

# Applications of soil moisture measurements from SMOS within the Copernicus European and global Flood Awareness Systems (EFAS & GloFAS)

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# Outline

1. Introduction to Copernicus Emergency Management Service
2. Soil moisture experiments with SMOS data
  - GloFAS data assimilation
  - Estimation of irrigation
3. Conclusions

# CEMS service components



**Copernicus EMS On Demand Mapping** provides on-demand detailed information for selected emergency situations that arise from natural or man-made disasters anywhere in the world.

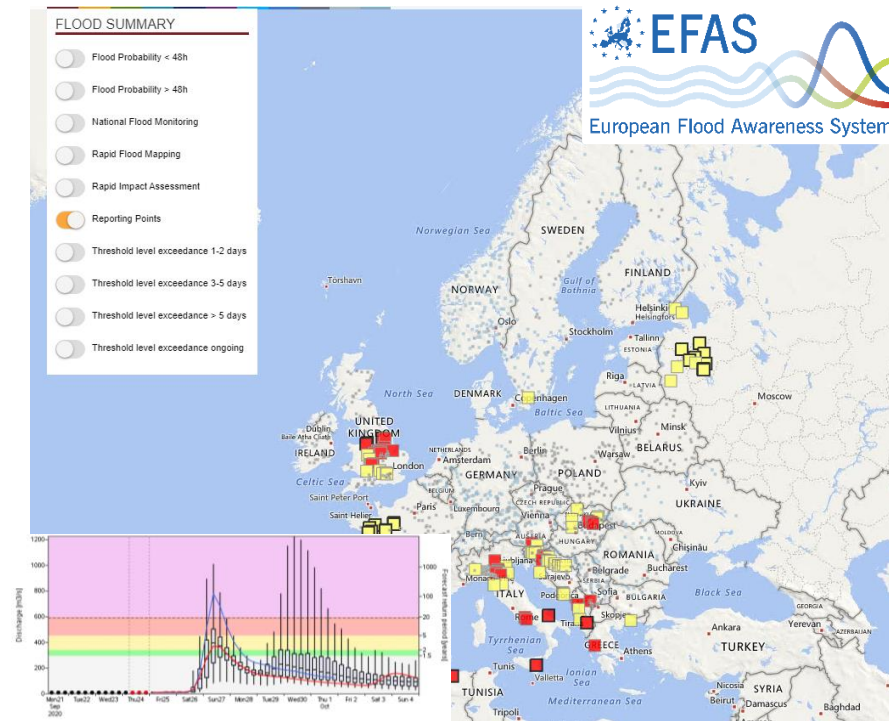
**Copernicus EMS Early Warning and Monitoring** offers critical geospatial information at European and global level through continuous observations and forecasts for floods, droughts and forest fires.

**Copernicus EMS Exposure Mapping** will offer global, harmonised and regularly updated information population, built-up areas and exposure.

# CEMS Flood Forecast Systems

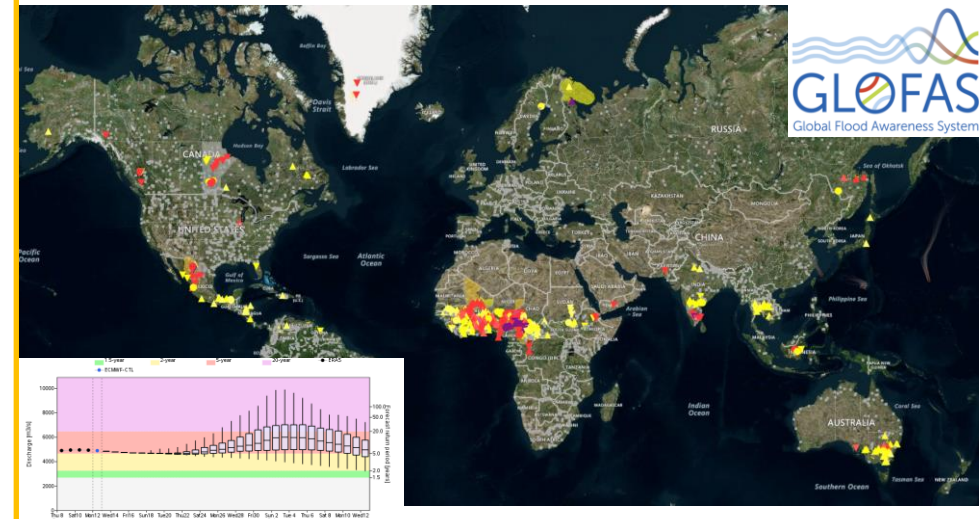
## European Flood Awareness System (EFAS)

