



### LDAS-Monde Sequential assimilation of satellite derived Vegetation and soil moisture products Applied to the Contiguous US

**Clement Albergel** <sup>1</sup>, Simon Munier <sup>1</sup>, Aymeric Bocher <sup>1</sup>, Bertrand Bonan <sup>1</sup>, Yongjun Zhang <sup>1</sup>, Clara Draper <sup>2</sup>, Delphine J. Leroux <sup>1</sup>, Jean-Christophe Calvet <sup>1</sup>

<sup>1</sup> CNRM UMR 3589, Météo-France/CNRS, Toulouse, France

<sup>2</sup> CIRES/NOAA Earth System Research Laboratory, Boulder, CO 80309, USA

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# + The summer 2018 heatwave impact on LSVs based on CGLS data

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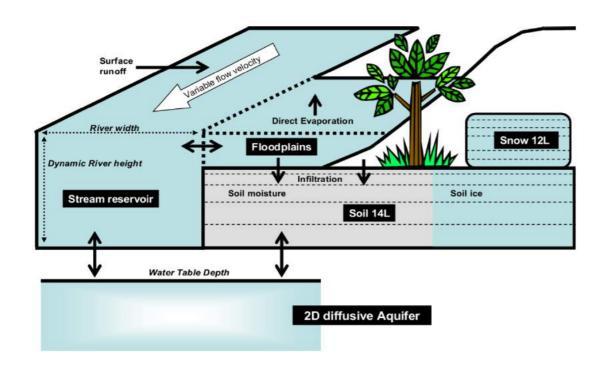
- Current fleet of Earth Satellite missions holds an unprecedent potential to quantify Land Surface Variables (LSVs) [Lettenmaier et al., 2015]
- Spatial and temporal gaps / Cannot observe all key LSVs
- Land Surface Models (LSMs) provide LSVs estimates at all time/location
- Through a weighted combination of both, LSVs can be better estimated than by either source of information alone [Reichle et al., 2007]
- Data assimilation

Spatially and temporally integrates the observed information into LSMs in a consistent way to unobserved locations, time steps and variables



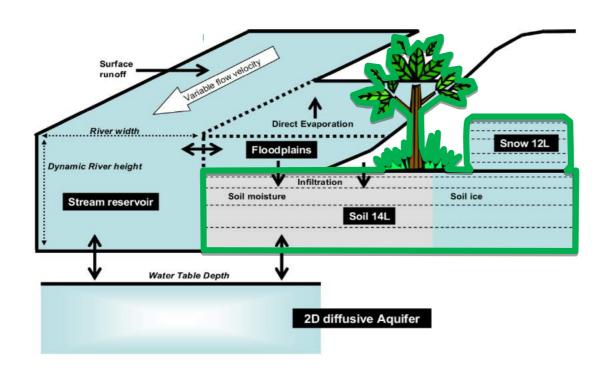


**LDAS-Monde**: Global capacity (sequential) integration of satellite derived observations into the SURFEX modelling platform



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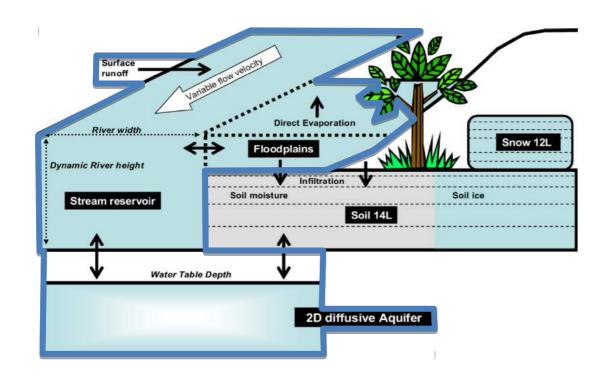
■ ISBA-A-gs: simulates the diurnal cycle of water and carbon fluxes, plant growth and key vegetation variables on a daily basis [Calvet et al., 1998, 2007, Gibelin et al., 2006]



