

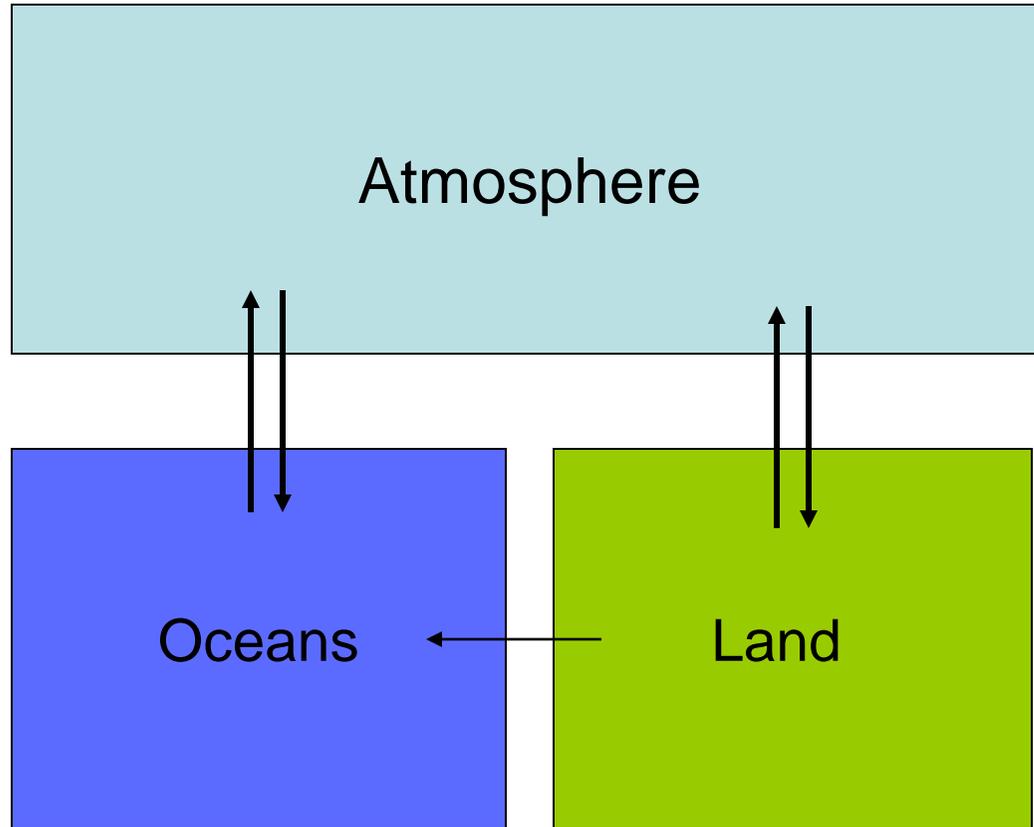
# Role of land-surface processes for seasonal prediction

