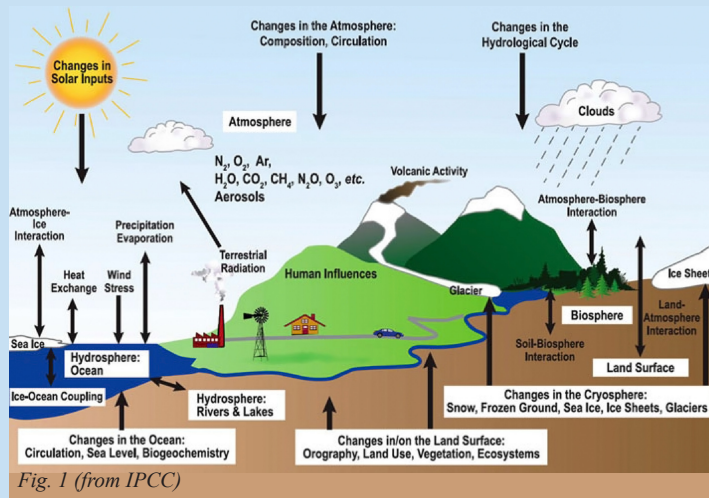
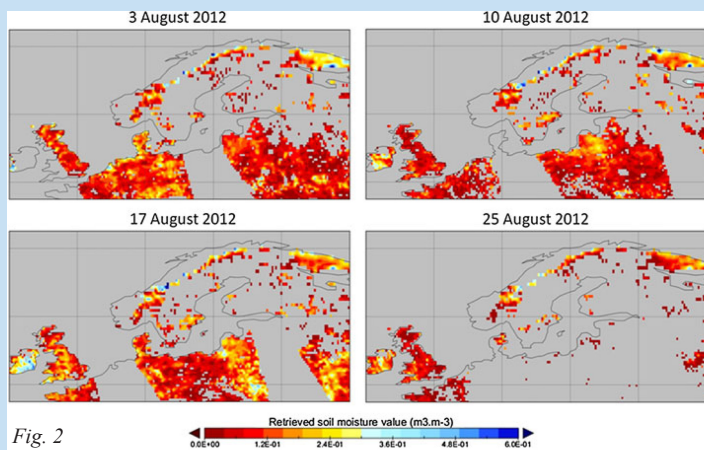


1. Introduction



Water stored on land is a key variable in the climate system (Seneviratne et al., 2010). (Fig. 1).



Hydrological observations have errors and are discrete in space and time so that their information has gaps (SMOS soil moisture data; Fig. 2). It is desirable to fill these gaps using additional information and computational techniques. Data assimilation (Kalnay, 2003; Lahoz et al., 2010a) is an objective way of doing this

The main challenges for observations/models of the land surface hydrological cycle are given in **panel 2**; the challenges for land data assimilation (DA) are given in **panel 3**. For further details see Lahoz and De Lannoy (2013).

3. Challenges for land data assimilation:

Assimilated satellite observations include retrievals of land surface temperature, soil moisture, snow water equivalent and snow cover area (Lahoz et al., 2010). Challenges include:

- Disaggregate data from large scales to small scales;
- Radiance DA to avoid inconsistencies between a priori information and land surface models;
- Multiple sensors and new sensors;
- Combination of state/forcing/parameter information;
- Representation of observation/forecast errors, and specification of biases in observational/model information;
- Advanced DA techniques (e.g., NILU Ensemble Kalman Filter; Fig. 4);
- Preserving water balance;

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- ESA CCI Soil moisture: <http://www.esa-soilmoisture-cci.org>
- NFR Project 202315/V3
- Thanks Finn Bjørklid (NILU)

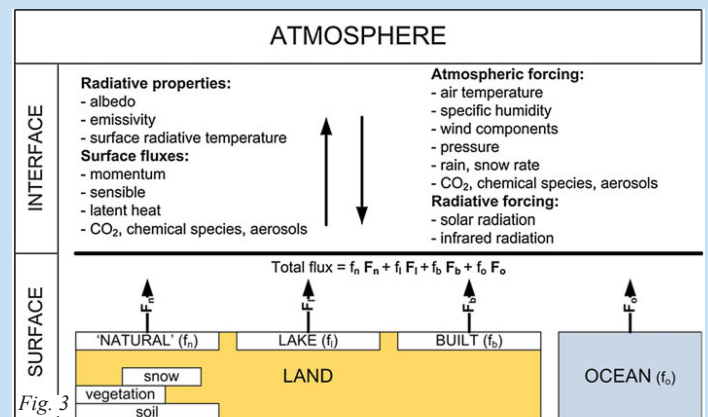
2. Challenges, observations/models of the land surface hydrological cycle:

Satellites observations:

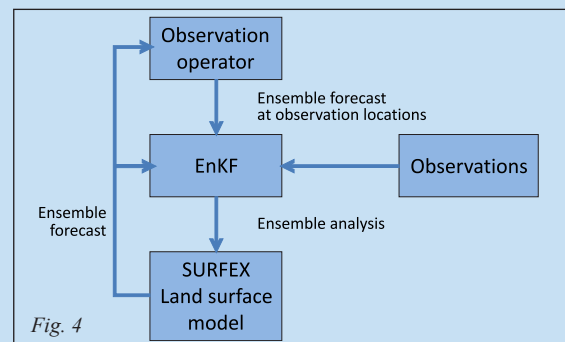
- Capture land system spatio-temporal variability;
- Length of time periods should allow trend identification and bias mitigation;
- Provide uncertainty information;
- Relate observations to key land system variables.

Models:

- Models (SURFEX model; Fig. 3) to describe physical processes, including coupling land surface with more specialized modules (e.g. dynamic vegetation, snow);
- Consistent parameter datasets to limit predictive uncertainty;
- High quality forcing data, e.g., precipitation.



Improvements in observational and model information from addressing these challenges, will improve application of DA ideas to understand the land hydrological cycle.



4. Conclusions:

To understand the land hydrological cycle, we need *observations* and *models*. *Data assimilation* combines observational and model information. Data assimilation *adds value to observations* by filling the gaps between them and *adds value to models* by constraining them with observations.

Collectively, new satellite missions, increasing attention to forecast uncertainty due to errors in land surface models, parameters and inputs, and development of advanced assimilation techniques will *close the largest gaps in our understanding of the hydrological cycle over land*.

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