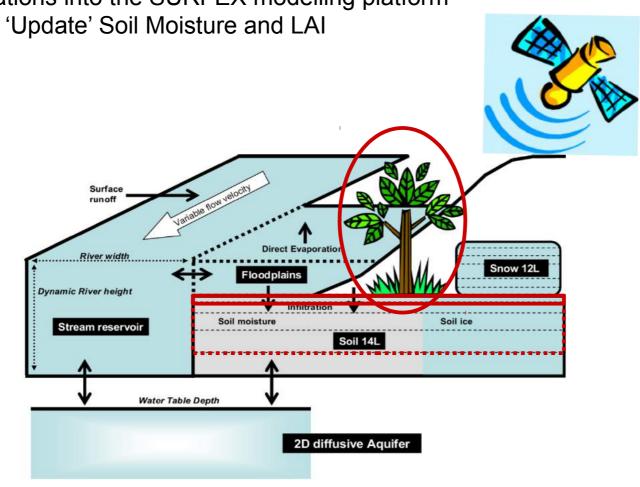
**LDAS-Monde**: Global capacity (sequential) integration of satellite derived observations into the SURFEX modelling platform

 CTRIP: TRIP based river routing system with CNRM developments for global hydrological applications

[Oki and Sud, 1998, Decharme et al., 2008, 2010]

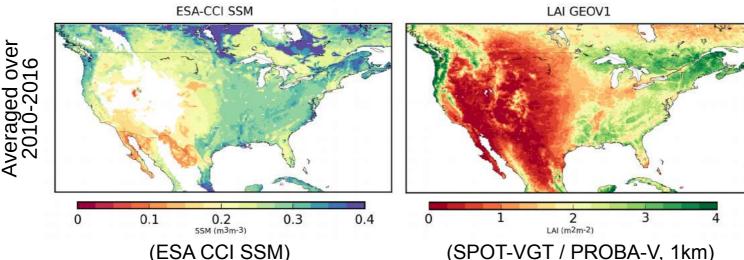


**LDAS-Monde**: Global capacity (sequential) integration of satellite derived observations into the SURFEX modelling platform



#### LDAS-Monde (Albergel et al., 2017, GMD; 2018 RS)

Model	Domaine	Atm. Forcing	DA Method	Assimilated Obs.	Observation Operator	Control Variables	Additional Option	
ISBA Multi-layer soil model CO <sub>2</sub> -responsive version (Interactive veg.)	Continental US (2010-2016, 0.25°x0.25°)	ERA-5 (HersBach, 2016)	SEKF	SSM (ESA CCI) LAI (GEOV1)	Second layer of soil (1-4cm) LAI	Layers of soil 2 to 8 (1-100cm) LAI	Coupling with CTRIP (0.5°)	



(ESA CCI SSM)
(seasonal bias correction)
http://www.esa-soilmoisture-cci.org

(SPOT-VGT / PROBA-V, 1km) http://land.copernicus.eu/global/





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- Direct impact of assimilating SSM on root-zone soil moisture as well as on LAI
- Direct impact of assimilating LAI on LAI and root-zone soil moisture
- Other variables benefit from the assimilation through biophysical processes and feedbacks in the model



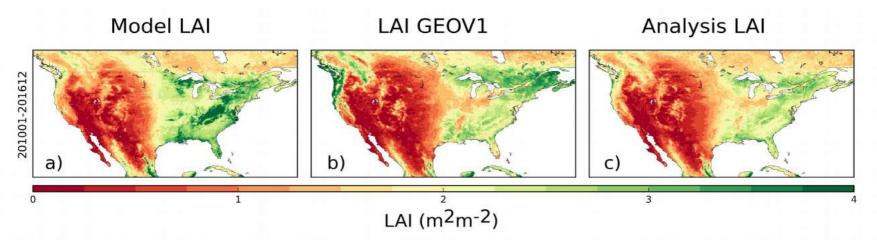


Analysis Impact is evaluated by comparing performance improvement relative to the open-loop

