

Monitoring Evapotranspiration and Drought using Thermal Remote Sensing



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TOOLS

ALEXI

Multi-scale ET
modeling

DMS

Thermal image
sharpening

STARFM

Multi-sensor data
fusion

ASSETS

GEO

Hourly SW/TIR
 5km/5km

MODIS

Daily 250m/1km

Landsat

16 day 30m/100m

Lsat-like

~20-60m/ --

APPLICATIONS

(daily/30 m to 10 km)



Crop water use
(Evapotranspiration)



Crop phenology
metrics



Crop stress
(drought early warning)

NEW ASSET

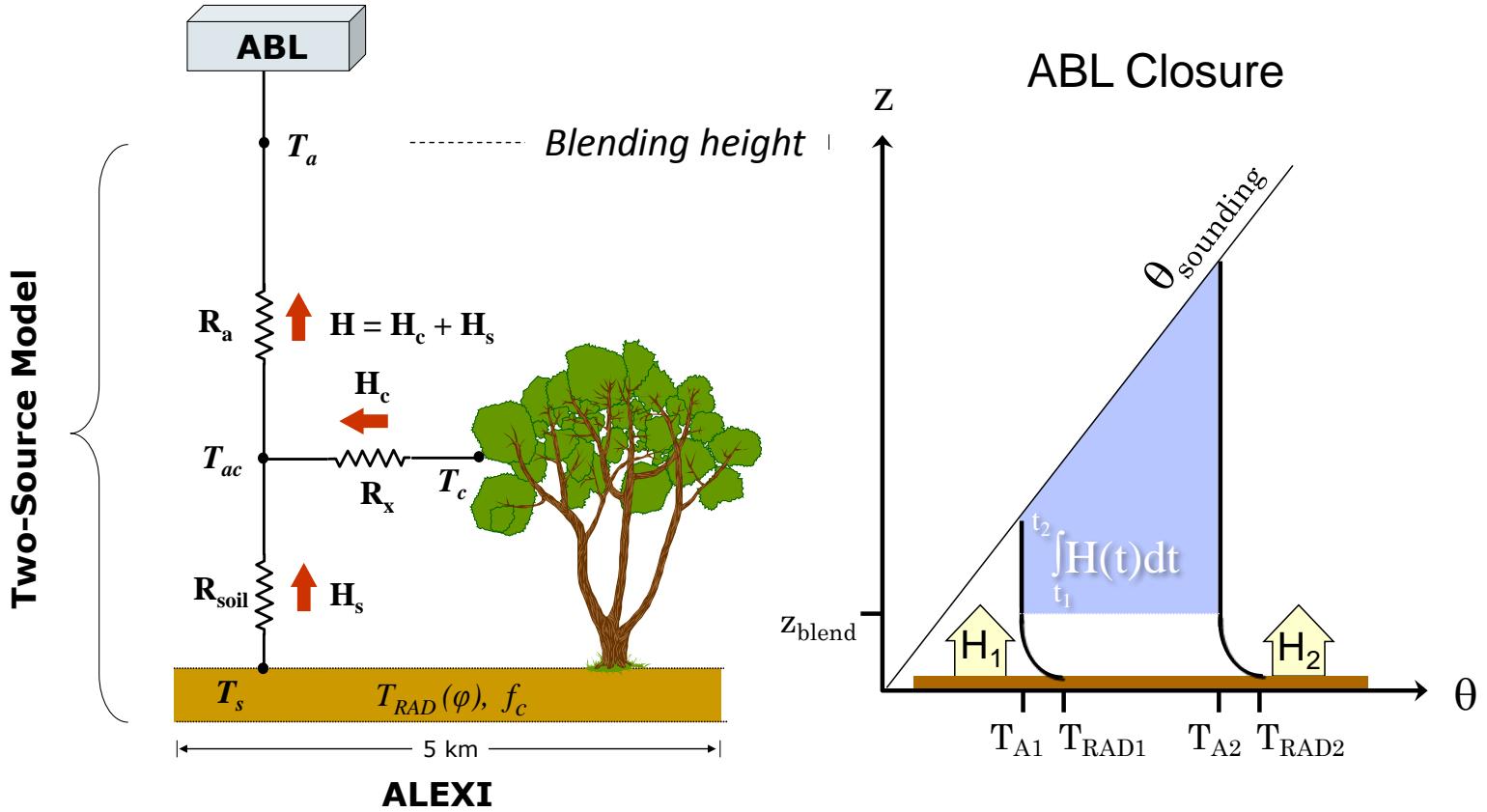
Ka LST

MW
All-sky > 5km

The background of the image is a high-resolution aerial photograph of agricultural land. The fields are organized into a grid-like pattern, with different colors representing various crops or soil types. Large, dark green circular shapes are overlaid on the image, covering approximately one-third of the total area. These circles are irregular in size and position, creating a textured, abstract effect.

MULTI-SCALE SATELLITE ET RETRIEVAL

... based on land surface temperature

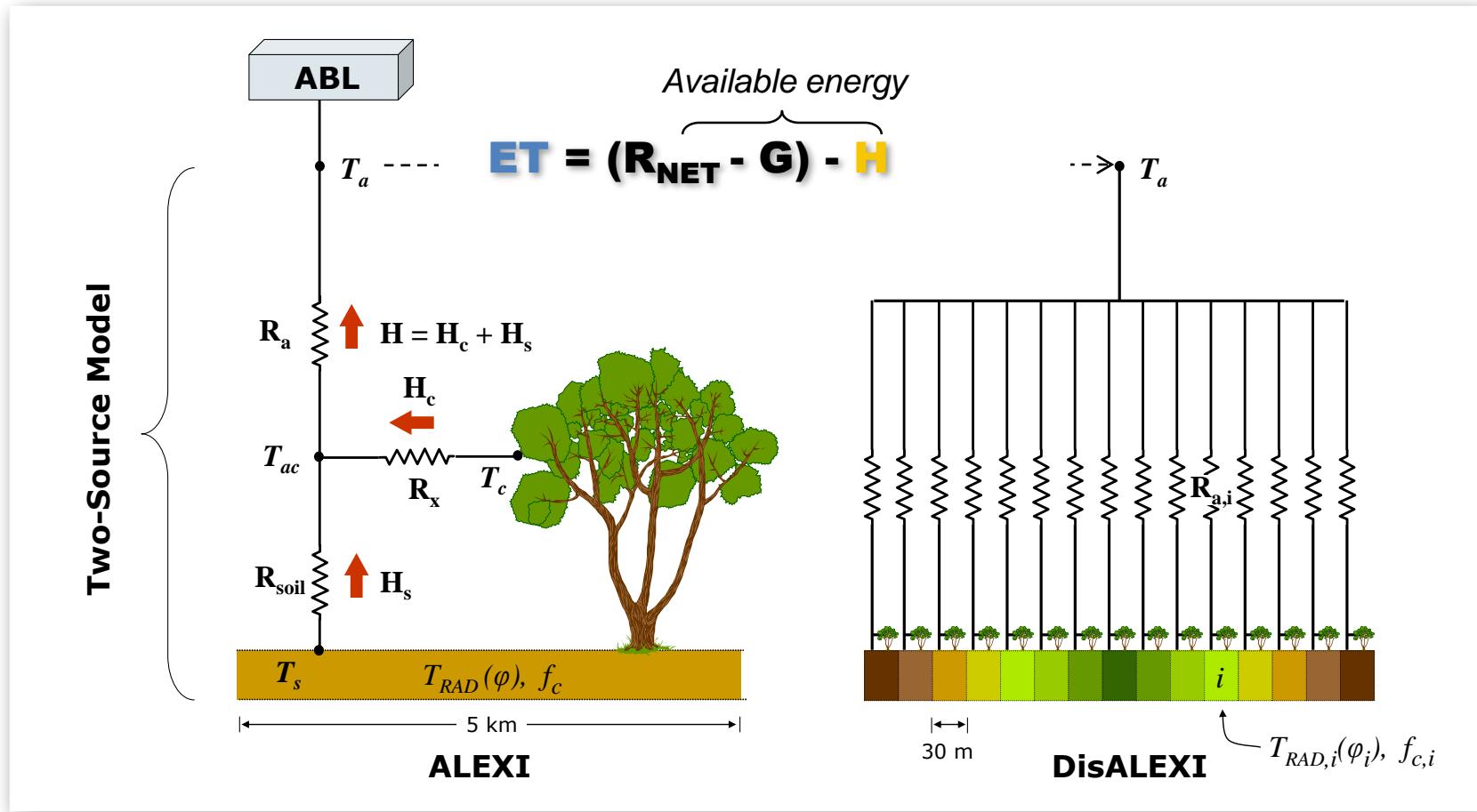


Regional scale

Surface temp: ΔT_{RAD} - Geostationary

Air temp: T_a - Diagnosed by ABL model

Given known radiative energy inputs, how much water loss is required to keep the soil and vegetation at the observed temperatures?