- European forecasts
- 5 km resolution, 6 hourly time step
  - Planned upgrade to ~1.6 km
- Open access 30 days after forecast has been produced
  - Before this data are restricted to flood warning agencies
- [efas.eu](http://efas.eu)



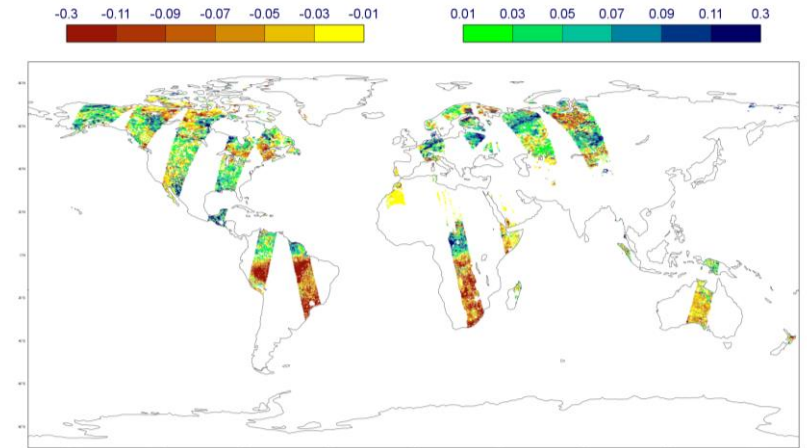
## Global Flood Awareness System (GloFAS)

- Global forecasts
- ~10 km resolution, daily time step
  - Planned upgrade to ~5 km
- Open access, >2000 registered users  
[globalfloods.eu](http://globalfloods.eu)



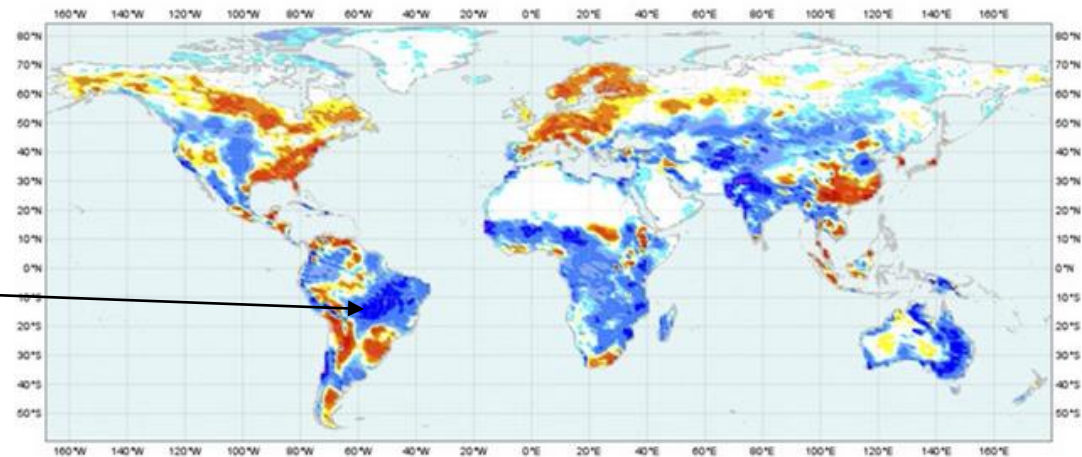
# Use of Satellite Data

1. Static input datasets
  - Topography – channel properties
  - Land cover – e.g. crop fractions
2. Data assimilation
  - ECMWF NWP forcings