Assimilated SSM & LAI (analysis has to be closer to them than the open-loop!)	http://www.esa-soilmoisture-cci.org https://land.copernicus.eu/
In situ measurements of soil moisture from USCRN network	https://www.ncdc.noaa.gov/crn
River discharge from USGS	https://waterdata.usgs.gov/nwis
Evapotranspiration from the GLEAM project	http://www.gleam.eu
Gross Primary Production from the FLUXCOM project	http://www.fluxcom.org

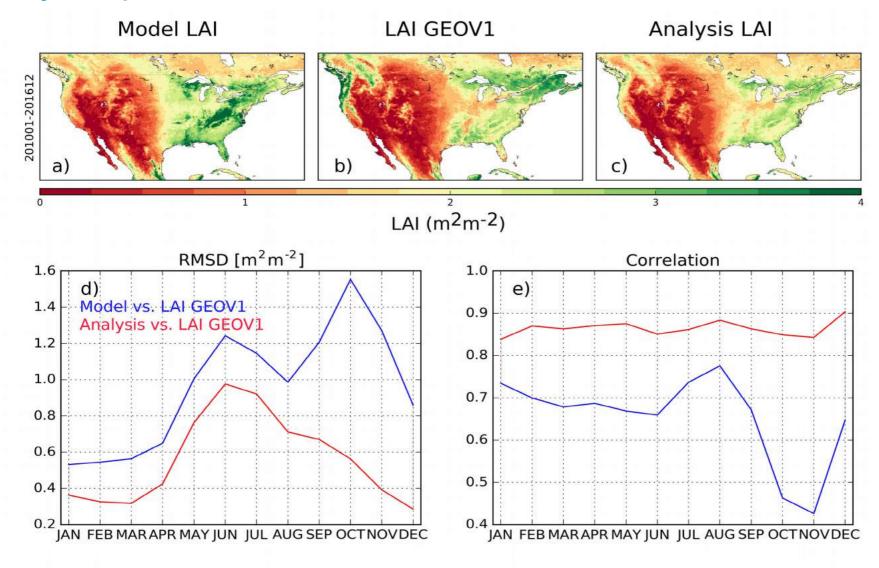


#### **Analysis impact: LAI**





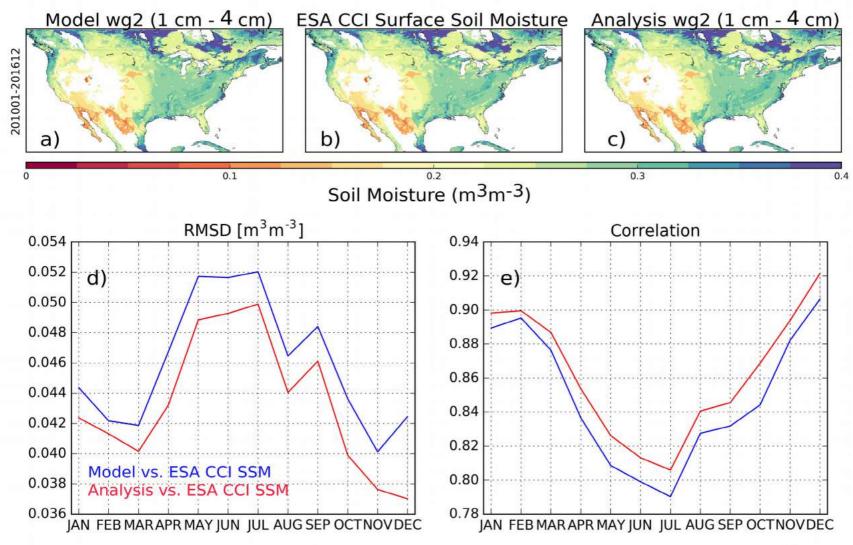
#### **Analysis impact: LAI**







#### **Analysis impact: SSM**

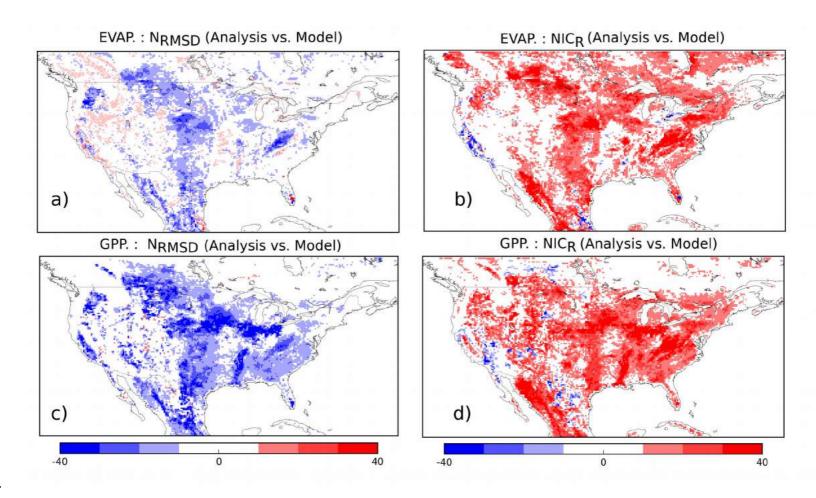






#### **Analysis impact: Evapotranspiration & Gross Primary Production**

- Normalised RMSDs and NIC (\*100) on R values to quantify improvement/degradation
- Blue (negative) colours for RMSD suggest that analysis is better (a & c panels)
- Red (positive) colours for R suggest analysis is better (b & d panels)