Regional scale

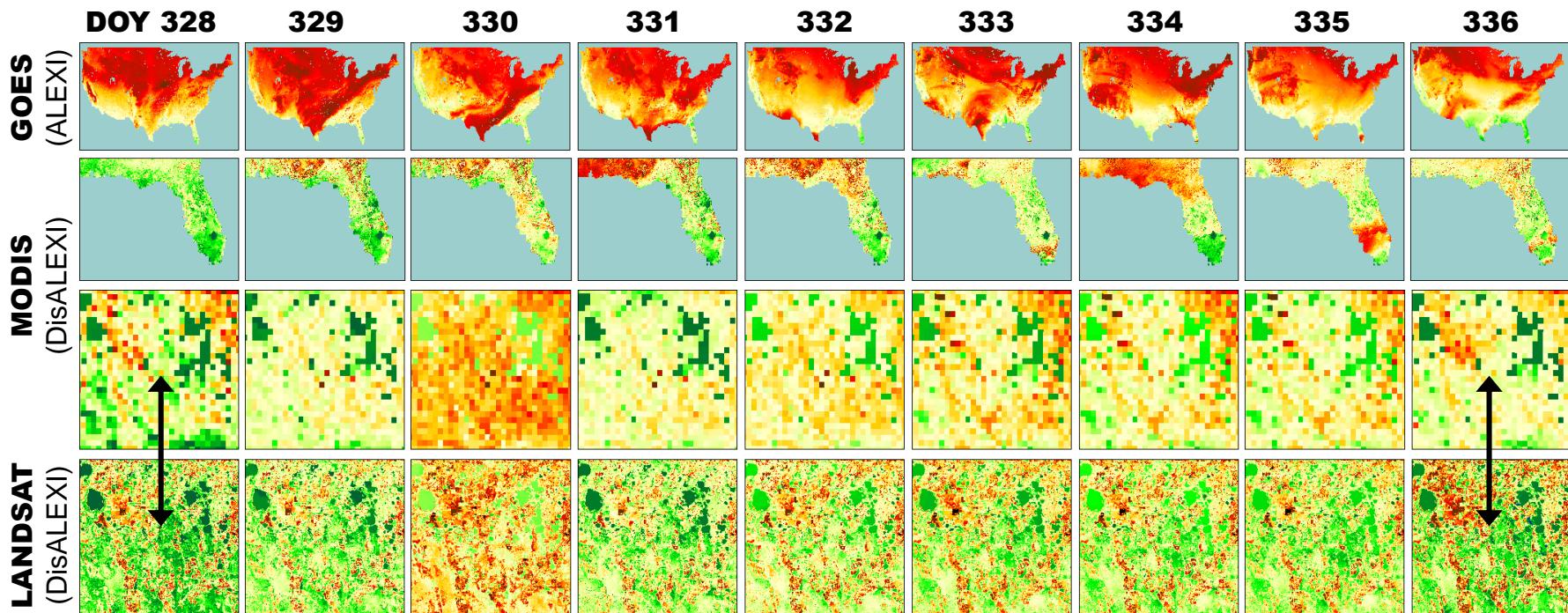
Surface temp: ΔT_{RAD} - Geostationary
 Air temp: T_a - ABL model