SMOS soil moisture innovations [obs-model] *From Patricia de Rosnay*

LDAS adds large amounts of soil water e.g. Australia, southern Africa and Brazil

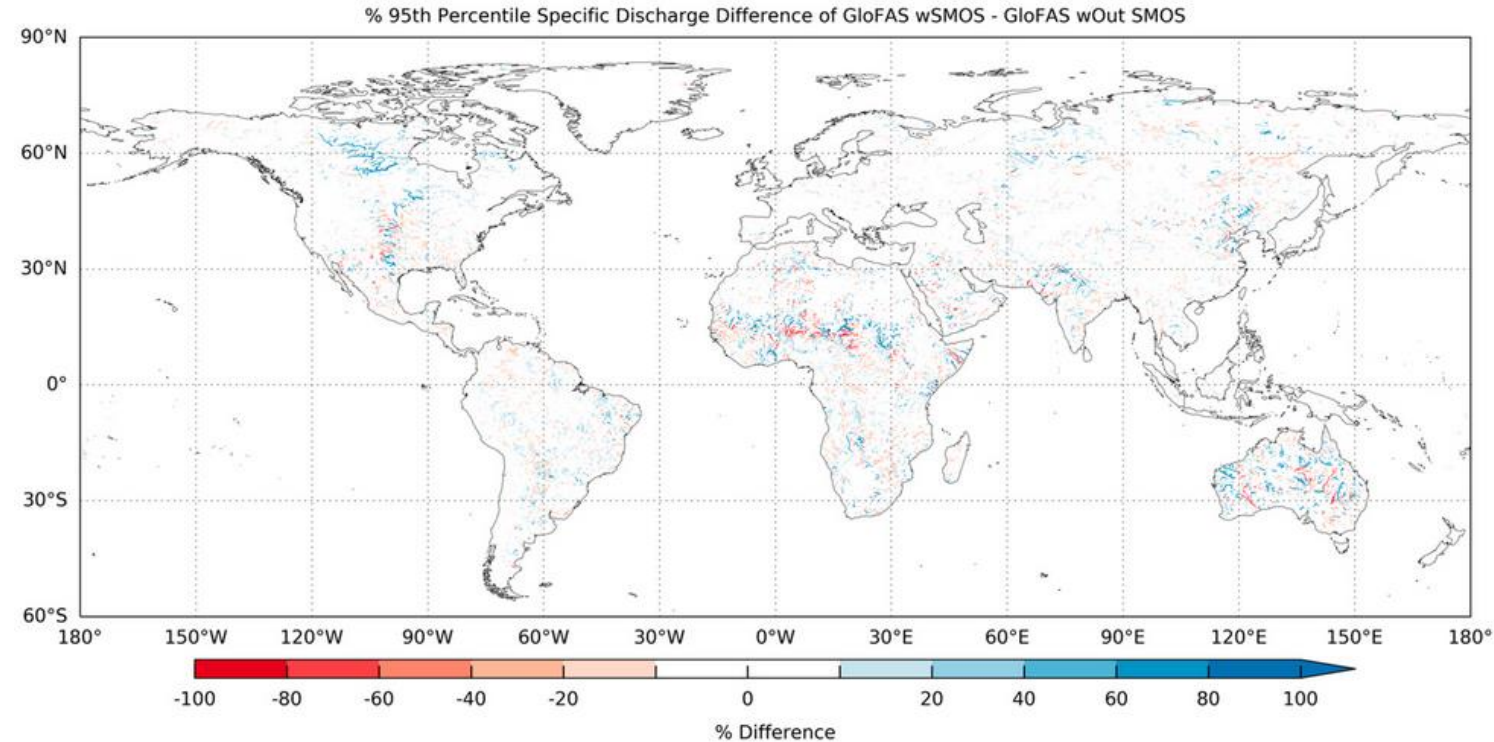
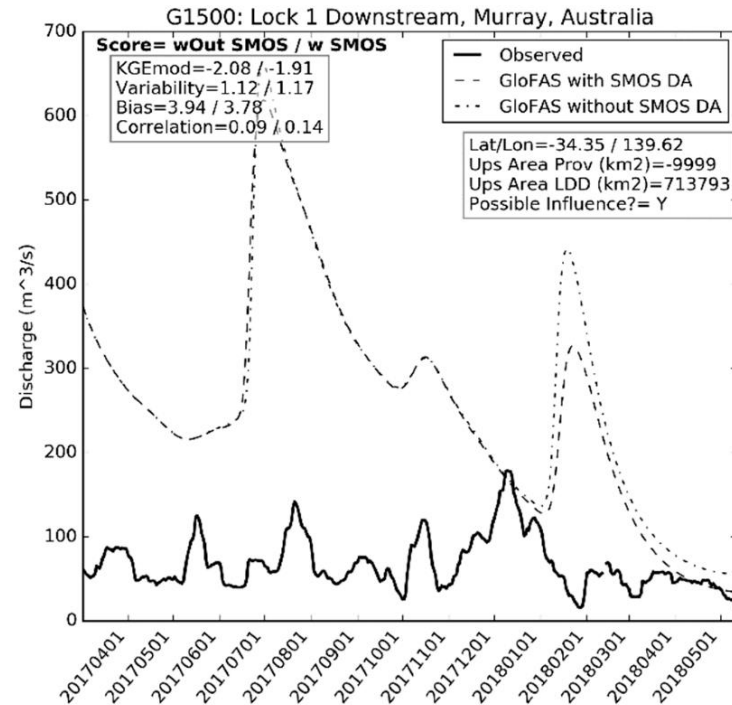