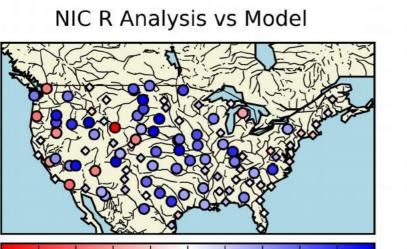


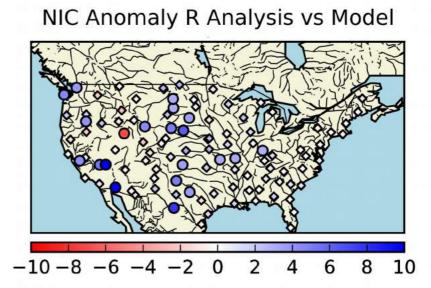
#### Vs. in situ soil moisture from USCRN network

(in situ 5cm vs ISBA 4-10cm, April-September 2010-2016, daily data)

8

10





110 (110) stations with significant R (Anomaly R)	Median R (Anomaly R)	Median ubRMSD				
Model	0.72 (0.60)	0.049				
Analysis	<b>0.74</b> (0.60)	0.048				

-10 - 8 - 6 - 4 - 2

NIC_R (NIC_ANO_R) >+3 % Blue circles
1
46 % (18 %)
Positive impact
<u>'</u>

NIC_R (NIC_ANO_R) < -3 % Red circles
/
8 % (1 %)
Negative impact

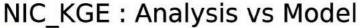
NIC\_NSE [-3,+3] Diamonds / 46 % (81 %) Neutral impact

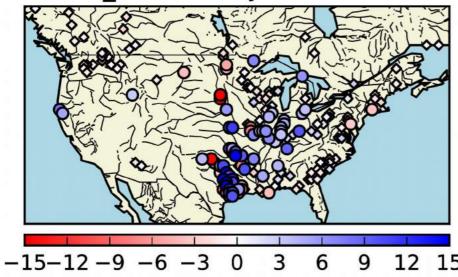




#### Vs. River discharge (USGS)

- Kling-Gupta efficiency (KGE) values are computed for each Exp. / stations (daily values scaled to the drainage area)
- NIC on KGE





Diamonds: neutral impact Red circles: negative impact Blue circle: positive impact