Landscape scale

T_{RAD} - Landsat, MODIS
 T_a - ALEXI

GOES/MODIS/Landsat FUSION

Daily Evapotranspiration – Orlando, FL, 2002



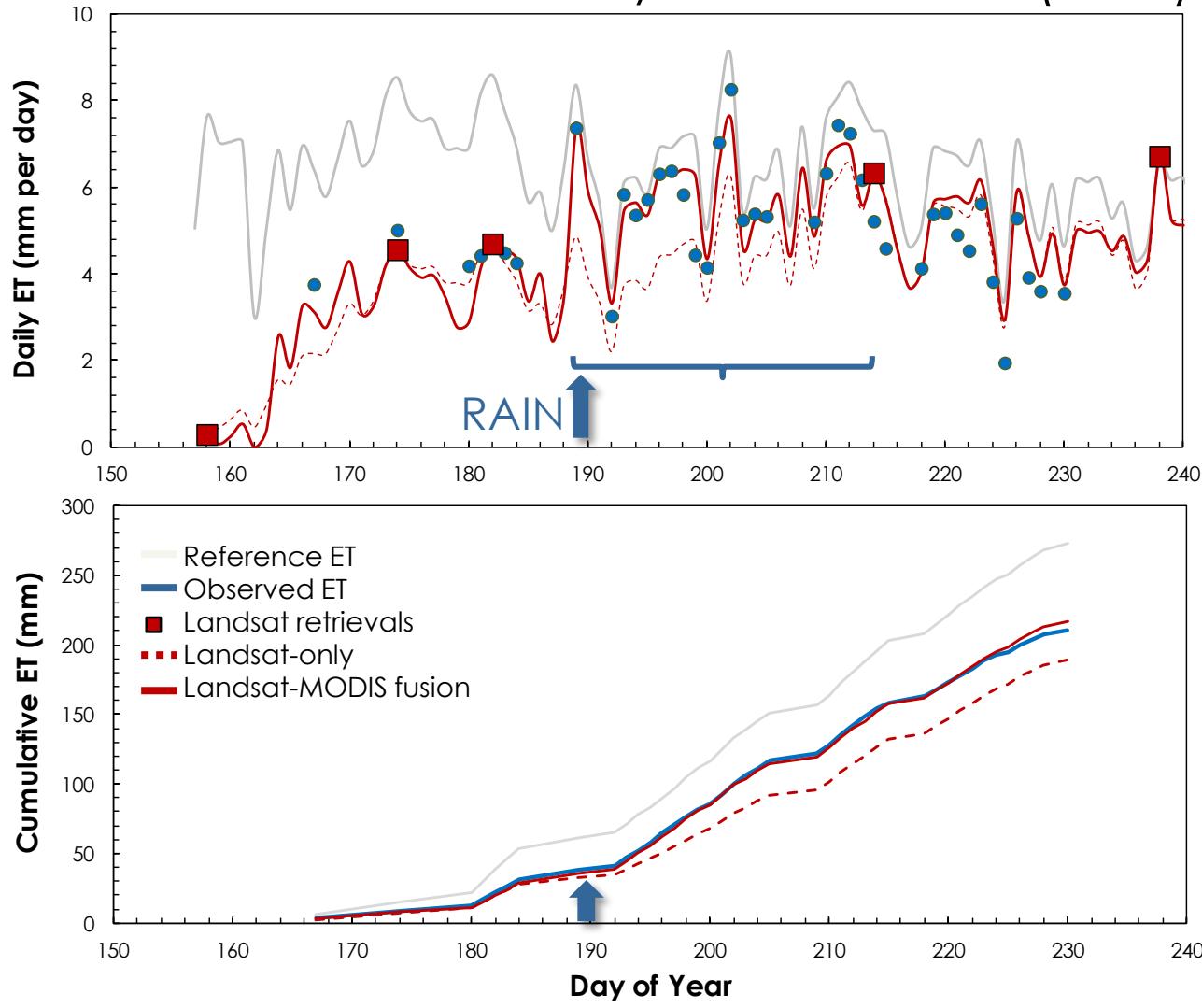
Landsat 5

Landsat 7

**Spatial Temporal Adaptive Reflectance Fusion Model
(STARFM)** *(Gao et al, 2006)*

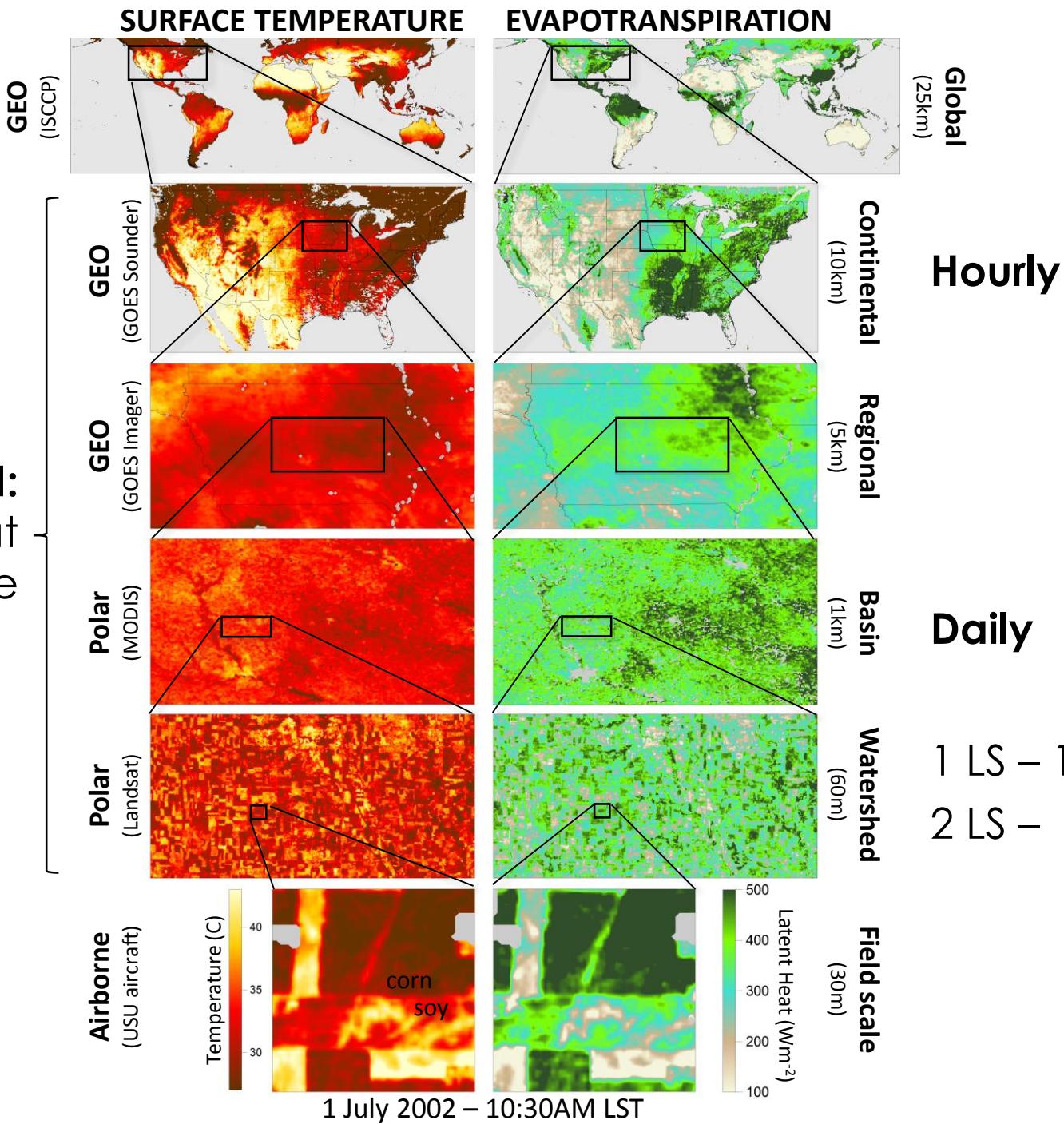
Validation using flux tower data

Rainfed soybean – SMEX02 (Iowa)

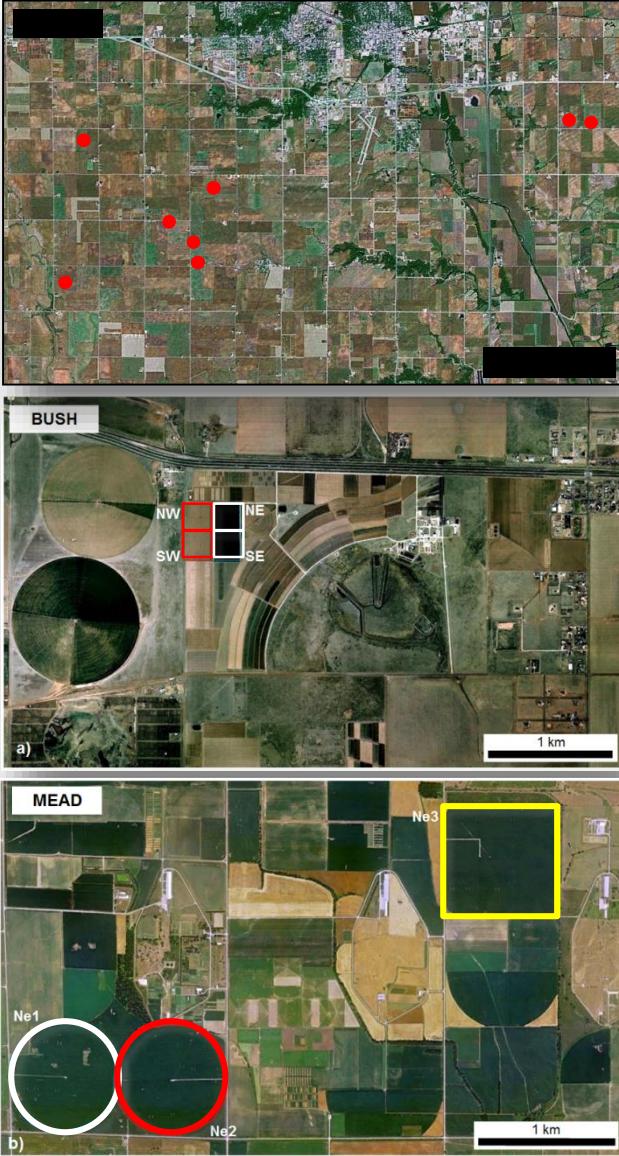


DATA FUSION:

daily ET at field scale



Evaluation of fused ET fluxes



SMEX02

Soil Moisture Experiment 2002
Ames, Iowa
Rainfed corn and soybean

BEAREX08

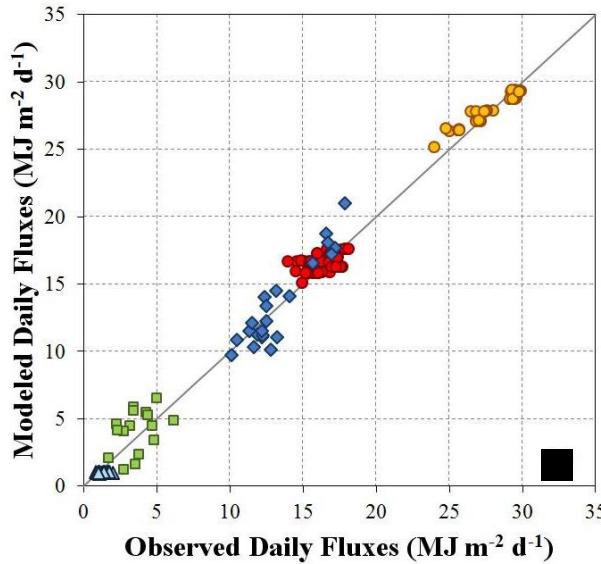
Bushland ET and Remote sensing Experiment 2008
Bushland, Texas
Rainfed and irrigated cotton

MEAD

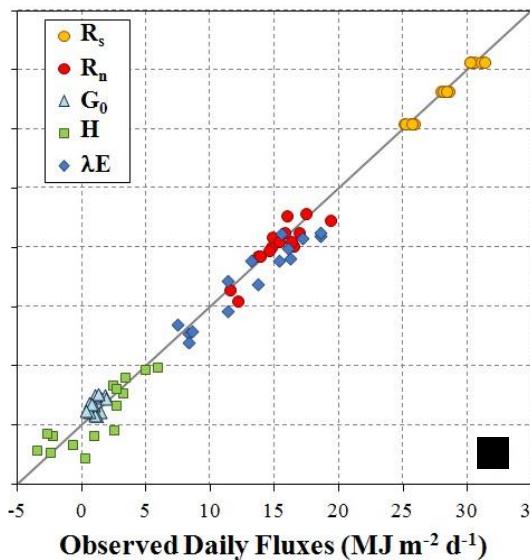
Ameriflux site (S. Verma)
Mead, NE
Rainfed and irrigated corn and soybean