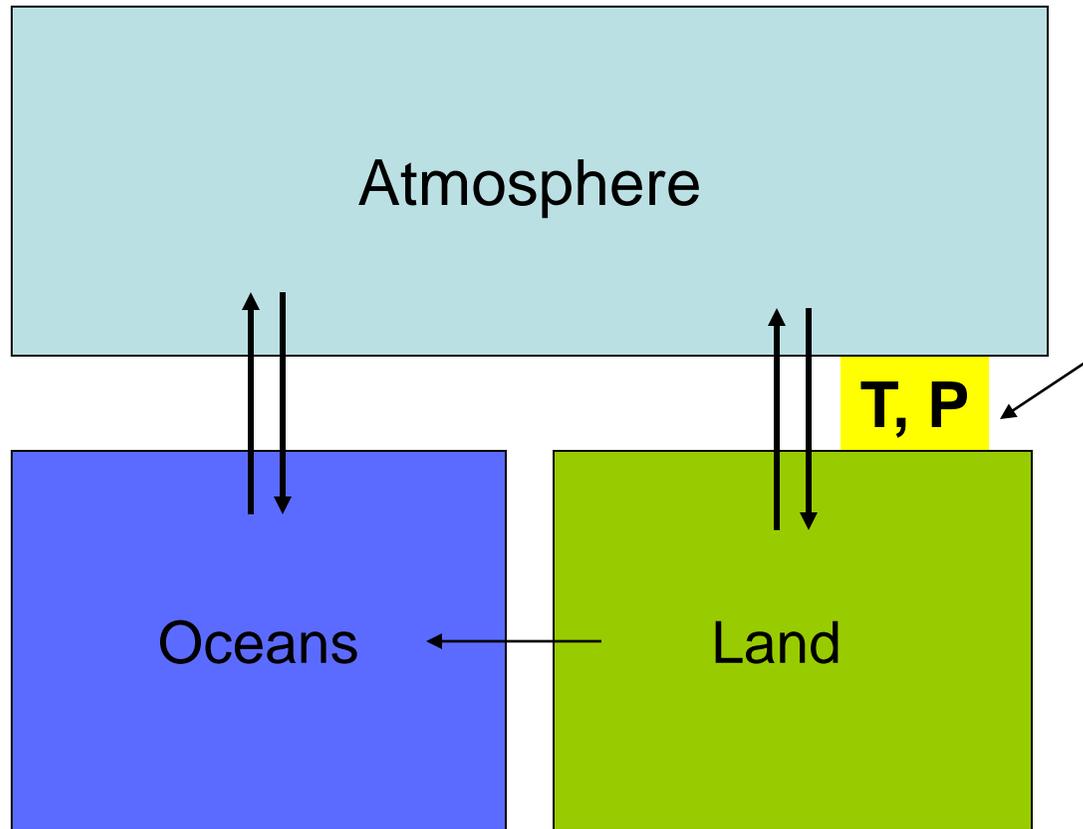
**Sonia I. Seneviratne**

Institute for Atmospheric and Climate Science, ETH Zurich, Switzerland  
sonia.seneviratne@env.ethz.ch

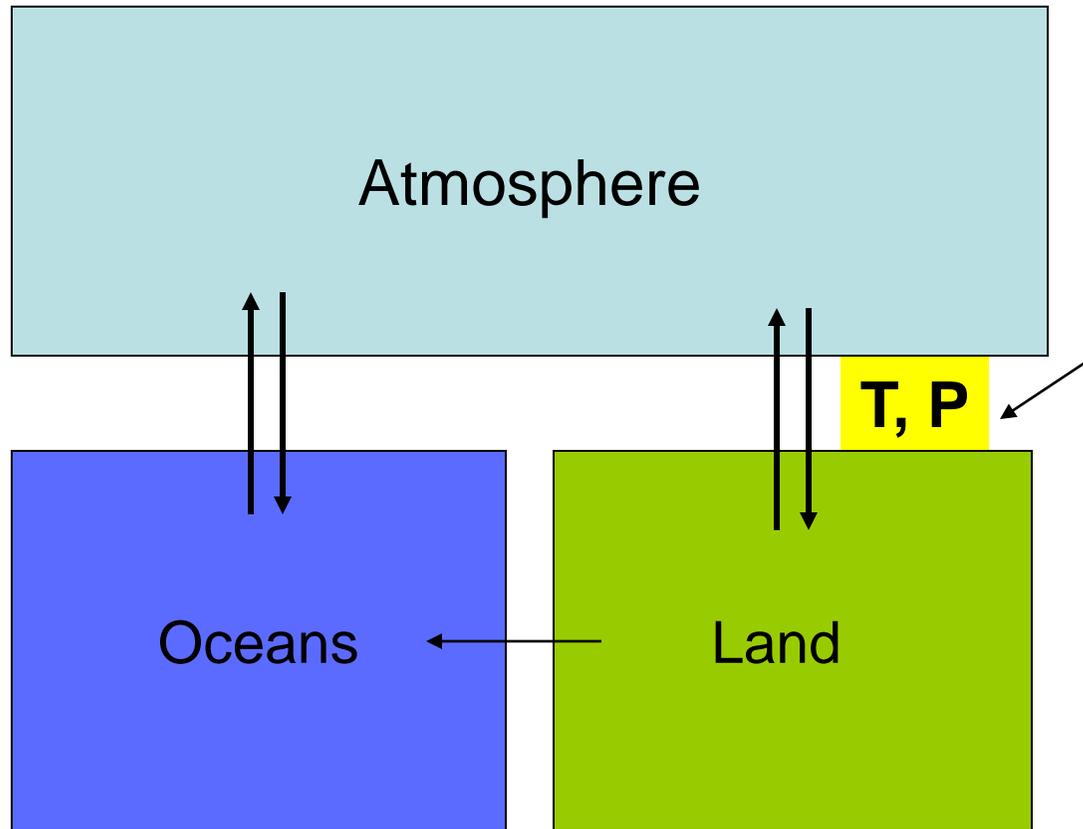
*Thanks to: R. Koster, B. Mueller, R. Orth*

- Basic relevant land-climate feedbacks and observational evidence
- Diagnosing prediction potential from land surface initialization
  - GLACE-2: Forecasting of atmospheric variables
  - Drought forecasting
- Discussion
- Conclusions





