# Session 6:

Remote sensing and data assimilation of hydrological components in the coming era of Earth Observation

# Introductory Remarks

*Eric F. Wood, Chair Princeton University* 

Hydrology delivering Earth System Science to Society 28th February, 2007 ~ 2nd March, 2007

Tsukuba, Japan

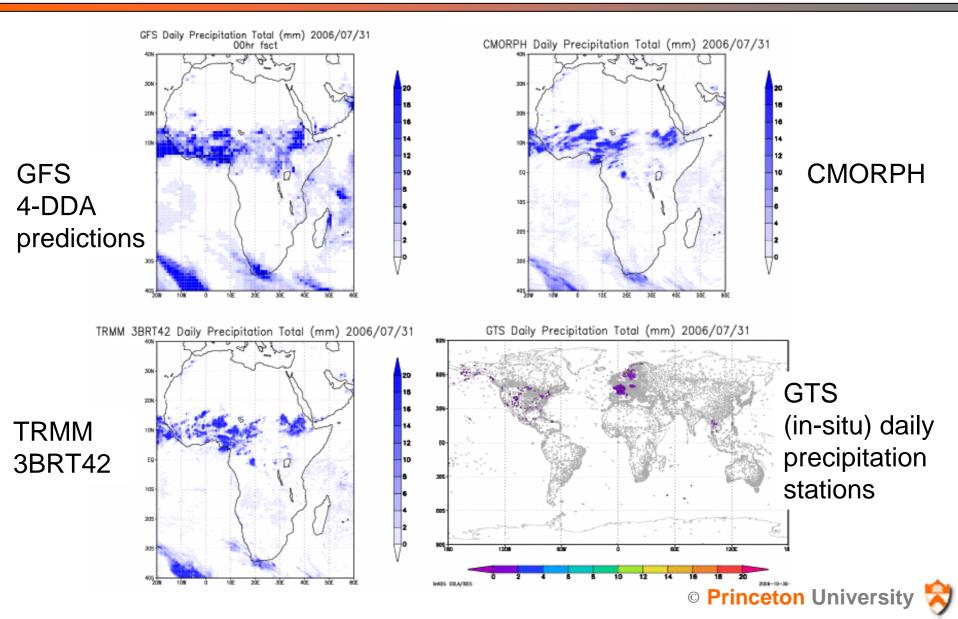


### The need for remote sensing Earth Observations

 We can't measure everywhere using in-situ measurements, so space observations offer consistent, global observations (e.g. precipitation)

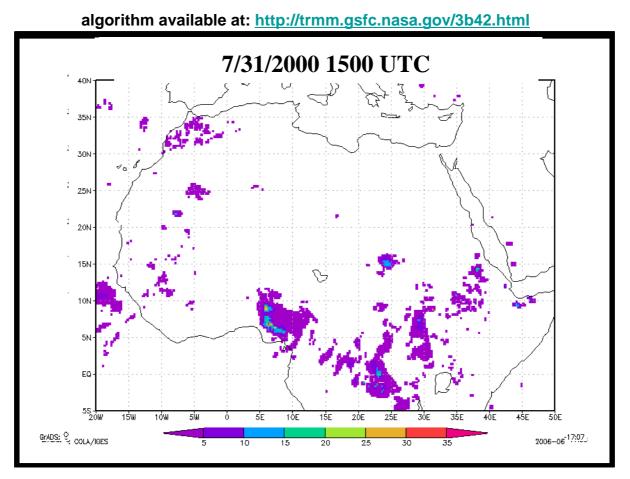


## Real-time data availability for hydrology – the options



# Data availability for hydrology – satellite precipitation

TRMM 3B42 merged high quality infrared precipitation product- 3hrly 0.25 x 0.25 degree gridded estimates of global precipitation [mm/hr] (instantaneous precipitation rate at the nominal observation time)



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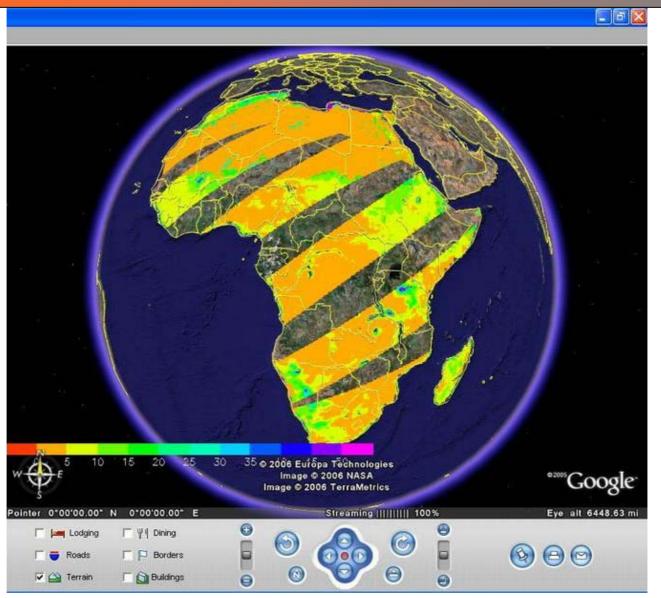
## The need for remote sensing Earth Observations