Model performance on Landsat dates

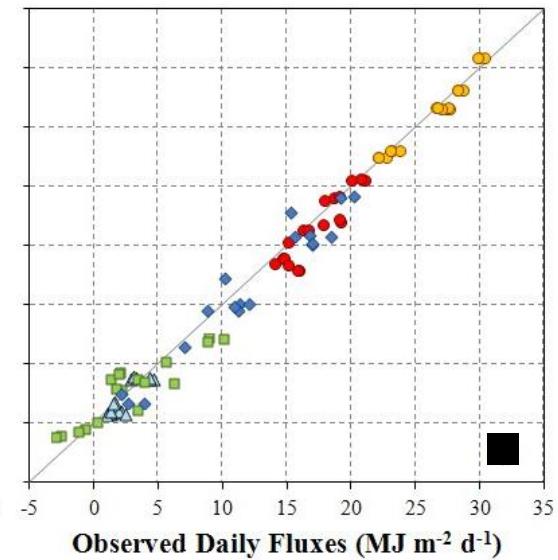
SMEX02



BEAREX08



MEAD



MAE: $1.08 \text{ MJ m}^{-2} \text{ d}^{-1}$

RE: 8%

$1.3 \text{ MJ m}^{-2} \text{ d}^{-1}$

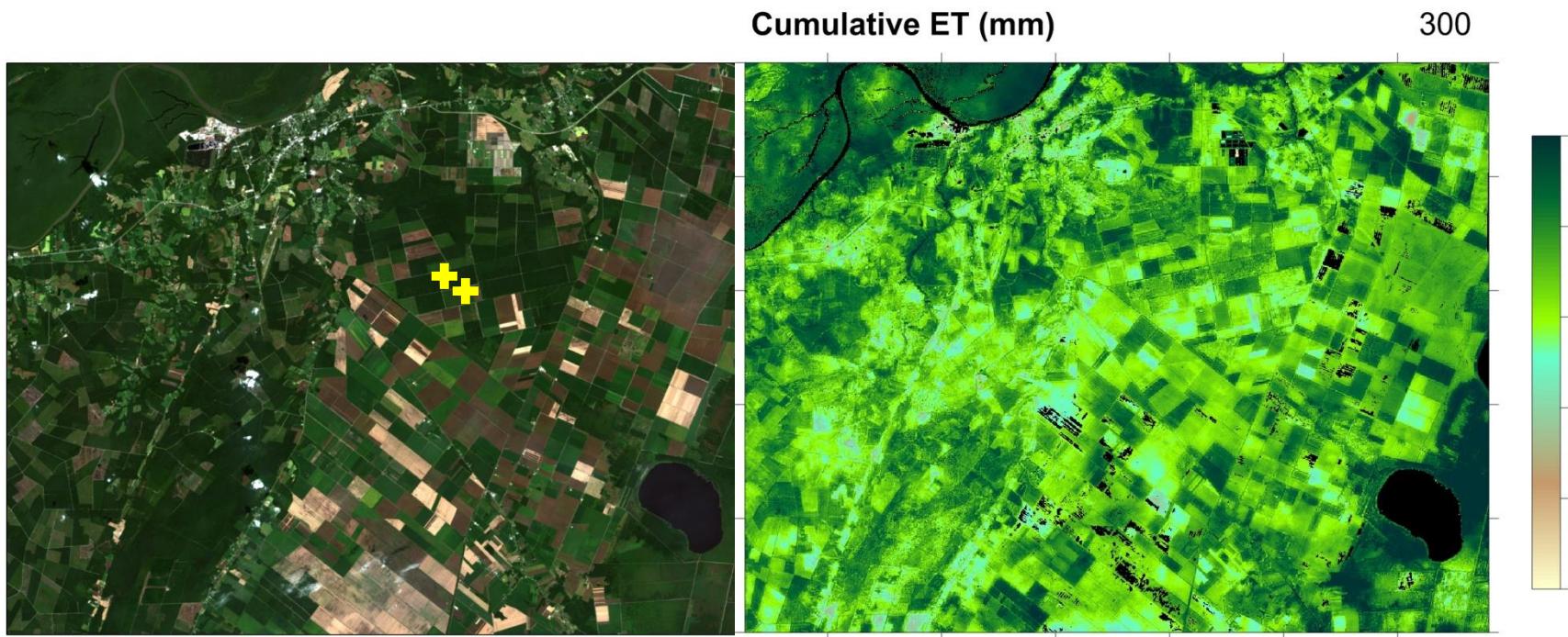
10%

$1.3 \text{ MJ m}^{-2} \text{ d}^{-1}$

11%

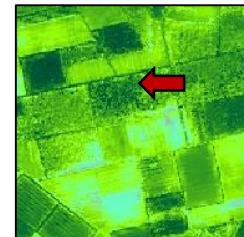
Loblolly Pine Plantation, NC

Effect of forest management

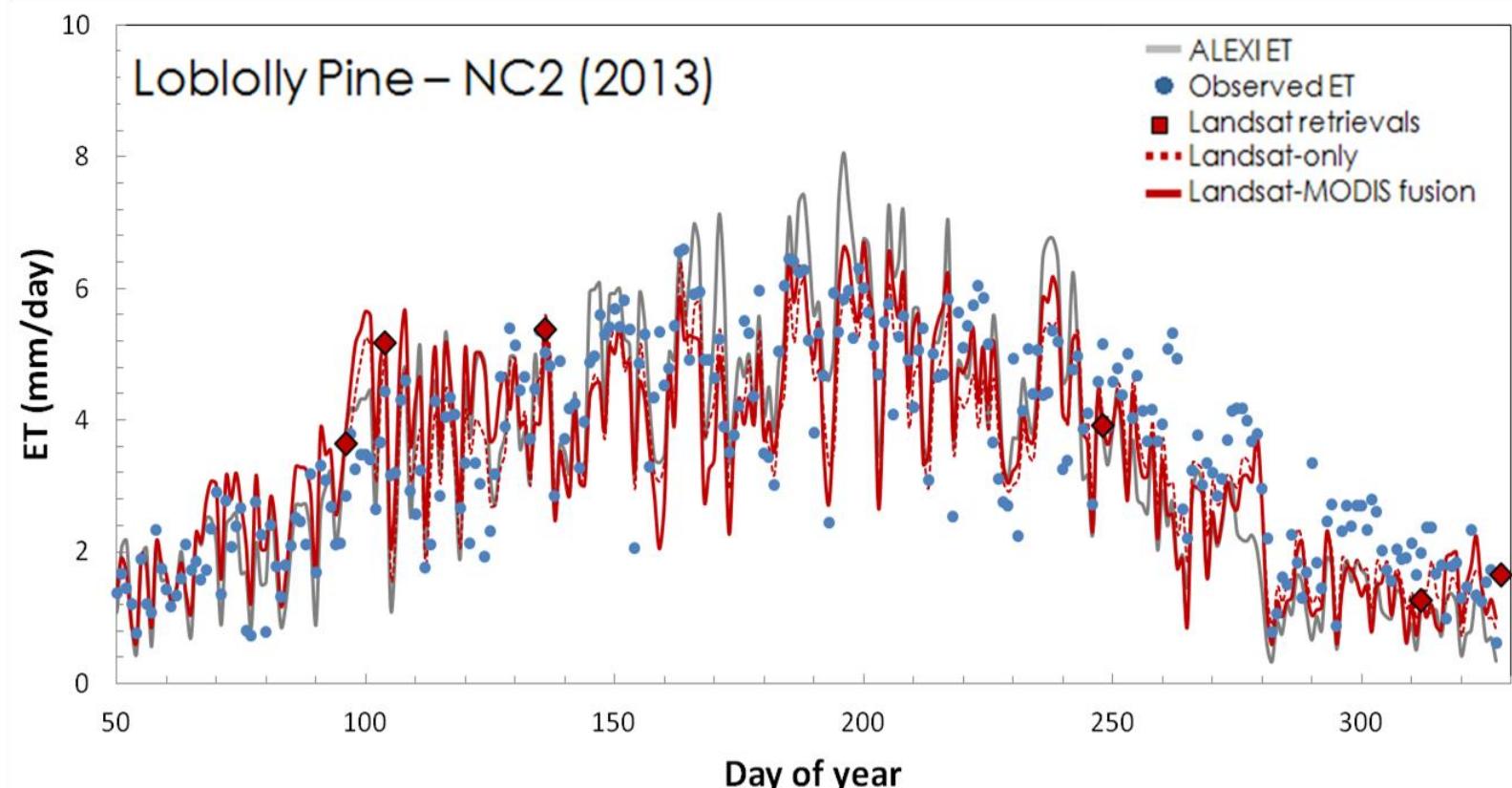


Landsat 8 - 2013

- ALEXI ET
- Observed ET
- Landsat retrievals
- Landsat-only
- Landsat-MODIS fusion



NC2:
Loblolly Pine (2013)



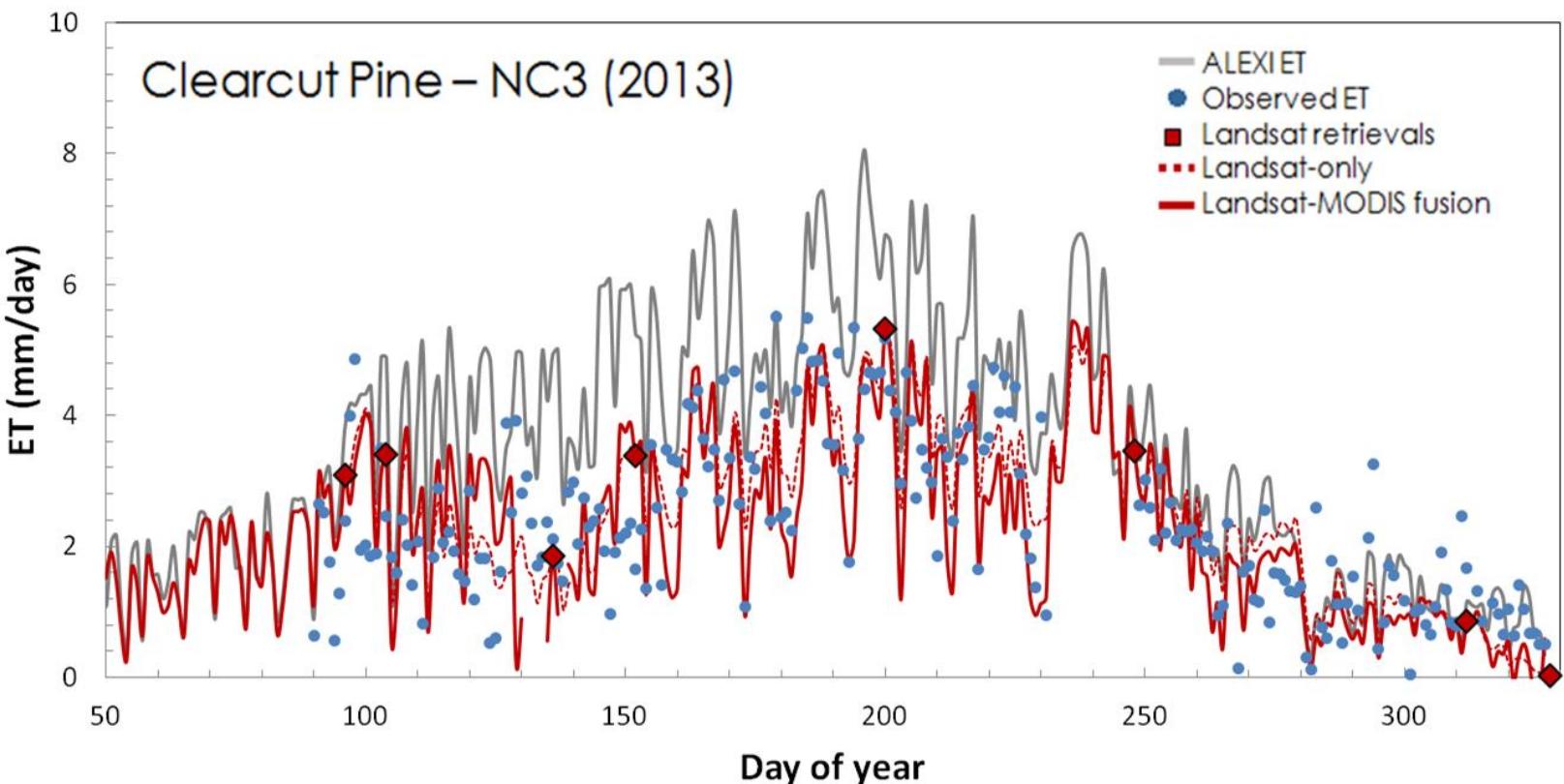
(Yang et al., 2015)