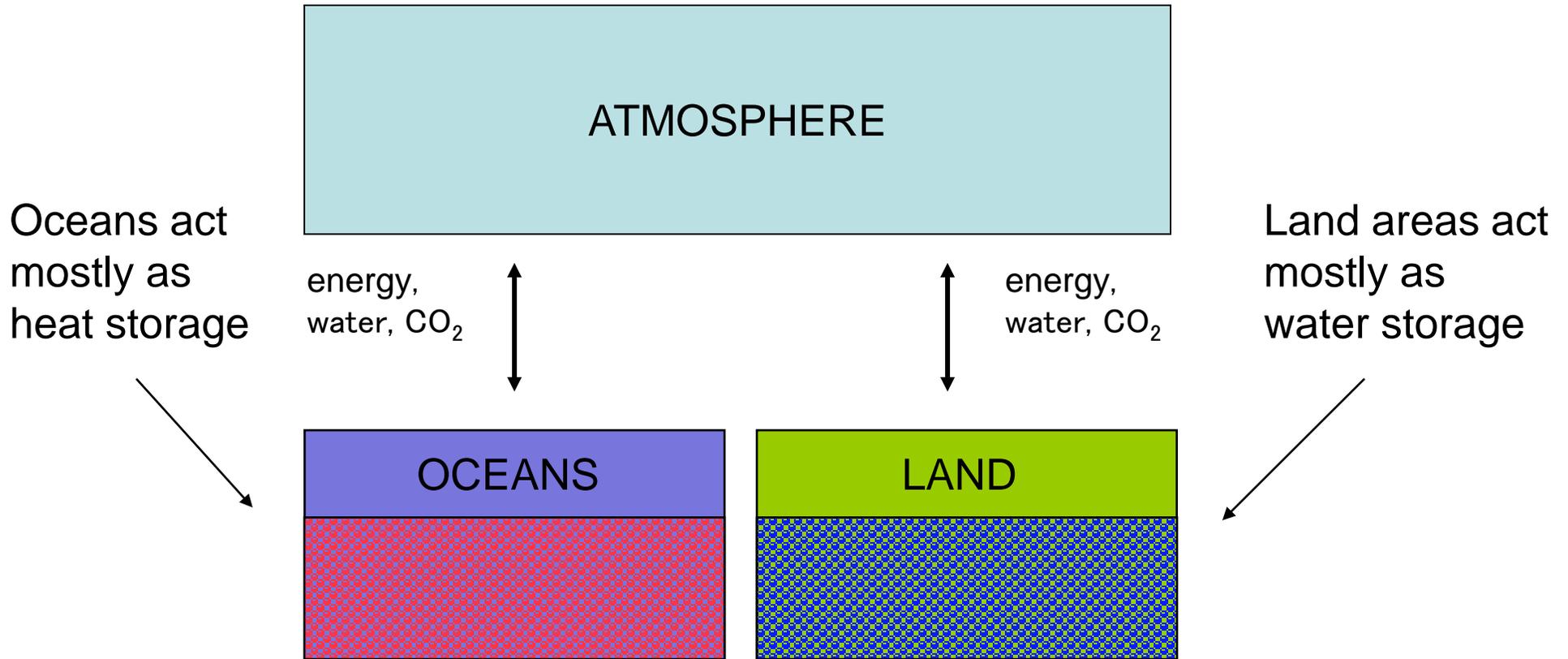
We are generally interested in climate over land!

