2017 minus Average (2002-2014)



15



17



16

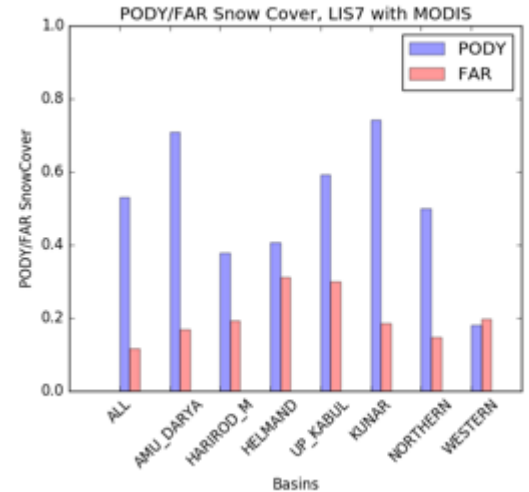
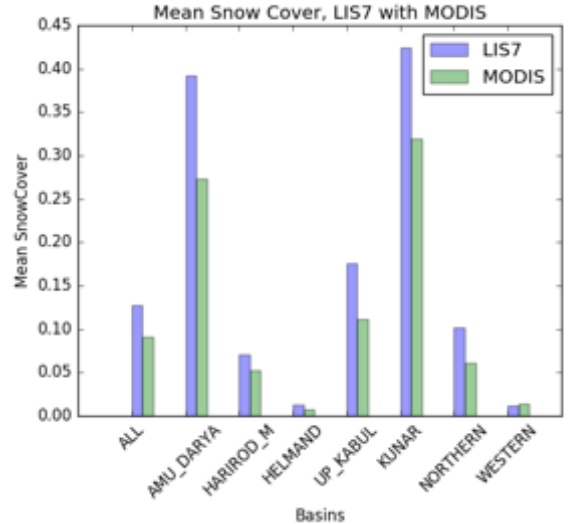
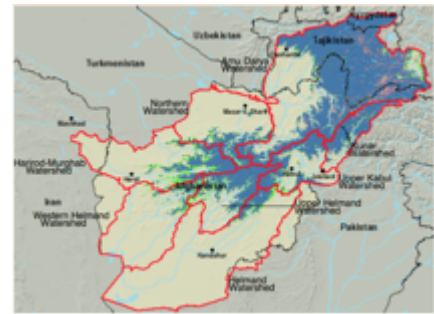
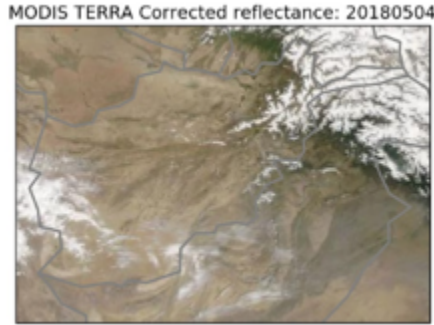
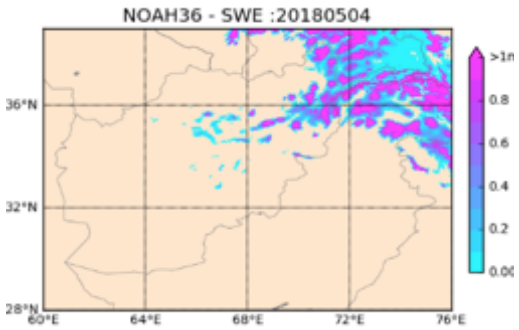
The Daily SWE by basin's tool allows us to compare the current season's evolution to historic season time series.

Central Asia Evaluation

Modeled estimates of SCA are routinely compared to MODIS.

Mean snow cover fraction corresponds well at the basin level during the peak snow season (Water Year 2016).

Probability of Detection (POD) > 0.6 and False Alarm Rate (FAR) < 0.2.



Central Asia Applications

These snow estimates contribute to Food Security Outlooks produced by FEWS NET.