[Figure from Zsoter et al., 2019](#)

# SMOS Data Assimilation within GloFAS

Data denial experiments with SMOS show muted impact on streamflow

- Biases and poor correlation remain

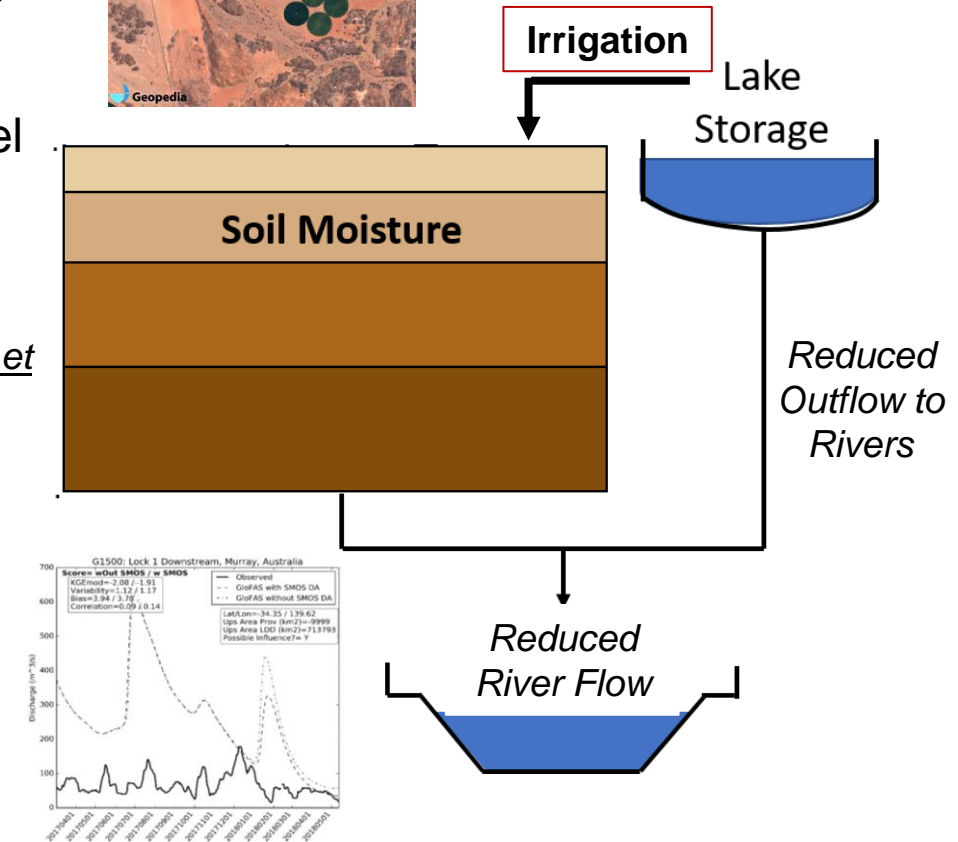
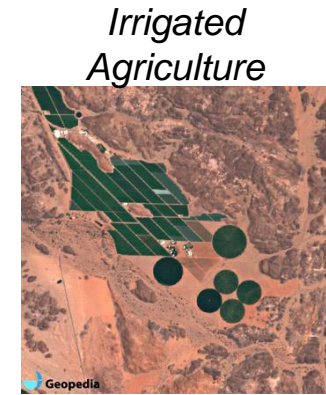