 We can't measure everywhere using in-situ measurements, so space observations offer consistent, global observations (e.g. precipitation)

 Some variables rarely or sparsely measured in-situ – e.g. surface soil moisture and evapotranspiration, so satellite observations offers the only feasible measurement system.



### Soil Moisture from Space



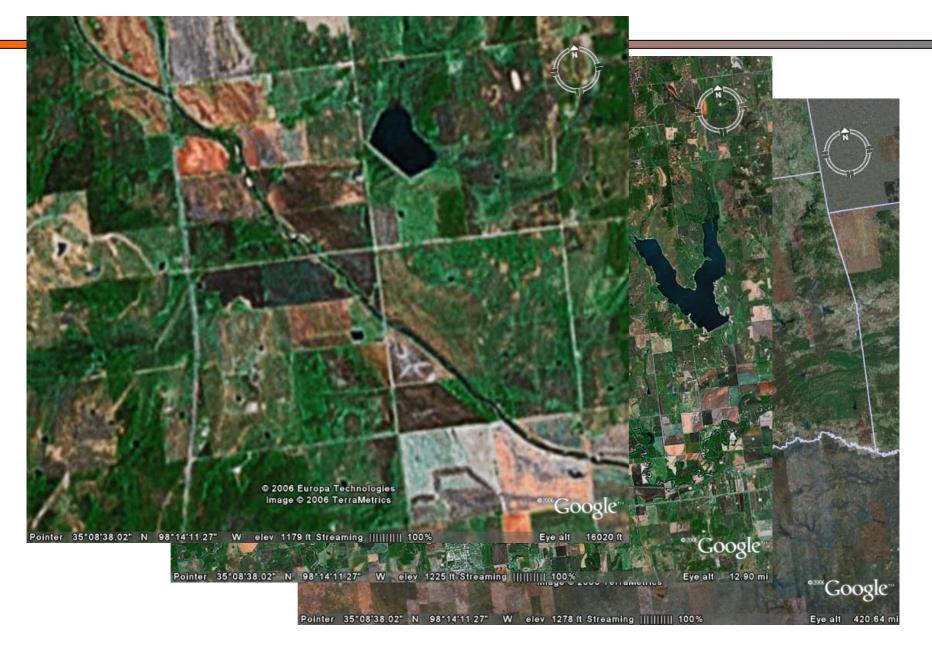
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### Challenges facing remote sensing