For the current 2017/18 season the low snowpack in Afghanistan is expected to impact crop production.



AFGHANISTAN Food Security Outlook

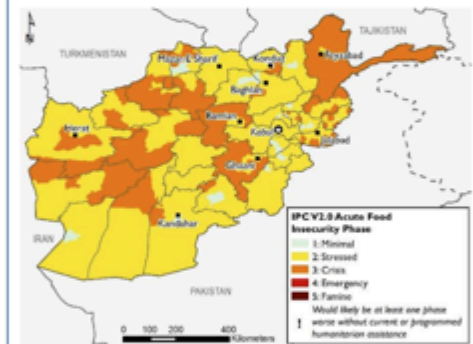
February to September 2018

Low snow accumulation and dry soil conditions likely to impact 2018 staple production

KEY MESSAGES

- The weakening of the casual labor market since 2014 has made it more difficult for poor households to earn sufficient income to support dietary needs during the lean season. Furthermore, 2017 rainfed production was poor in some provinces, including in Ghor, Balkh, Jawzjan, Takhar, Badakhshan, Samangan, Herat, Baghlan, and Sar-i-Pul Provinces. Poor households affected by poor own production or who were not able to find sufficient employment to support food purchases are likely experiencing Crisis (IPC Phase 3) outcomes until local spring labor opportunities facilitate access to income and market purchases of food.
- The ongoing conflict between various insurgent groups, primarily the Taliban and IS, and the Government of the Islamic Republic of Afghanistan has increased in geographic extent and severity in recent years, with more than 1.1 million people displaced since the beginning of 2016. Beyond displacement, insecurity has continued to disrupt normal livelihoods by limiting access to farms, rangelands, markets, and labor opportunities, and by reducing local economic activity. Although seasonal improvements in access to food and income will occur during the spring and summer months, displacement and reduced non-agricultural labor opportunities are expected to drive Stressed (IPC Phase 2) outcomes throughout much of the country during the scenario period, with many households in Crisis (IPC Phase 3).

Current food security outcomes, February 2018



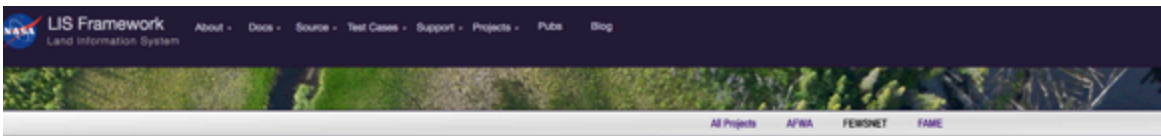
Source: FEWS NET

FEWS NET classification is IPC-compatible. IPC-compatible analysis follows key IPC protocols but does not necessarily reflect the consensus of national food security partners.

Beyond displacement, insecurity has continued to disrupt normal livelihoods by limiting access to farms, rangelands, markets, and labor opportunities, and by reducing local economic activity. Although seasonal improvements in access to food and income will occur during the spring and summer months, displacement and reduced non-agricultural labor opportunities are expected to drive Stressed (IPC Phase 2) outcomes throughout much of the country during the scenario period, with many households in Crisis (IPC Phase 3).

Summary & Next Steps

- *NASA GSFC routinely runs custom instances of the NASA Land Information System for FEWS NET to provide information on water availability in Africa and Central Asia.*
- *Ongoing efforts include:*
 - Expanding the modeling domain globally &*
 - Developing hydrologic forecasts using input from NMME models.*



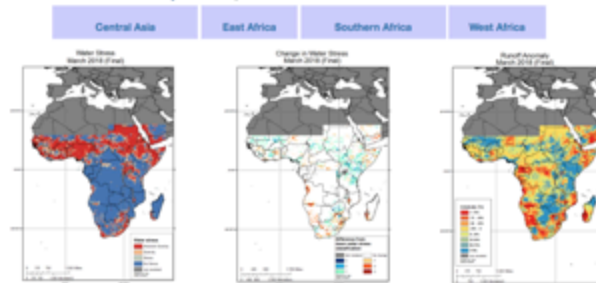
Projects / Famine Early Warning Systems Network