Effect of forest management

- ALEXI ET
- Observed ET
- Landsat retrievals
- Landsat-only
- Landsat-MODIS fusion

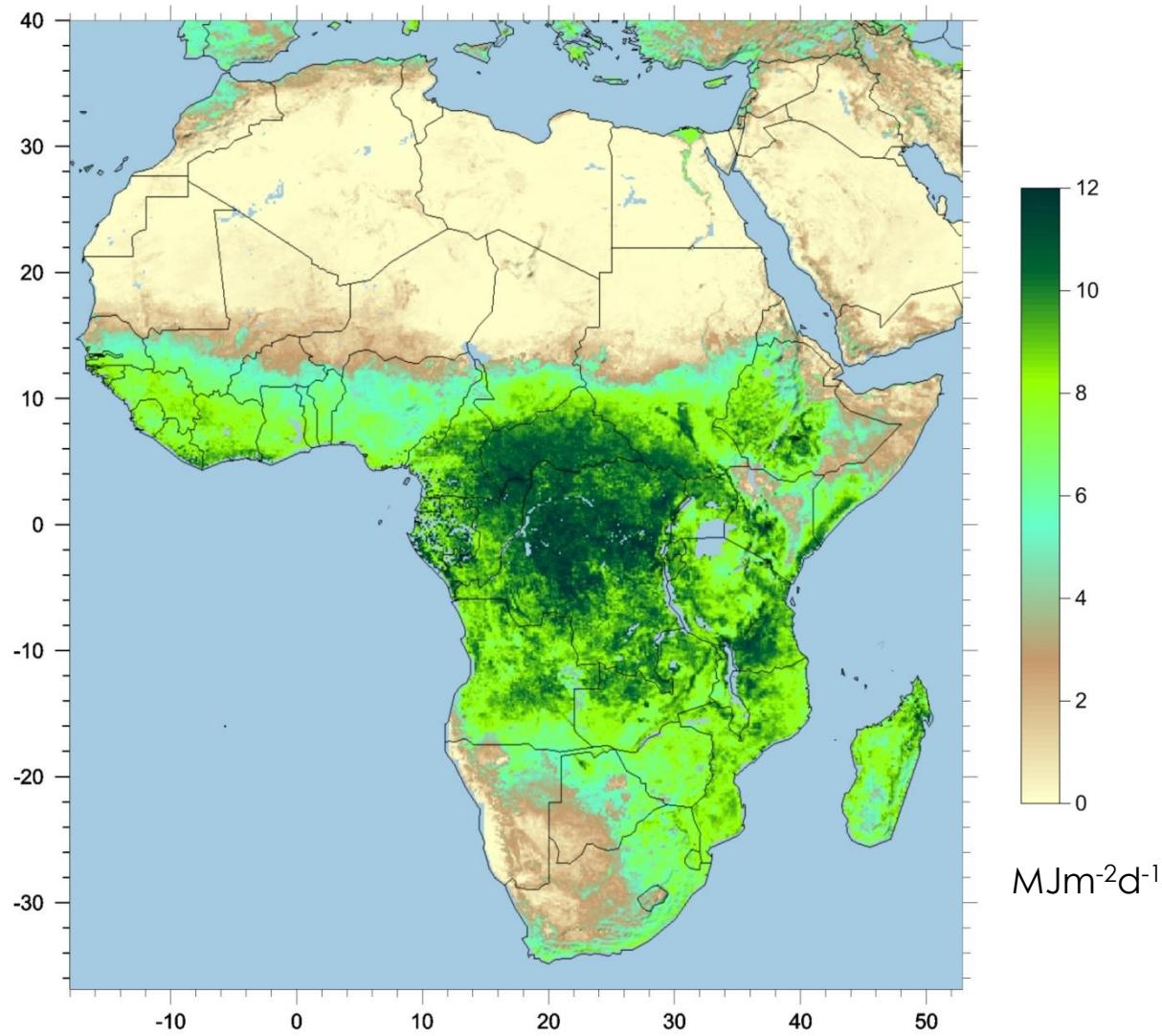


NC3:
Clearcut Pine (2013)

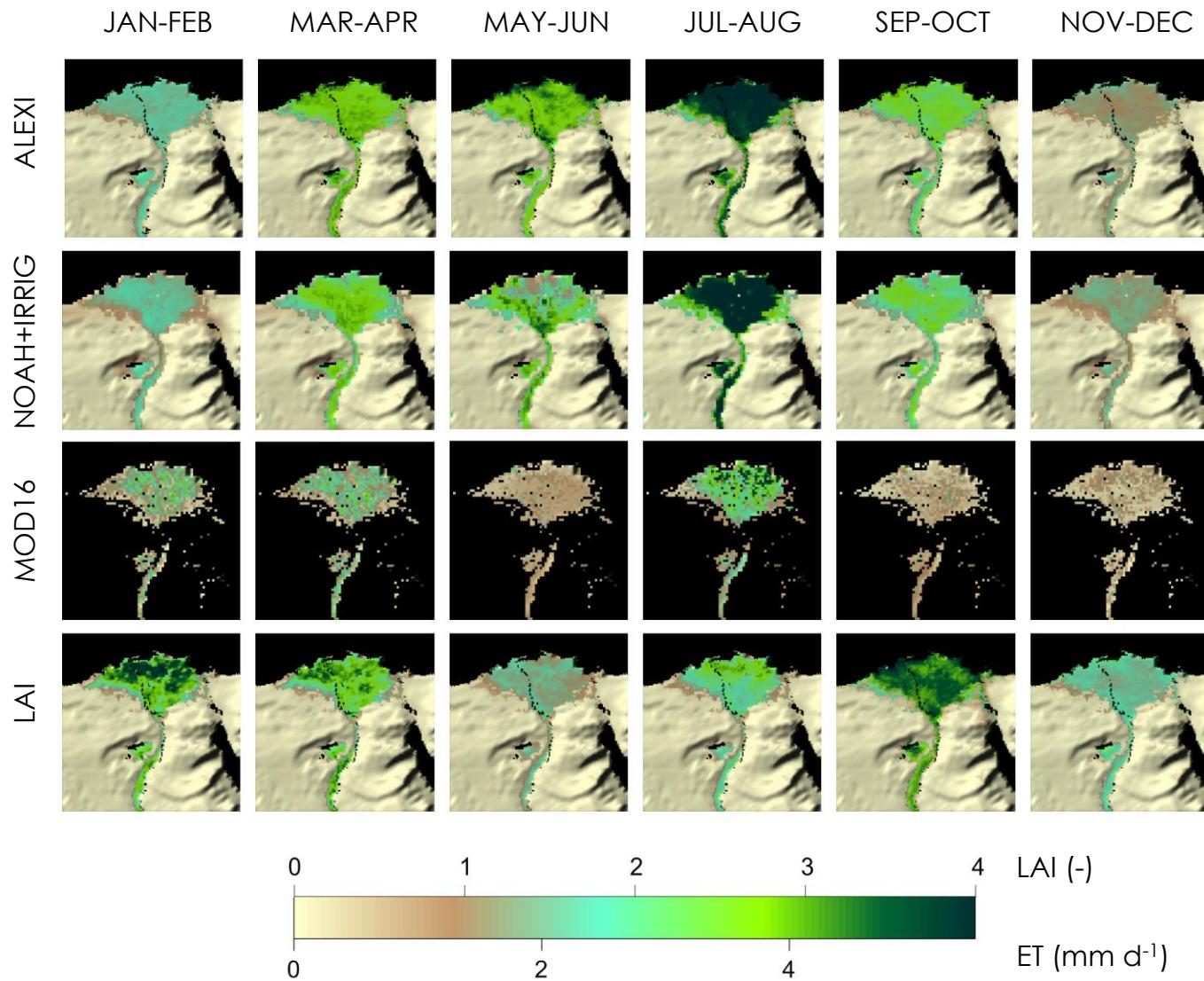


(Yang et al., 2015)