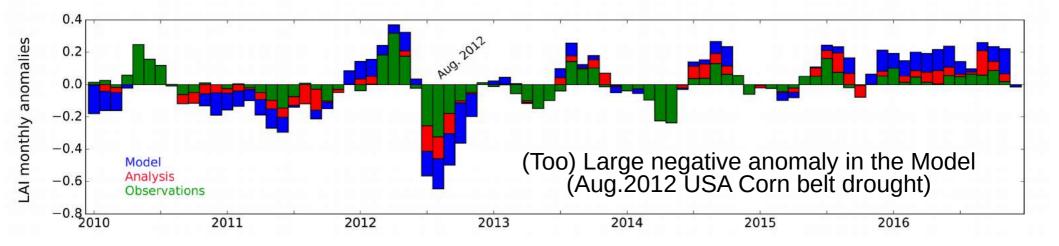
258 out of 531 Stations with KGE Greater than 0	Positive Impact: >+3	Negative Impact: <−3	Neutral Impact [-3; +3]			
NIC <sub>KGE</sub>	26%	12%	62%			
$N_{RE\sigma}$	22%	1%	77%			
$N_{RE}u$	34%	1%	65%			





### Monitoring agricultural drought

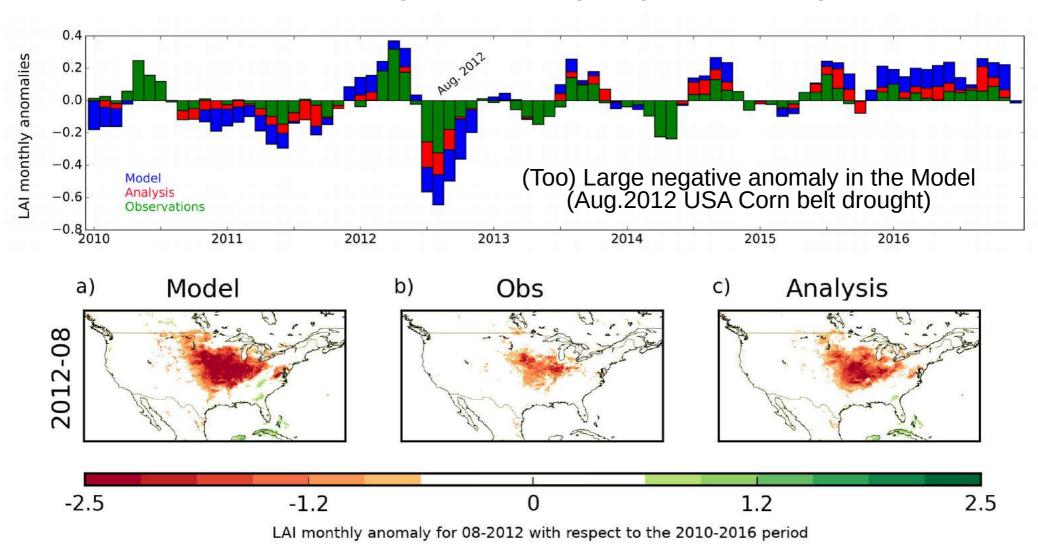
Can LDAS-Monde provides a good monitoring of agricultural drougth?





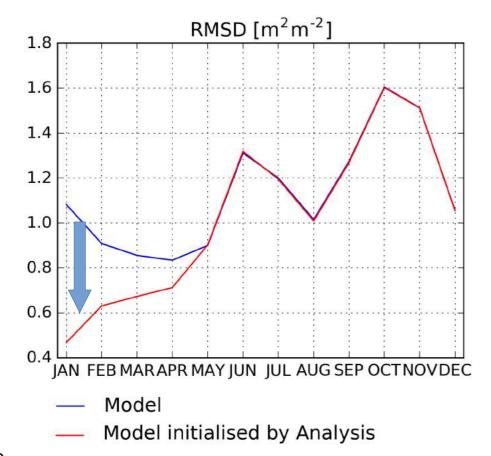
#### Monitoring agricultural drought

Can LDAS-Monde provides a good monitoring of agricultural drougth?