Famine Early Warning Systems Network

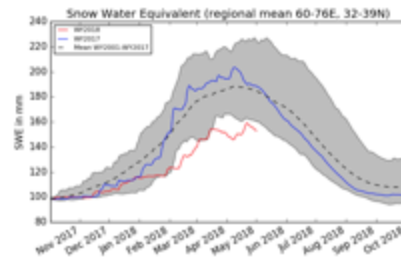
The Famine Early Warning Systems Network is a leading provider of early warning and analysis on food insecurity. Created by USAID in 1985 to help decision-makers plan for humanitarian crises, FEWS NET provides evidence-based analysis on some 35 countries. Implementing team members include NASA, NOAA, USDA, and USGS, along with Chemonics International Inc. and Kimerica.

LIS Visualization Tool: LISATLAS

Related Links: FEWSNET products, FEWSNET on LDAS



Central Asia - Snow Water Equivalent time series with climatology



<https://lis.gsfc.nasa.gov/projects/fewsnet>

References

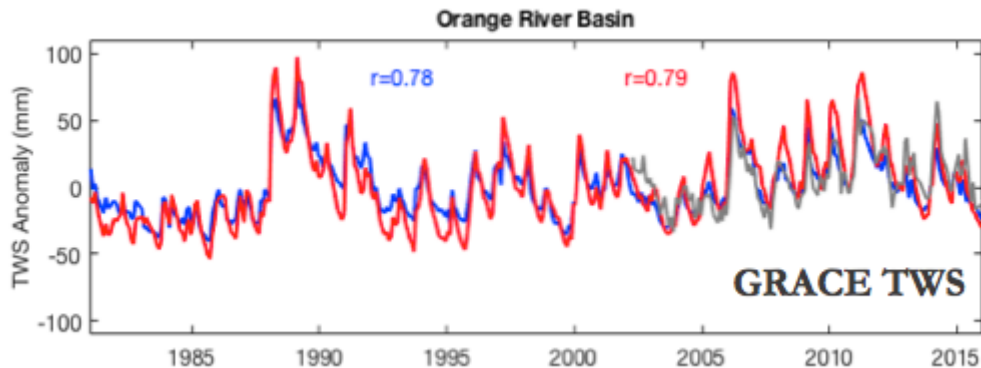
- [1] Rodell, Matthew, et al. "The observed state of the water cycle in the early twenty-first century." *Journal of Climate* 28.21 (2015): 8289-8318.
- [2] Kumar, S. *et al.* Land information system: An interoperable framework for high resolution land surface modeling. *Environmental Modelling and Software* 21, 1402–1415 (2006).
- [3] McNally, A., et al. (2017). A land data assimilation system for sub-Saharan Africa food and water security applications. *Scientific Data*, 4, 170012.
- [4] Funk, C. *et al.* The climate hazards infrared precipitation with stations—a new environmental record for monitoring extremes. *Scientific Data* 2, 150066 (2015).
- [5] Bosilovich, M. *et al.* MERRA-2: Initial Evaluation of the Climate. *NASA Technical Report Series on Global Modeling and Data Assimilation, NASA/TM-2015 39*, 136 (2015).
- [6] Senay, Gabriel B., et al. "Operational evapotranspiration mapping using remote sensing and weather datasets: A new parameterization for the SSEB approach." *JAWRA Journal of the American Water Resources Association* 49.3 (2013): 577-591.
- [7] Verdin, K.L., 2017, Hydrologic Derivatives for Modeling and Analysis—A new global high-resolution database: U.S. Geological Survey Data Series 1053, 16 p., <https://doi.org/10.3133/ds1053>.
- [8] JRC's GHS_POP_GPW4_GLOBE_R2015A Population data
- [9] Linard, Catherine, et al. "Population distribution, settlement patterns and accessibility across Africa in 2010." *PloS one* 7.2 (2012): e31743.
- [10] Falkenmark, Malin. "The massive water scarcity now threatening Africa: why isn't it being addressed?." *Ambio* (1989): 112-118.