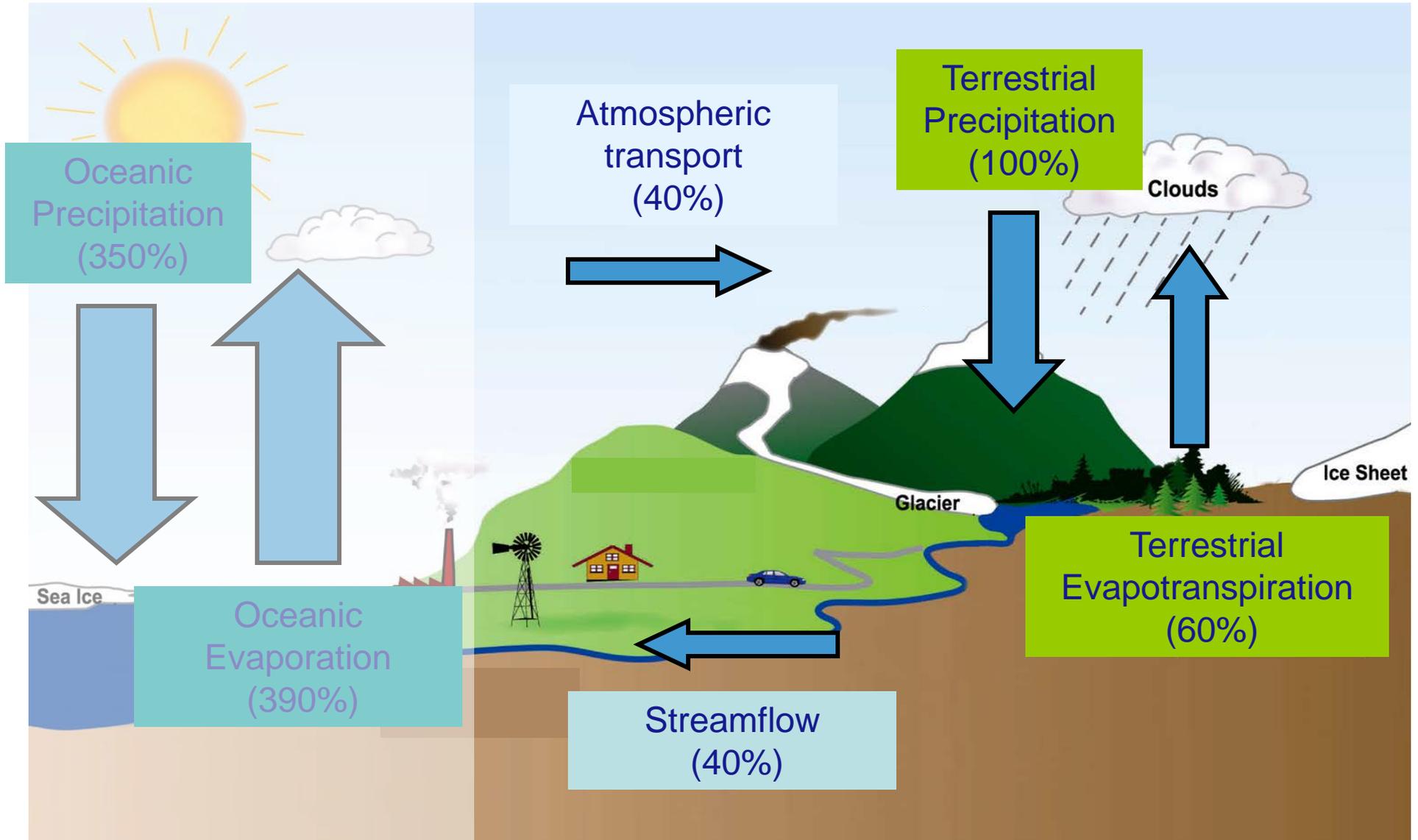


We are generally interested in climate over land!

Land climate is strongly affected by land processes in several regions

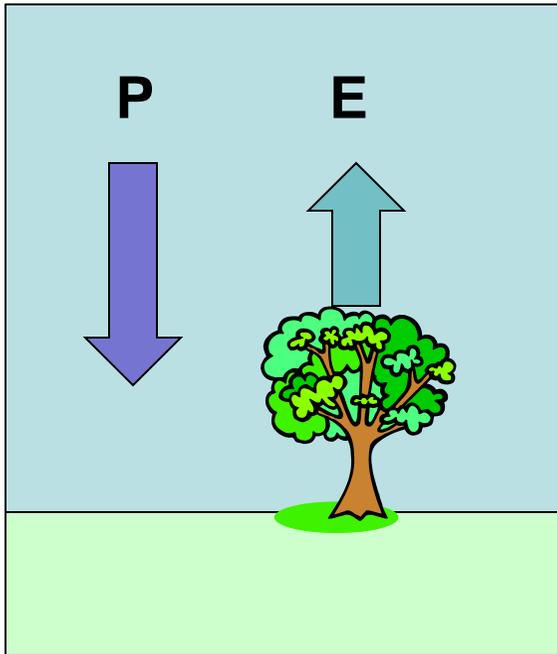
## Land water storage vs oceans' heat storage:





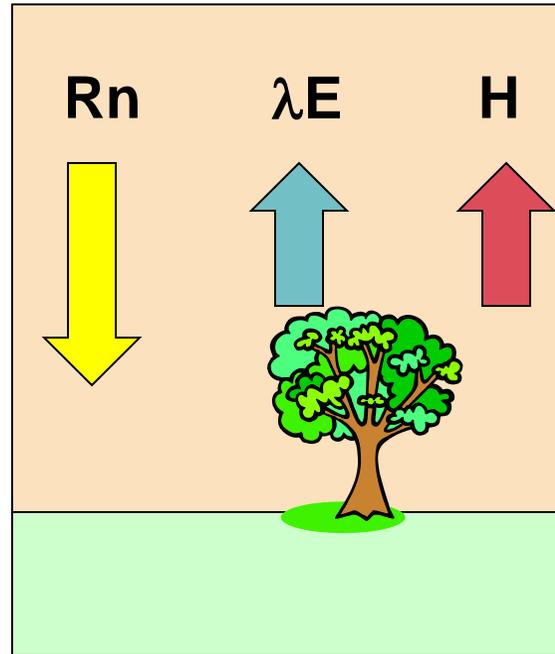
(Flux estimates: Oki and Kanae, Science 2006)