1. Issues of scale, and sub-pixel "contamination"

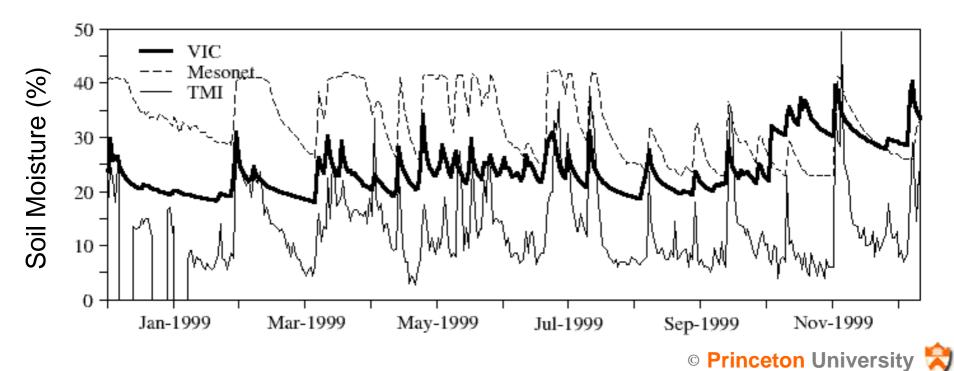


### Spatial Variability of Land Surface

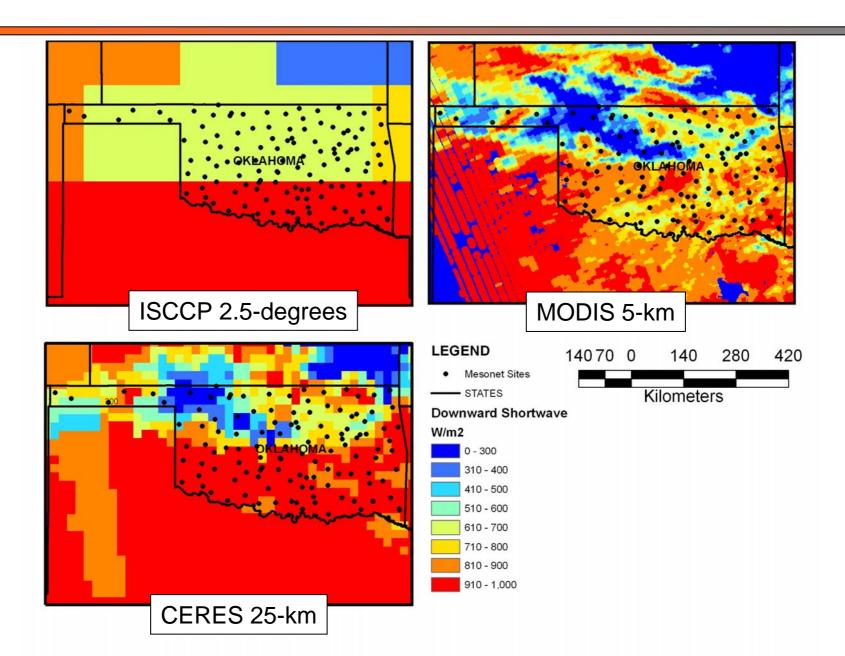


Time series from different sources, measured at different scales behave differently; yet they are correlated and show skill in data assimilation – how to evaluate them?

In-situ (points over the region); VIC (10 km); TMI (~35 km)



#### Surface insolation products for estimating ET (land heat fluxes)



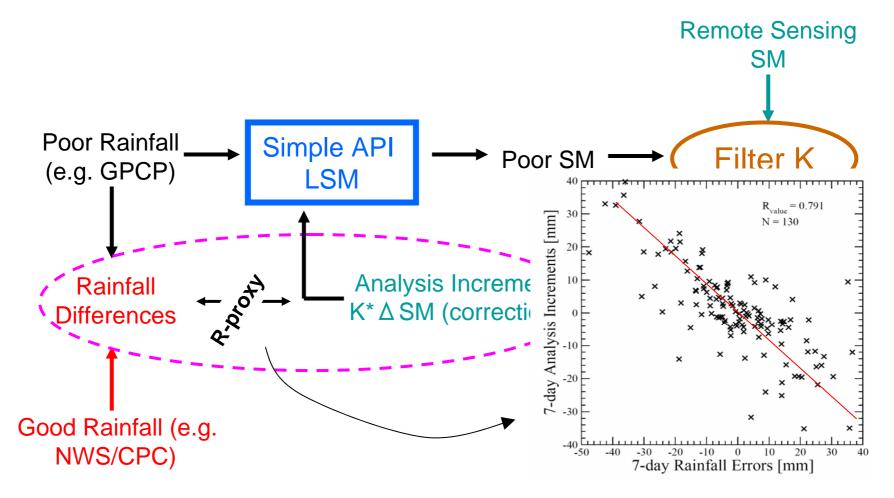
## Challenges facing remote sensing

- 1. Issues of scale, and sub-pixel "contamination"
- 2. Remote sensing validation (and calibration) at largescales: a new paradigm is needed.