FLDAS Africa evaluation:

2016. McNally, A., S. Shukla, K. Arsenault, S. Wang, C. Peters-Lidard, J.P. Verdin: Evaluating ESA CCI soil moisture in East Africa. *International Journal of Applied Earth Observation*. doi:10.1016/j.jag.2016.01.001 [-Abstract](#)
2017. McNally, Amy, Kristi Arsenault, Sujay Kumar, Shradhanand Shukla, Pete Peterson, Shugong Wang, Chris Funk, Christa D. Peters-Lidard, and James P. Verdin. "A land data assimilation system for sub-Saharan Africa food and water security applications." *Scientific Data* 4. doi: [10.1038/sdata.2017.12](https://doi.org/10.1038/sdata.2017.12)
2017. Jung, H. C., Getirana, A., Policelli, F., **McNally, A.**, Arsenault, K. R., Kumar, S., ... & Peters-Lidard, C. D. Upper Blue Nile Basin Water Budget from a Multi-Model Perspective. *Journal of Hydrology*. <https://doi.org/10.1016/j.jhydrol.2017.10.040>

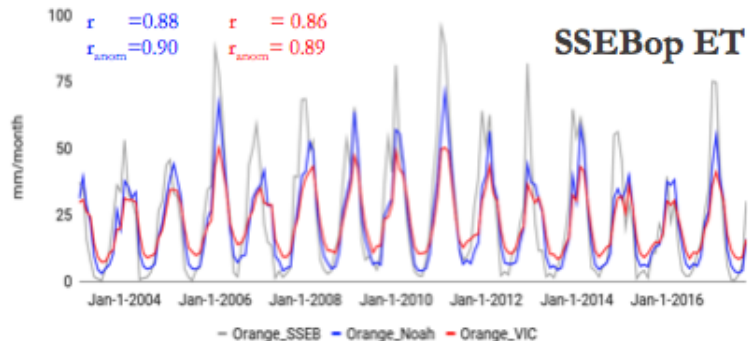
FLDAS Model Evaluation

Southern Africa - Orange Basin

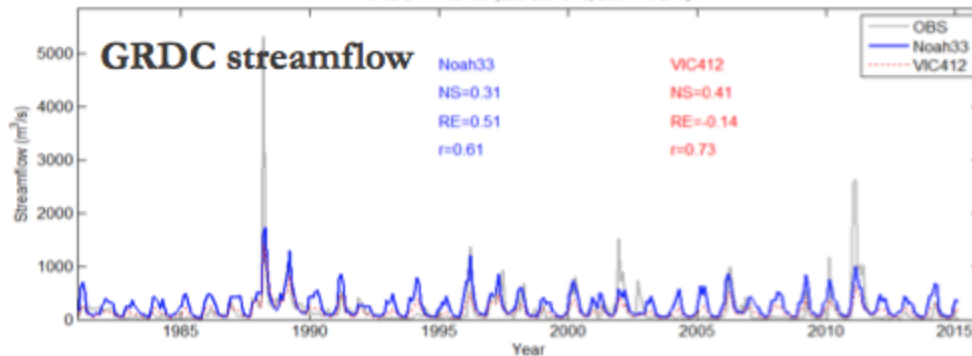


FLDAS Noah33 & VIC412 estimates are well correlated with GRACE terrestrial water storage (TWS), SSEBop evapotranspiration (ET) and in situ streamflow in the Orange Basin, Southern Africa