## Water



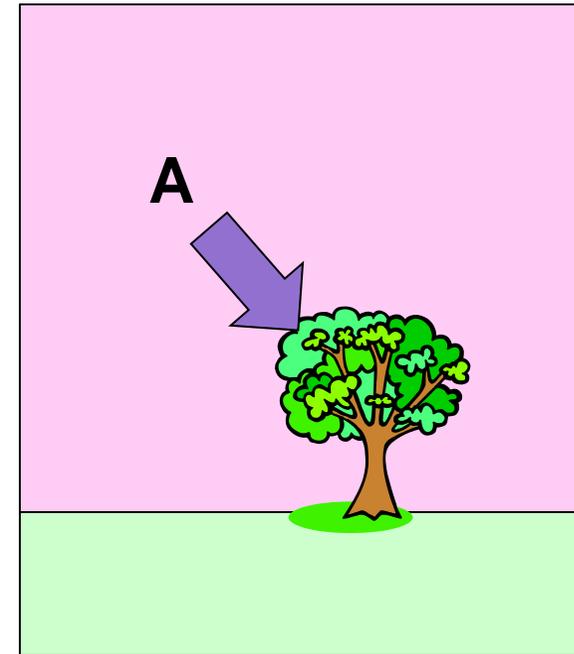
$$E = 60\%P$$

## Energy



$$\lambda E = 50-60\%Rn$$

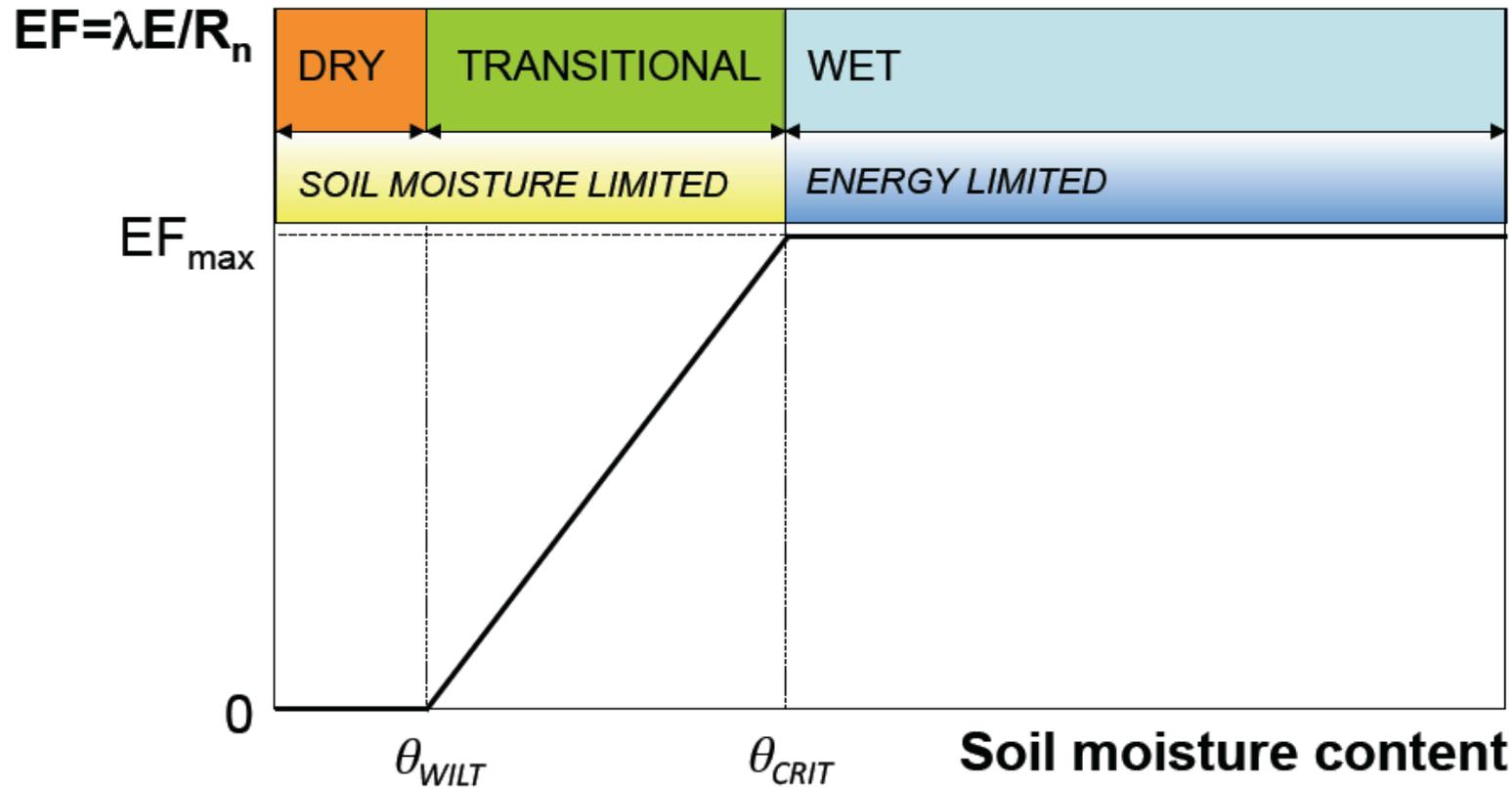
## Carbon



**A is a sink for 30%  
of C emissions**

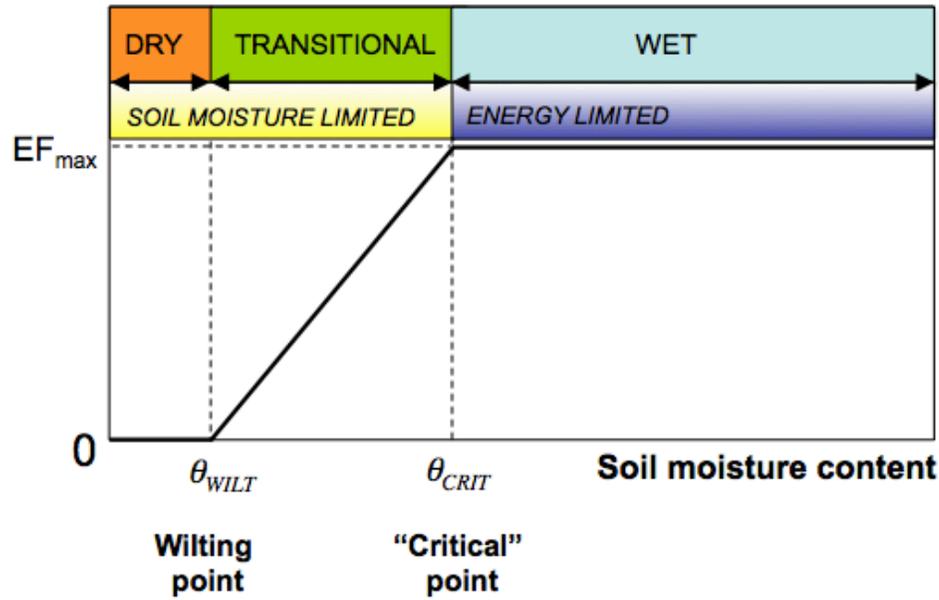