- ECMWF LDAS corrects for random errors, not systematic ones
- Process errors in Australia for example, maybe poor representation of processes such as irrigation and lake storage

# Estimating Irrigation from SMOS

- Hydrological performance in irrigated basins is often poor
- Can EO identify & quantify irrigation?; How do we incorporate this into our hydrological modelling?
- Irrigation computed as difference in soil moisture between a model which doesn't consider irrigation and satellite soil moisture which does
- Using SMOS (*sat*) and ERA5-Land (*mod*), irrigation (*I*) as: (Zaussinger et al., 2019)

$$I(t) = \frac{d\theta^{sat}}{dt} - \frac{d\theta^{mod}}{dt}$$



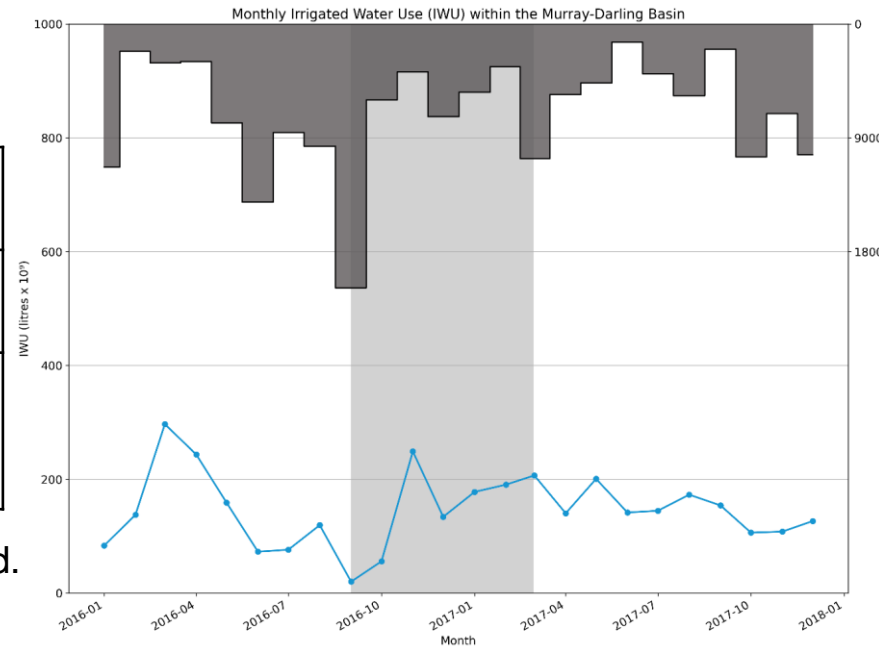
# Estimating Irrigation from SMOS

Results: Murray-Darling, Australia

- Irrigation estimated too broadly spatially
- Temporal trend does not correlate to the growing season
- Estimated IWU volume under-estimates