Orange Basin: SSEBopV4, Noah33 and VIC412

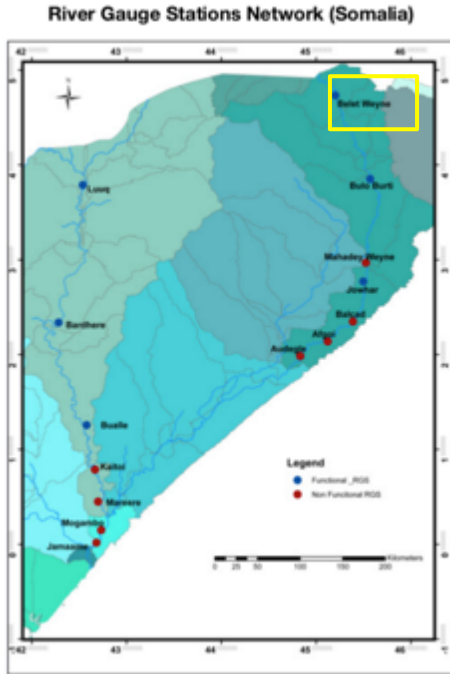


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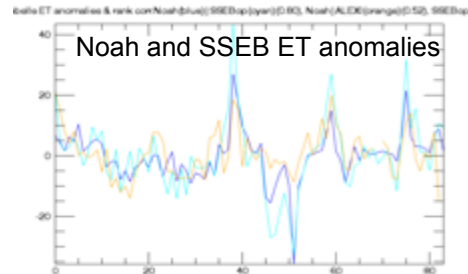
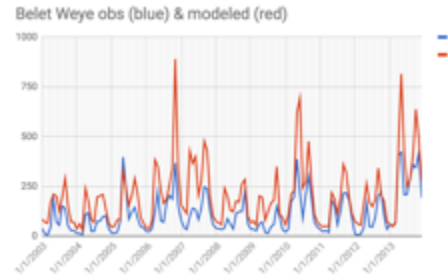
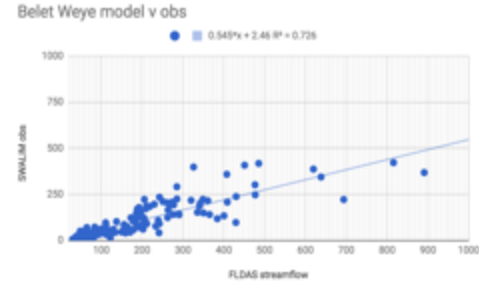


FLDAS Model Evaluation

Eastern Africa - Shabelle Basin



Gauges from FAO's *Somalia Water and Land Information Management (SWALIM)* <http://sddr.faoswalim.org/>



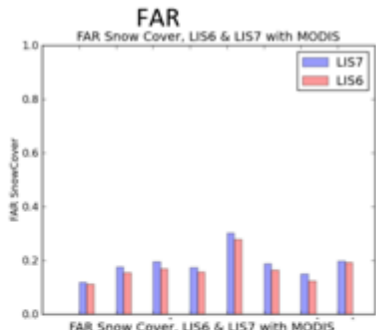
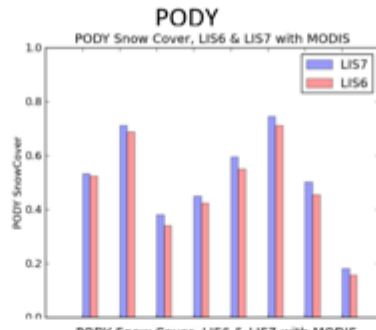
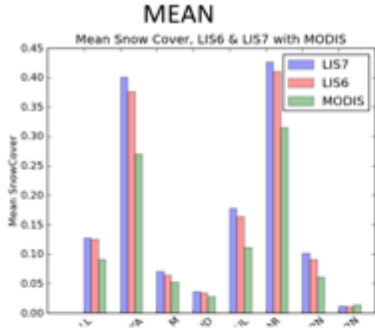
FLDAS Noah33 streamflow estimates are well correlated ($R=0.73$) with in situ streamflow from SWALIM. Correlations decline downstream, likely due to human abstractions.

Central Asia Evaluation

Mean snow cover fraction corresponds well at the basin level during different water years.

When considering a full water year, probability of Detection (POD) metrics are moderate and False Alarm Rate (FAR) is relatively low. Cloud contamination likely contributes to poor statistics.

WY2016



WY2015