(see also Budyko 1956)

## Evaporative fraction

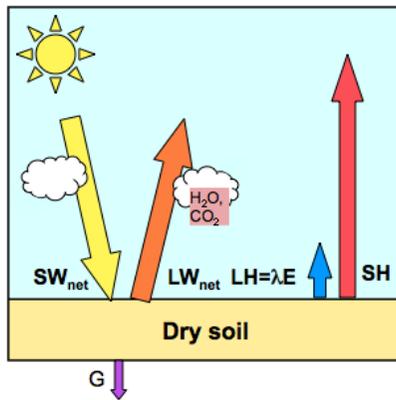


(Seneviratne et al. 2010, Earth-Science Reviews)

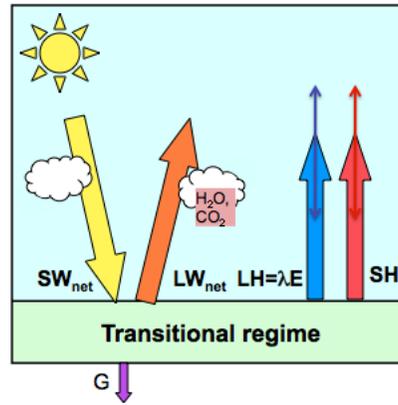
Evaporative fraction  $EF = \lambda E / R_n$