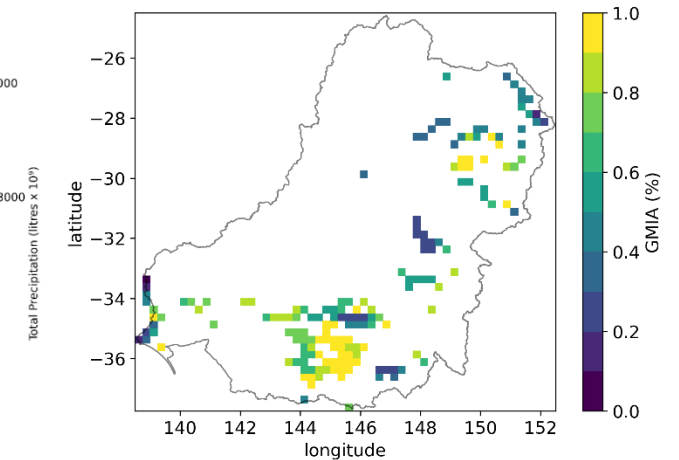
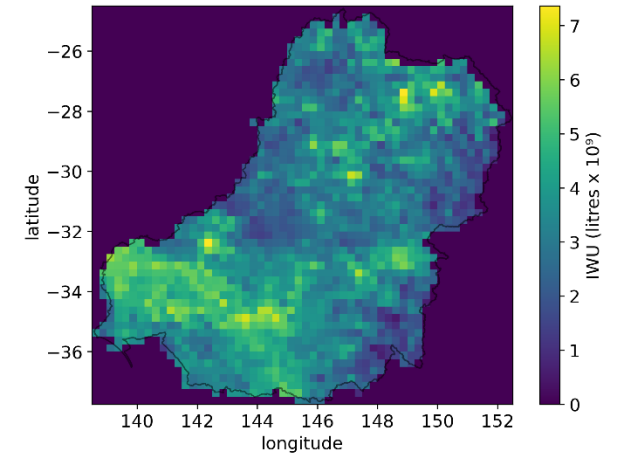
| Time Period  | Estimated IWU (x10 <sup>9</sup> l) | Error (x10 <sup>9</sup> l) |
|--|------------------------------------|----------------------------|
| 1 <sup>st</sup> July 2016 - 30 <sup>th</sup> June 2017                           | 1,712                              | -4665                      |
| 1 <sup>st</sup> September 2016 - 28 <sup>th</sup> February 2017 (growing season) | 827                                | -5550                      |

Estimated total IWU during specified time period. Error is versus reported irrigation totals from agricultural census



Estimated monthly basin total IWU (blue line). Light grey = growing season, dark grey = basin total precipitation

Annual average irrigated water use estimated from SMOS



Observed irrigated fraction (GMIA)