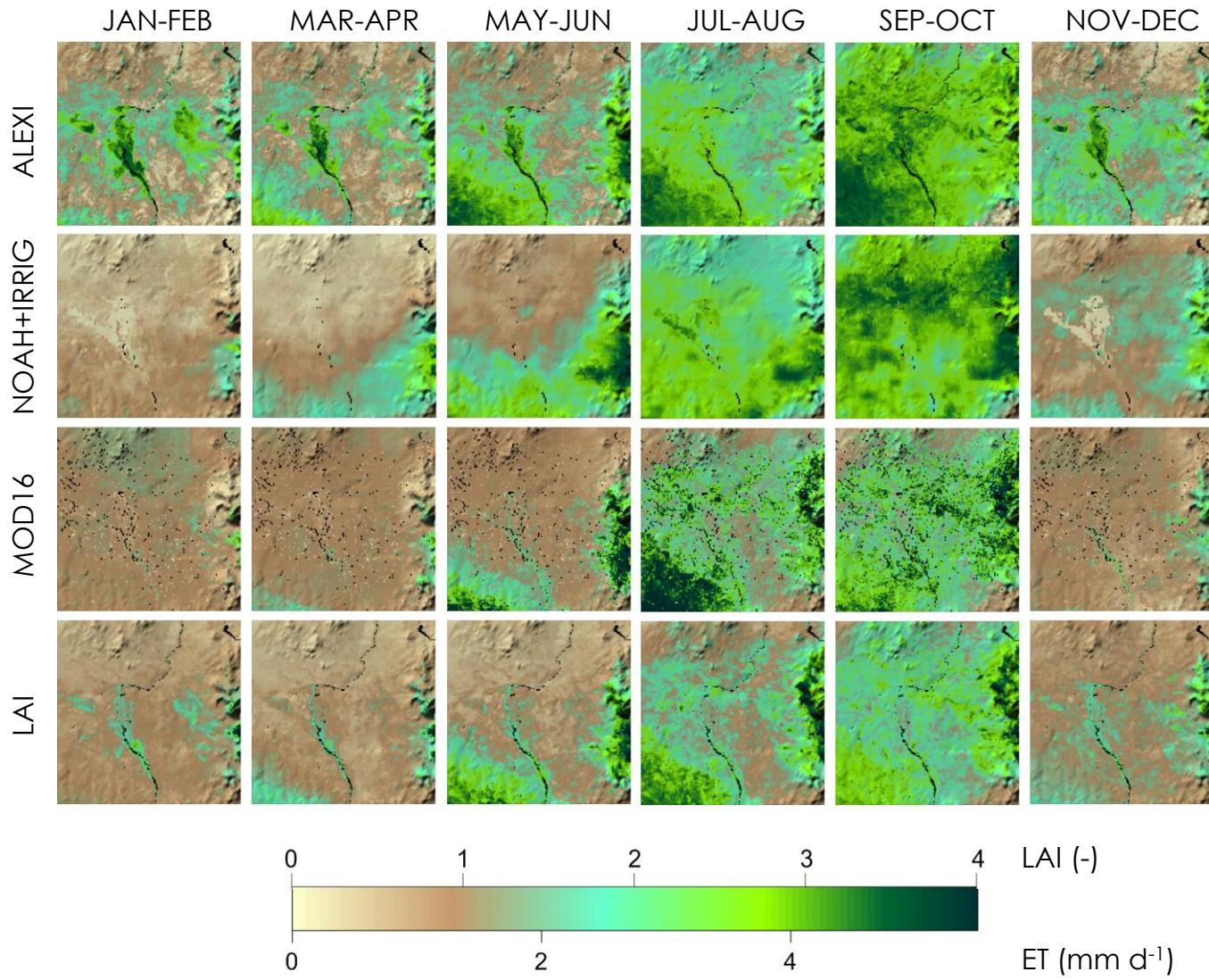
Annual composite of daily latent heat flux for 2013



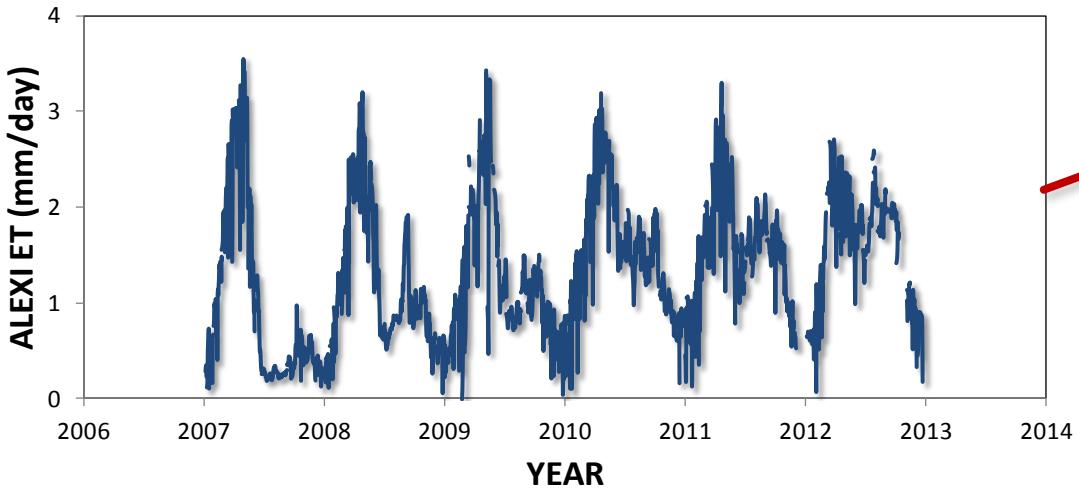
NILE DELTA



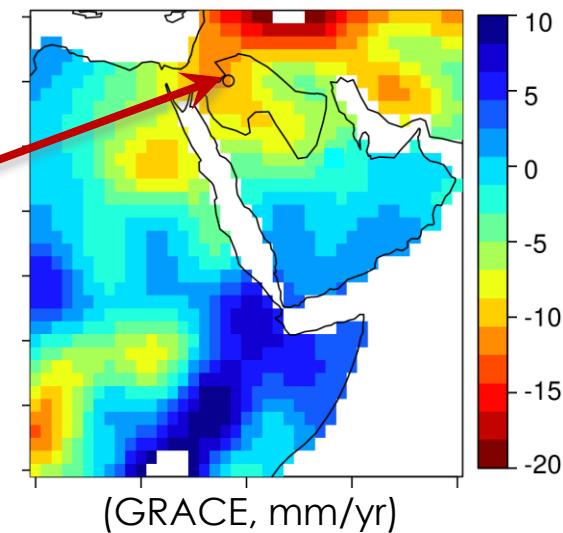
SUDD WETLAND



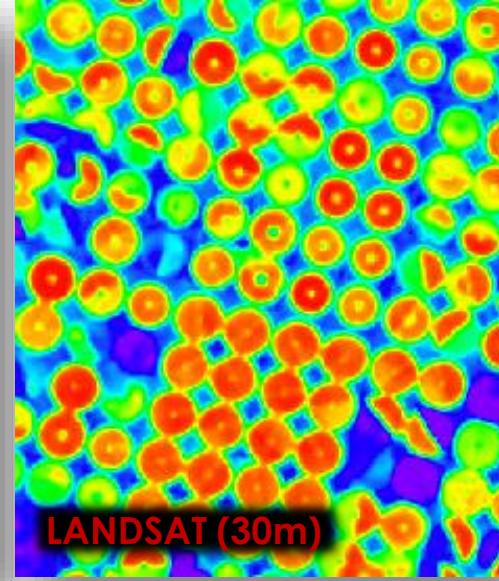
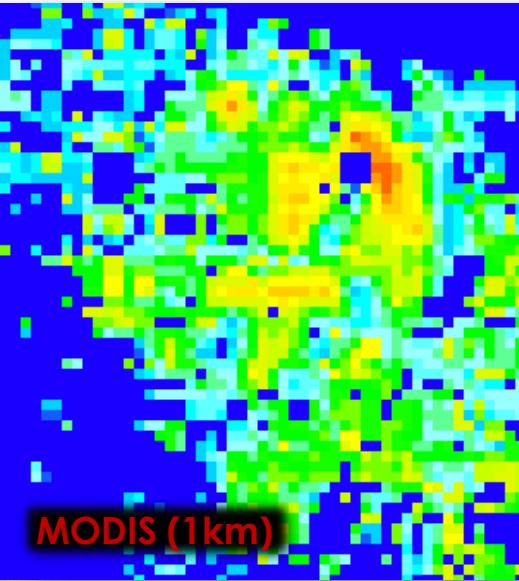
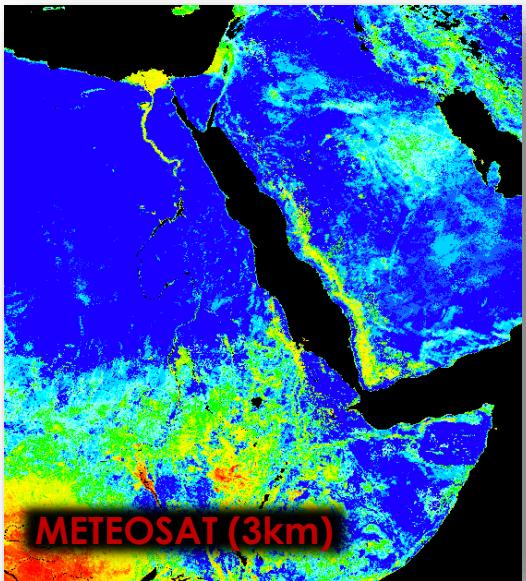
ALEXI ET – AL JOWF IRRIGATION SITE