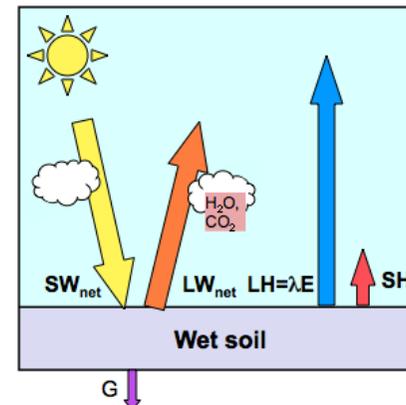
Dry climate regime



Transitional climate regime

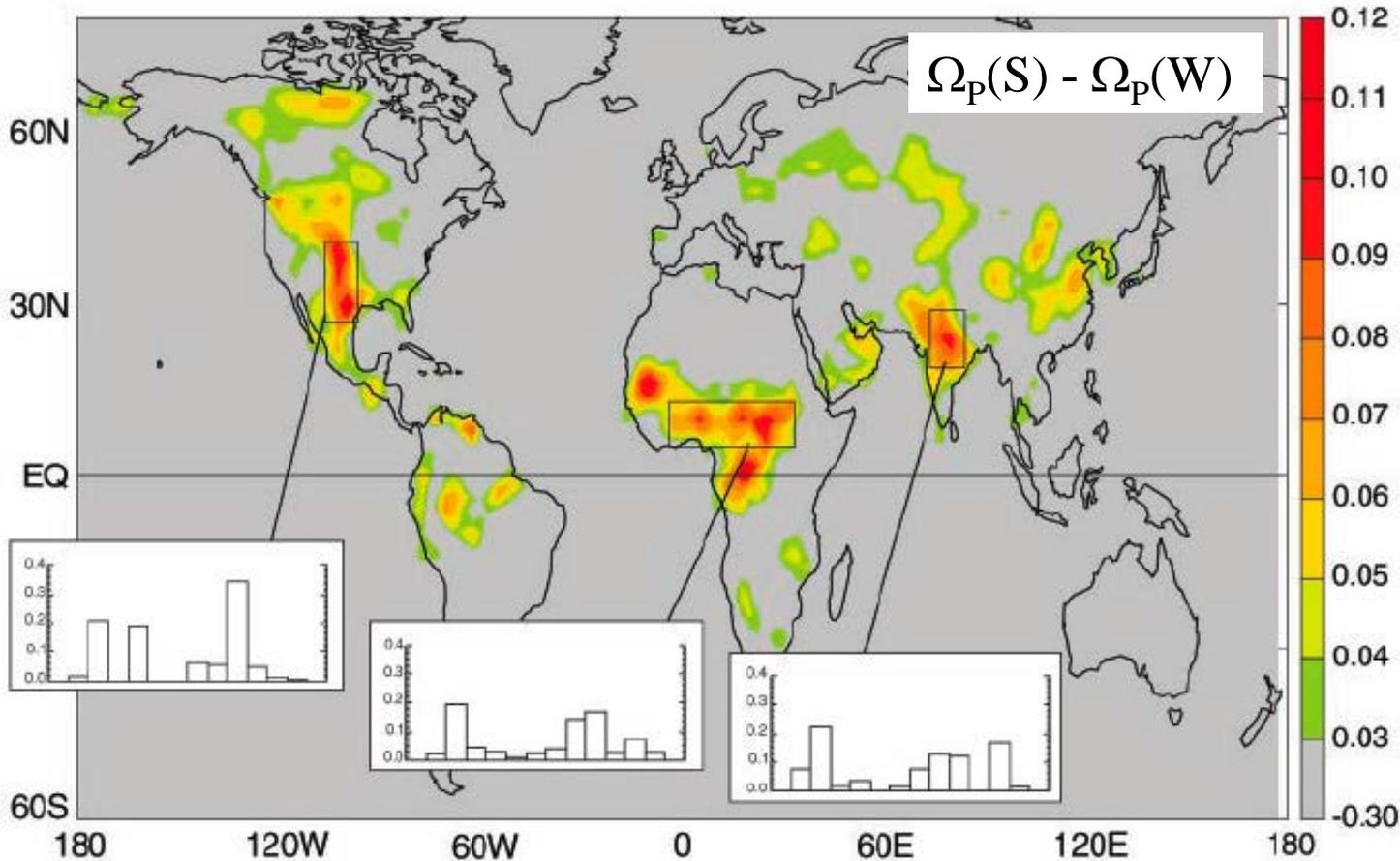


Wet climate regime



# Global Land-Atmosphere Coupling Experiment (GLACE)

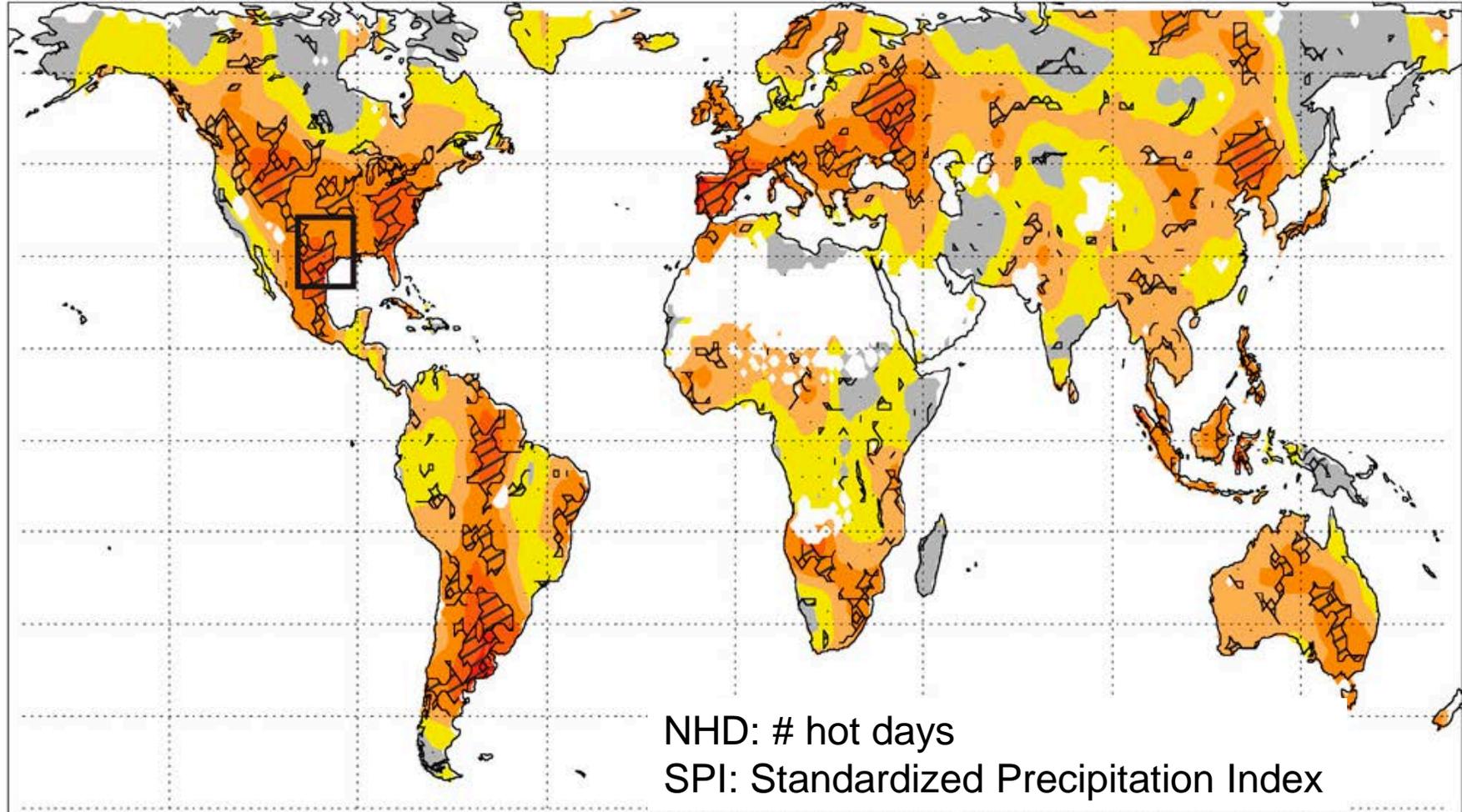
12 AGCMs, experiments for JJA 1994



**Strong SM-P  
coupling in  
transitional  
zones between  
dry and wet  
climates**

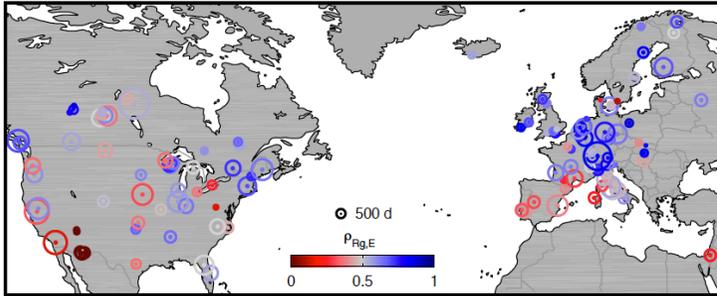
(Koster et al. 2004, Science)