### From monitoring to forecasting

- Could analysis provide better initial conditions than model run? Does the impact last in time?
  - Use analysis initial conditions at 01/01/2016 to start a
     12-month Model run
  - Compare with a 'simple' model run
  - Evaluation against LAI observations over (2010-2016)
- > Persistence for several weeks / months on LAI



Model initialised with Analysis 2016-01 2016-02 2016-03 2016-04

0

-0.49

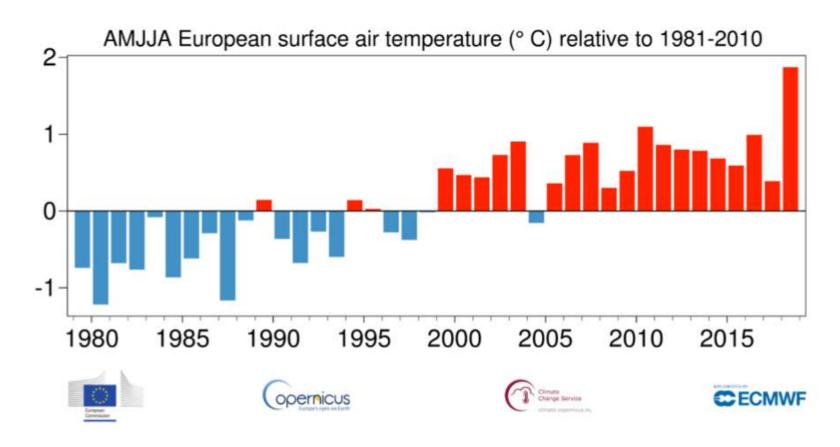
0.49

RMSD differences: Model -

## LDAS-Monde sequential assimilation of satellite derived vegetation and soil moisture products

- Direct impact on model control variables (Soil moisture and LAI)
- Other variables benefit from the assimilation through biophysical processes and feedbacks in the model (e.g., Evap., GPP, river discharge)
- → Powerful tool to monitor land surface variables (Extreme events like agricultural drought, also)
- → High potential of the analysis for initialising forecasts (Analysis provides better initial conditions than a model run)
- Towards higher spatial resolution LDAS-Monde using ECMWF IFS HRES (0.1°x0.1°)
- Combining ERA5 (0.25°x0.25°) and HRES (0.1°x0.1°) to force LDAS-Monde permits to have long Reanalysis with Real Time [and even forecast] capacities





**Evolution of near-surface air temperature anomalies.** This chart produced by C3S shows that the near-surface air temperature anomaly in Europe in the period of April to August (AMJJA), calculated relative to the 1981–2010 average for those months, was much larger in 2018 than in any previous year since 1979.

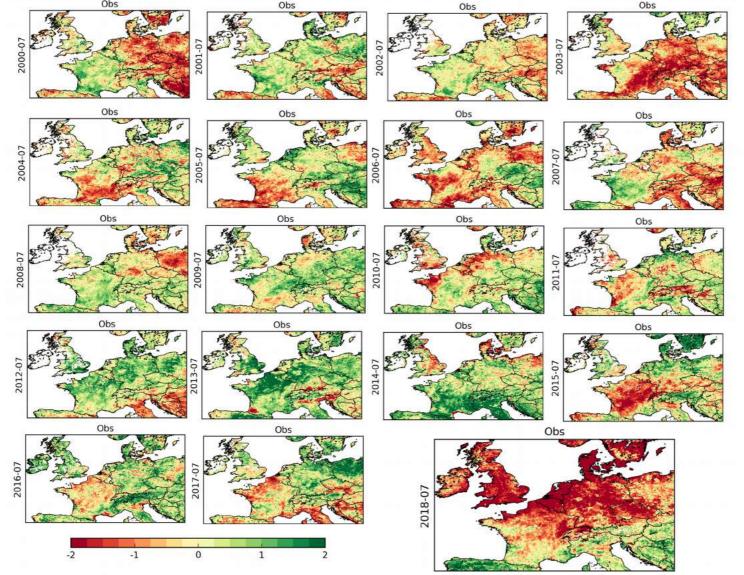
From Magnusson et al., 2018, ECMWF newsletter #154