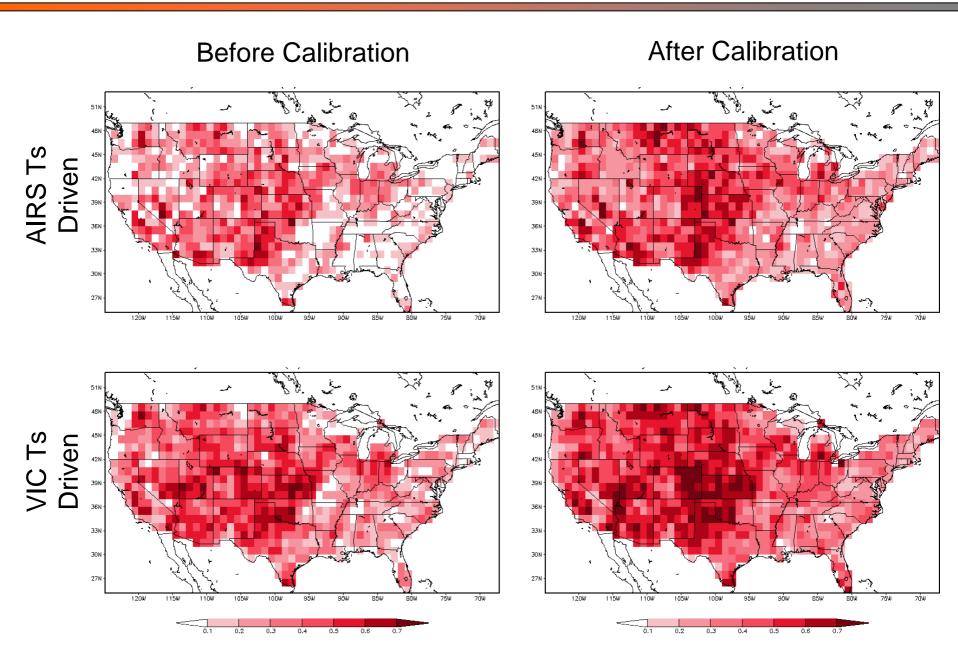


# Soil Moisture "Value" in Data Assimilation

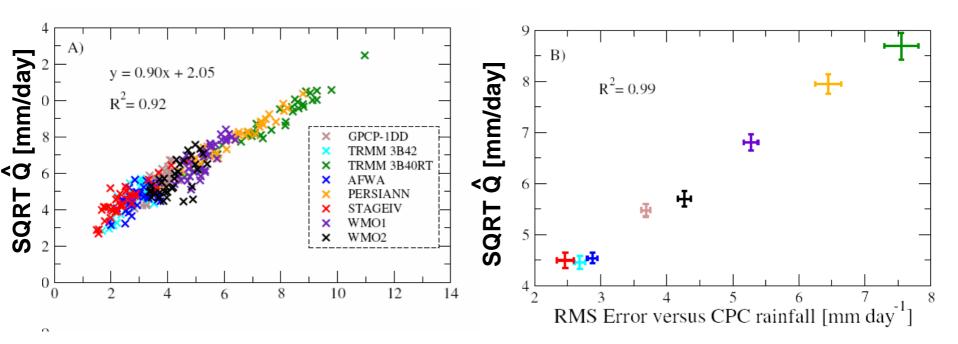
Instead of comparing remotely-sensed soil moisture to ground measurements, look for how much the soil moisture product can contribute when it is assimilated into a (simple) land surface model (LSM) driven by poor rainfall forcings (after Crow, 2007).



### "R Proxies" using VIC and AIRS Ts for soil moisture retrievals

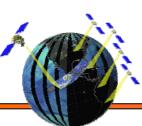


#### Using adaptive filtering to estimate errors from <u>satellite-</u> retrieved precipitation (from Crow and Bolton, 2007)

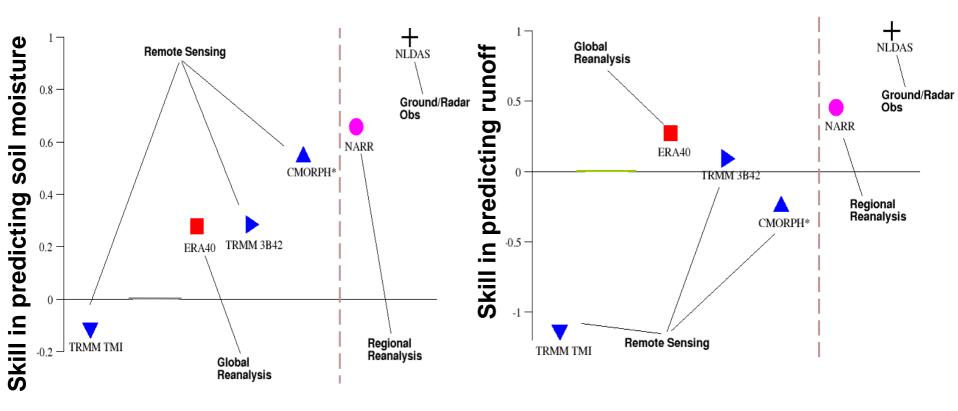


Estimated forecast error versus CPC for all 1-degree SGP boxes (July 1, 2002 to Dec 31, 2005)

Estimated forecast error versus CPC spatially averaged for all boxes and retrievals.



# Estimated skill scores in using <u>satellite-retrieved</u> precipitation in hydrologic models



Skill depends on the land surface model, and the scales (spatial and temporal) used in the analysis.

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## Challenges facing remote sensing

- 1. Issues of scale, and sub-pixel "contamination"
- 2. Remote sensing validation (and calibration) at largescales: a new paradigm is needed.
- 3. Efficient algorithms for data assimilation that considers scale and dynamics: multi-scale data assimilation



# The potential for a space-borne Global Water Cycle observation system