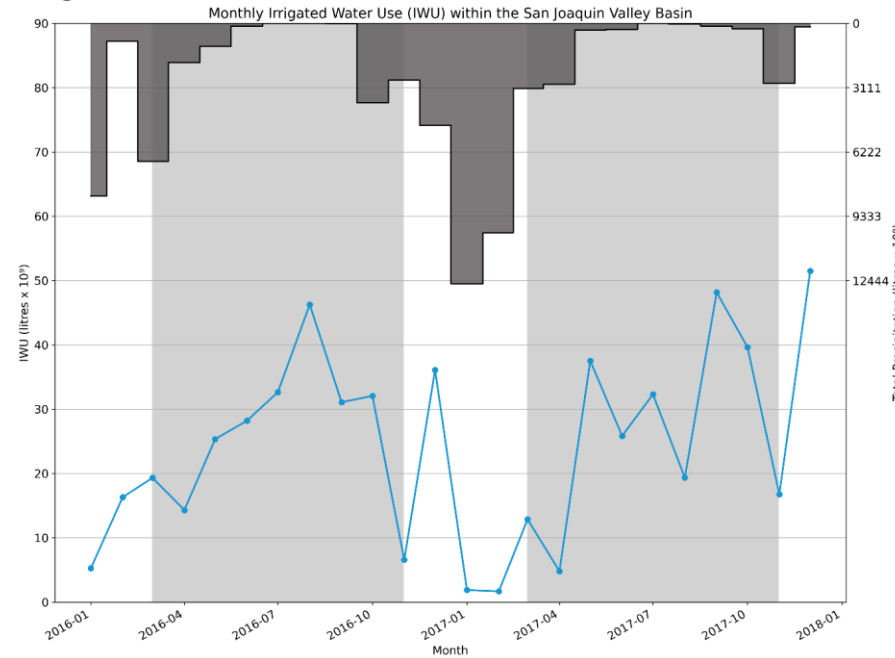
# Estimating Irrigation from SMOS

Results: San Joaquin, USA

- Spatial estimates of irrigation extent are closer (but easier to match)
- Slightly better monthly trend of increasing irrigation during growing season
- Estimated IWU volume much larger than reported data

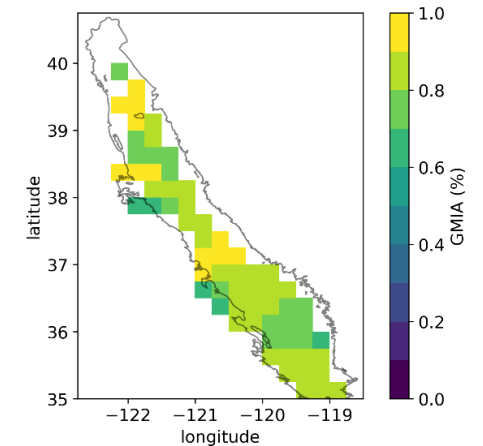
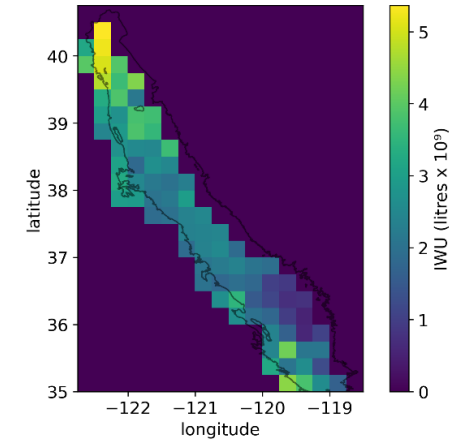
| Time Period   | Estimated IWU ( $\times 10^9$ l) | Error ( $\times 10^9$ l) |
|---|----------------------------------|--------------------------|
| 1 <sup>st</sup> Jan 2017 – 31 <sup>st</sup> Dec 2017                  | 218                              | 188                      |
| 1 <sup>st</sup> Mar 2017 – 30 <sup>th</sup> Oct 2017 (growing season) | 160                              | 130                      |

Estimated total IWU during specified time period.  
Error is versus reported irrigation totals from agricultural census



Estimated monthly basin total IWU (blue line). Light grey = growing season, dark grey = basin total precipitation

Annual average irrigated water use estimated from SMOS



Observed irrigated fraction (GMIA)

## Conclusions

- In our chosen study areas SMOS was not a reliable detector of irrigation
  - Irrigated and non-irrigated areas had similar signals
- Chosen areas could have efficient irrigation systems – require very high sensor accuracy to detect
  - Potentially more success in flood irrigated areas e.g. India, Spain
- Low spatial resolution – difficult to detect irrigation at field scales
- Previous studies with similar methods have had success
  - Perhaps an issue with our noise filtering

### **Future work:**

- DestinE – high resolution land surface modelling and accompanying DA
- Post-processing of river discharge forecasts with EO estimated discharge