#### The Earth Observations point of view: CGLS GEOV2

Leaf Area Index monthly anomaly (scaled by stdv) over 2000-09/2018

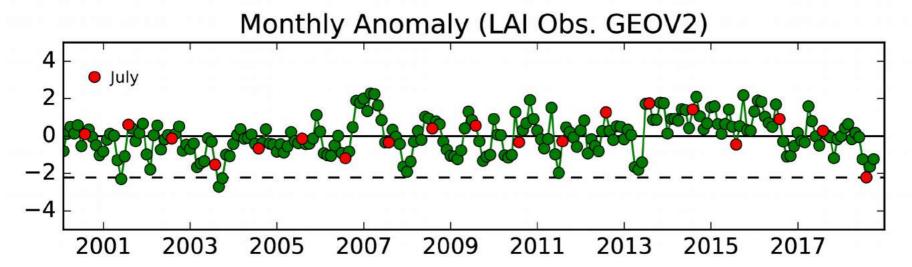






#### The Earth Observations point of view: CGLS GEOV2

Leaf Area Index monthly anomaly (scaled by stdv) over 2000-09/2018



- Dashed line illustrates july 2018 values
- Summer 2018 : 4 months in a row with strong negative anomaly!

#### % of the domain with monthly anomalies greater than -2 stdv (July)

07/2	07/2	07/2	07/2	07/2	07/2	07/2	07/2	07/2	07/2	07/2	07/2	07/2	07/2	07/2	07/2	07/2	07/2	07/2
000	001	002	003	004	005	006	007	008	009	010	011	012	013	014	015	016	017	018
5	0.4	0.25	5	0.6	0.8	1.84	1.14	0.22	0.03	0.67	0.70	0.28	0.7	0.25	2	0.10	0.6	18.8

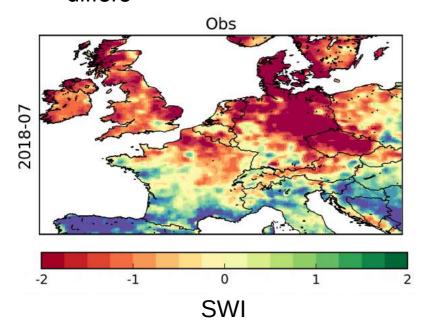
Larger area affected in Jul. 2018 than Jul. 2003 (in 2003 Aug./Sept. where most affected)

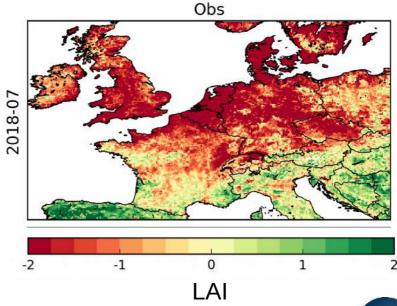




#### The Earth Observations point of view: CGLS ASCAT SWI [0-100]

- SWI monthly anomaly (scaled by stdv) over 2008-09/2018
- SWI patterns (top left) in agreement with LAI (bottom right) although time period differs

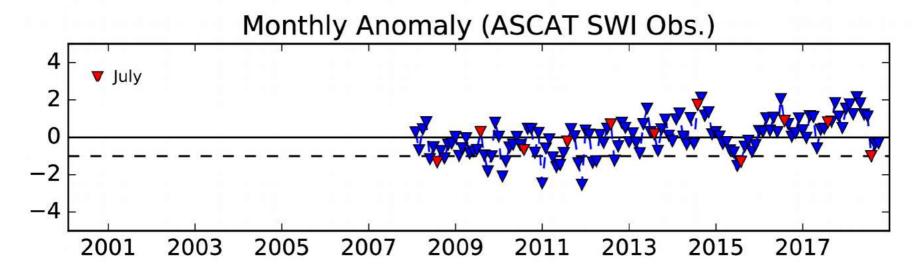






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Dashed line illustrates july 2018 values

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000	001	002	003	004	005	006	007	008	009	010	011	012	013	014	015	016	017	018
								2.2	0.04	1.75	0.17	1.5	0.5	0.06	3.02	0.01	0.01	10.0