GROUNDWATER DEPLETION TRENDS

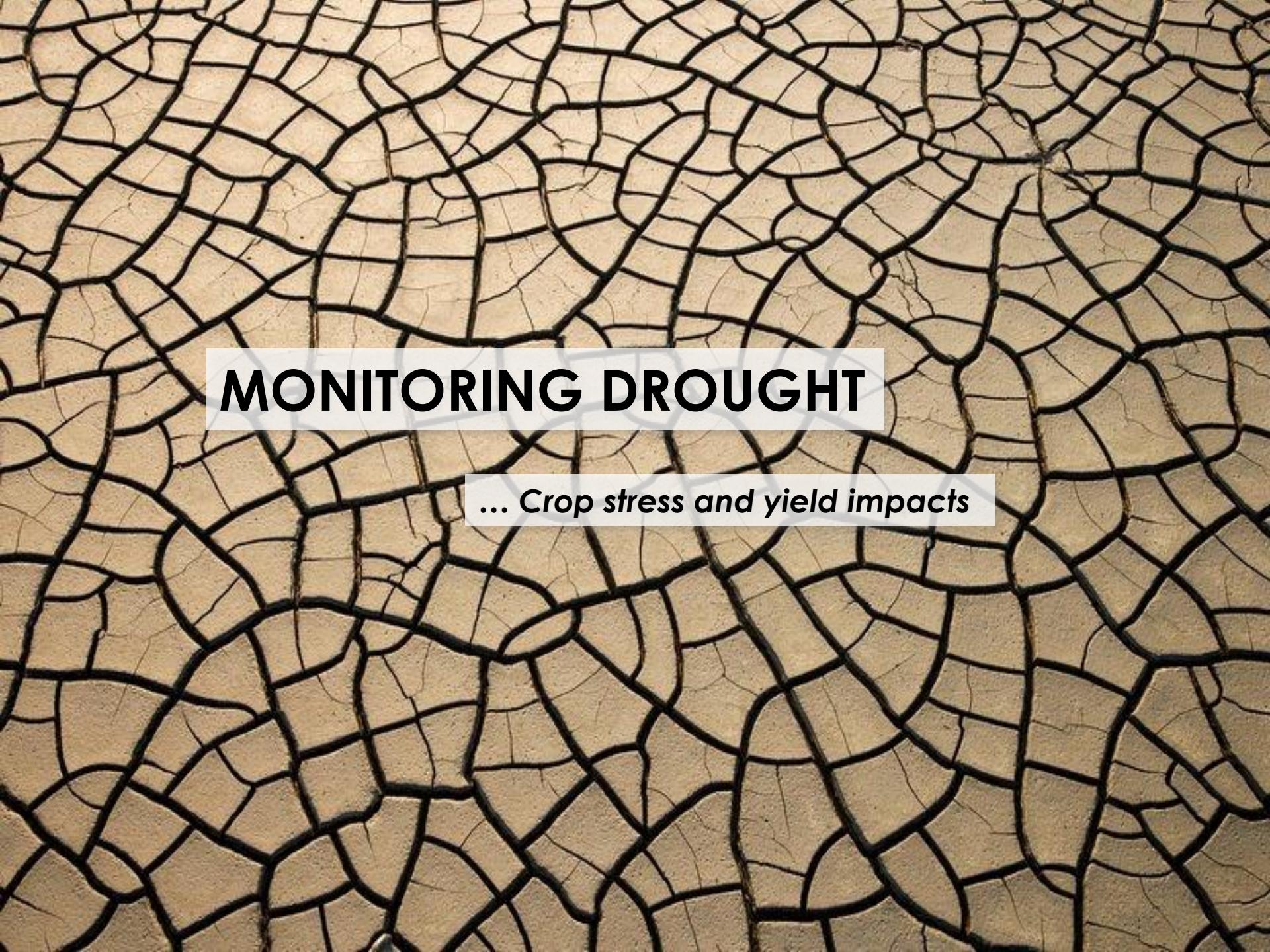


MULTI-SCALE ET MAPPING (mm/day)



Water use by irrigation

Houborg et al., 4th International Symposium of RAQRS, Valencia, Spain, Sept. 2014



MONITORING DROUGHT

... Crop stress and yield impacts

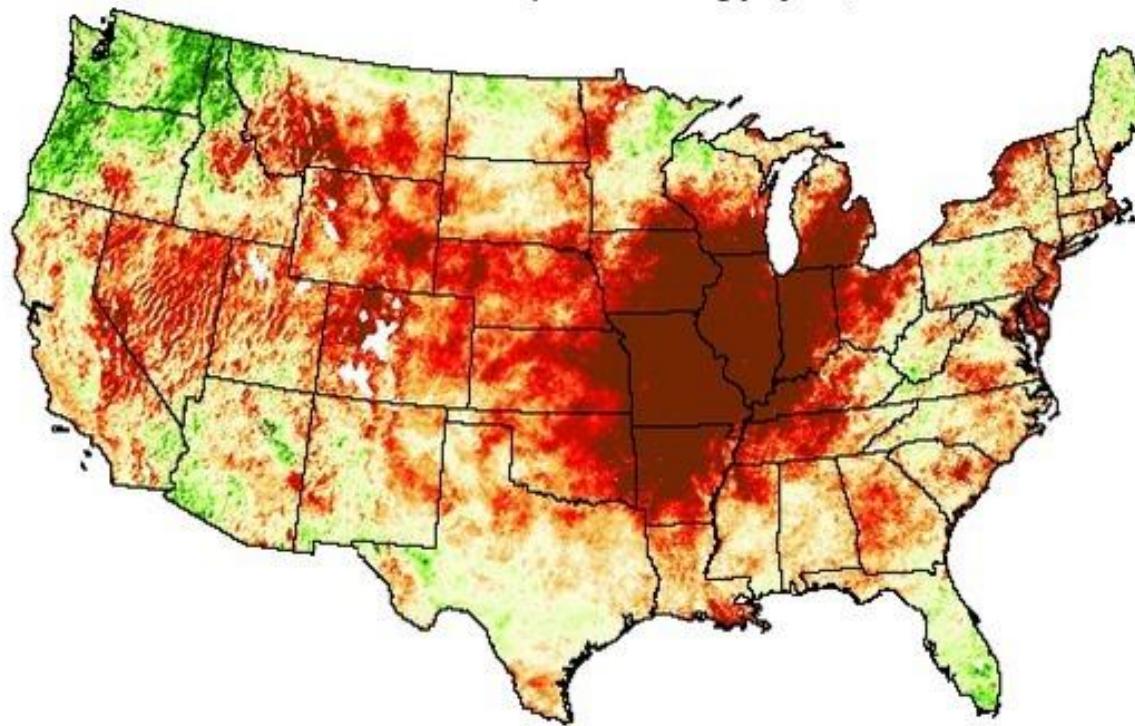
ESI Methodology

ALEXI ESI represents temporal anomalies in the ratio of actual ET to potential ET.

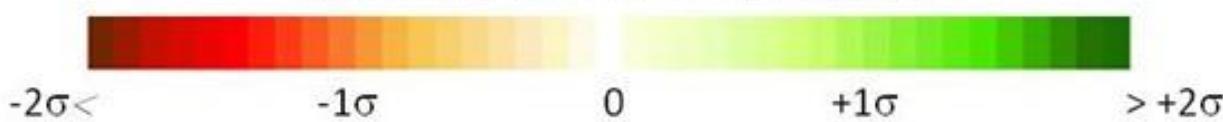
- ESI does not require precipitation data, ***the current surface moisture state is deduced directly from the remotely sensed LST***, therefore it may be more robust in regions with minimal in-situ precipitation monitoring.
- Signatures of vegetation stress are manifested in the LST signal before any deterioration of vegetation cover occurs, for as example as indicated in NDVI, so TIR-based indices such as ESI can provide an effective early warning signal of impending agricultural drought.
- ALEXI ESI inherently includes non-precipitation related moisture signals (such as irrigation; vegetation rooted to groundwater; lateral flows) that need to be modeled a priori in prognostic LSM schemes.
- ALEXI ESI provides an independent assessment of current drought conditions, supplementing precipitation and modeling-based indices – an invaluable resource to decision-makers who usually depend on a convergence of information in the decision making process.

Evaporative Stress Index 4km

3 month composite ending July 28, 2012

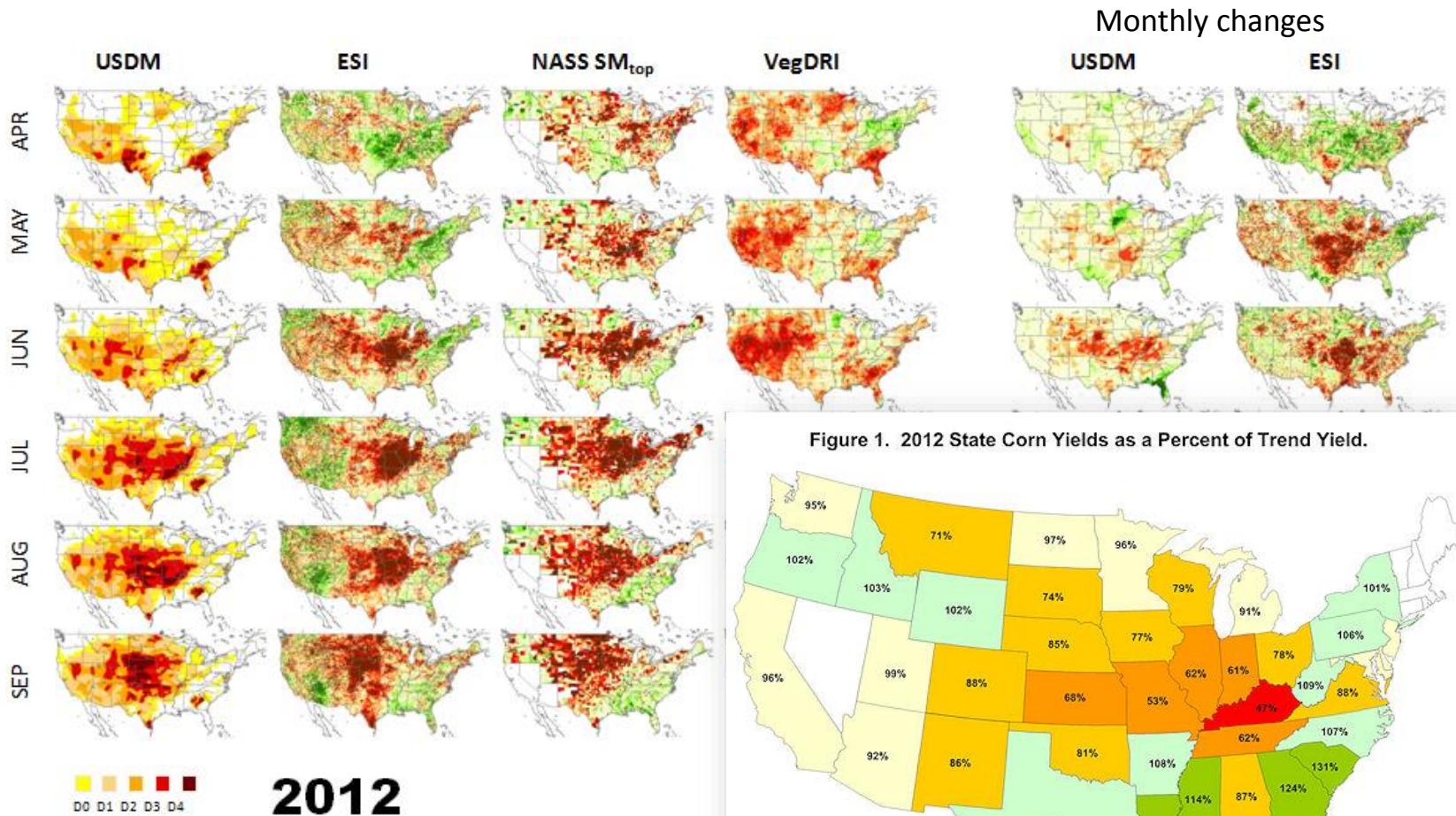


Standardized ET/PET anomalies



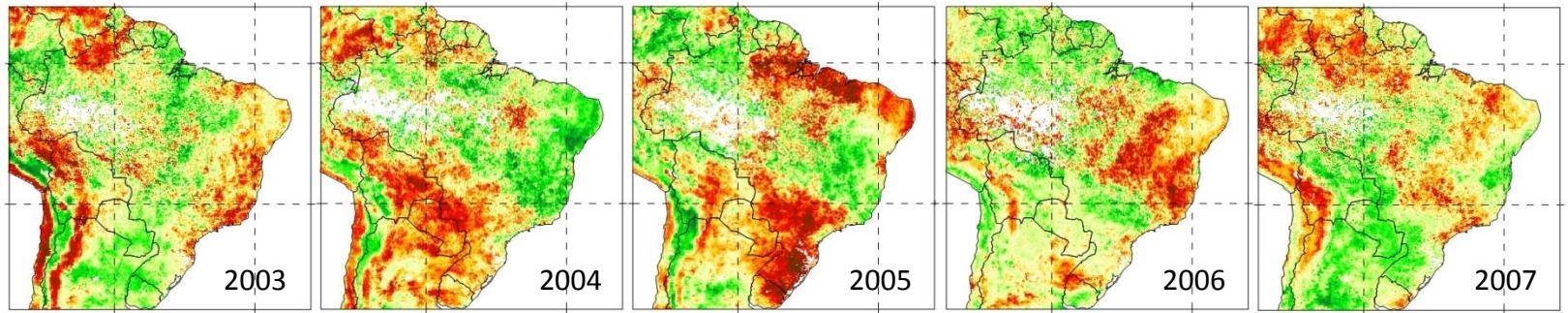
Atmosphere-Land Exchange Inverse Model (ALEXI)
(Anderson et al., 1997, 2007)

2012 FLASH DROUGHT



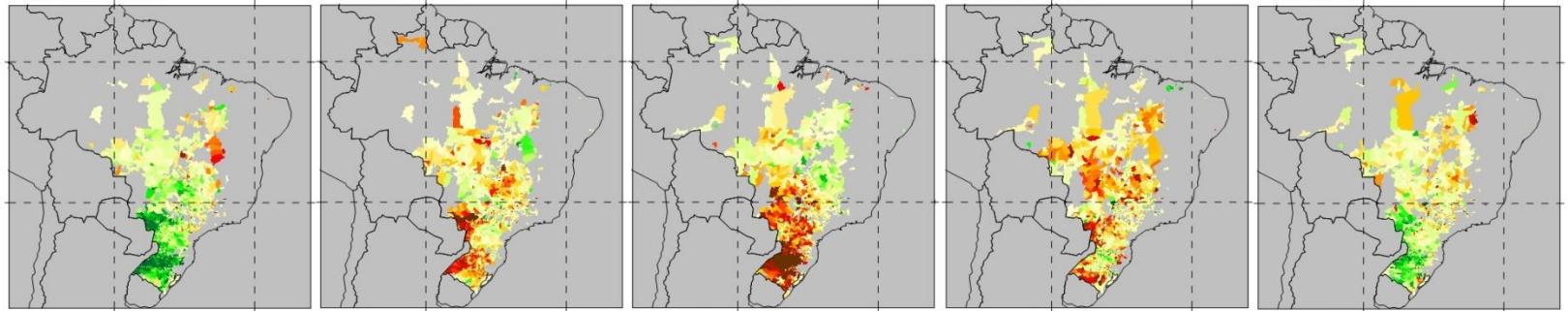
ANNUAL MUNICIPAL LEVEL SOYBEAN YIELD ANOMALIES

ESI (JFM)

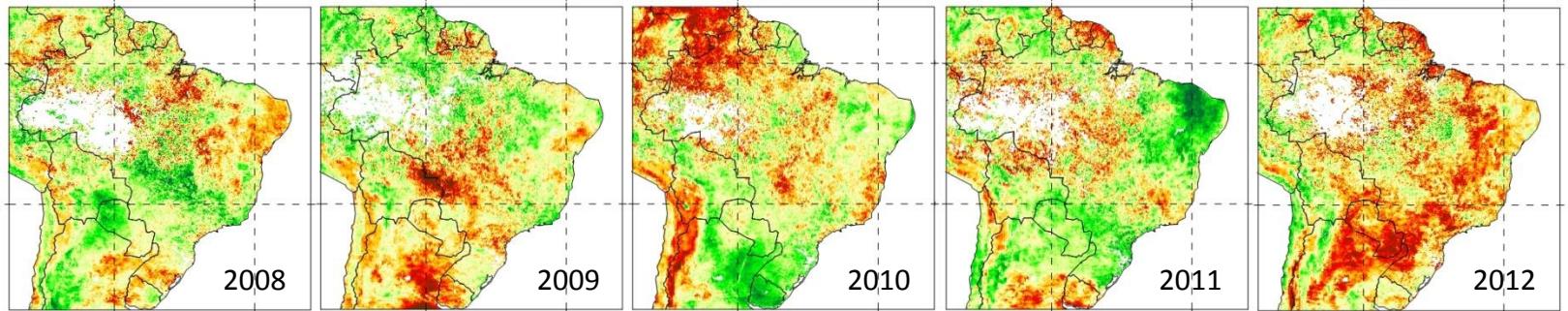


ESI (s)

Yield anomaly

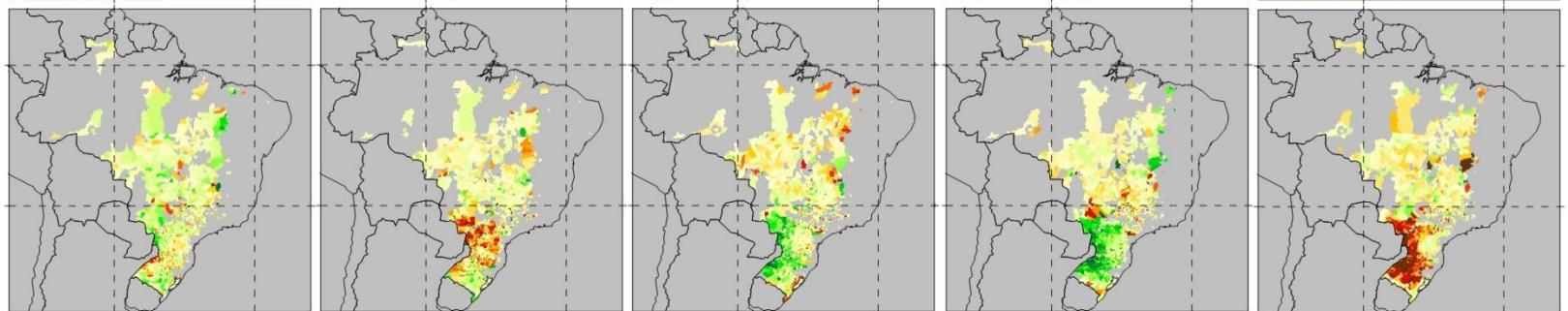


ESI (JFM)



Yield anomaly (kg/ha)

Yield anomaly



Advantages

- Diagnostically captures non-precipitation related moisture sources/sinks (irrigation, shallow groundwater, drainage)
- Capacity to map from global to sub-field scales using TIR-based data fusion
- Captures thermal canopy stress signal – agricultural drought monitoring
- Fast response to variable soil evaporation rates (not captured by NDVI alone)