## Correlation NHD E-Int and preceding 3mn SPI CRU



(Mueller and Seneviratne 2012, PNAS)

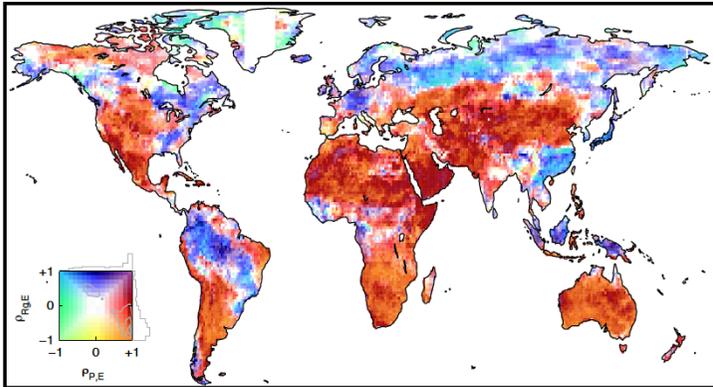
E drivers, FLUXNET dataset



$\rho(E, R_g)$   
- blue: high  
- red: small

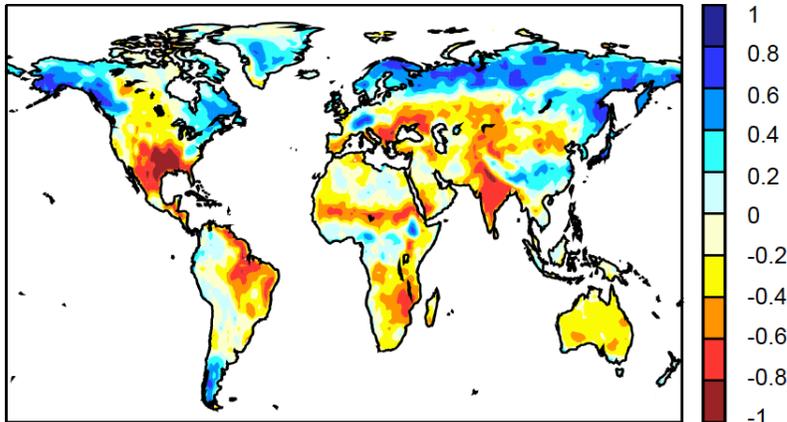
**Estimates of ET regimes**  
**Blue: Radiation limited**  
**Red: Soil moisture limited**

E drivers, GSWP dataset



$\rho(E, R_g)$  &  $\rho(E, P)$   
- blue: high  $\rho(E, R_g)$ , small  $\rho(E, P)$   
- red: small  $\rho(E, R_g)$ , high  $\rho(E, P)$

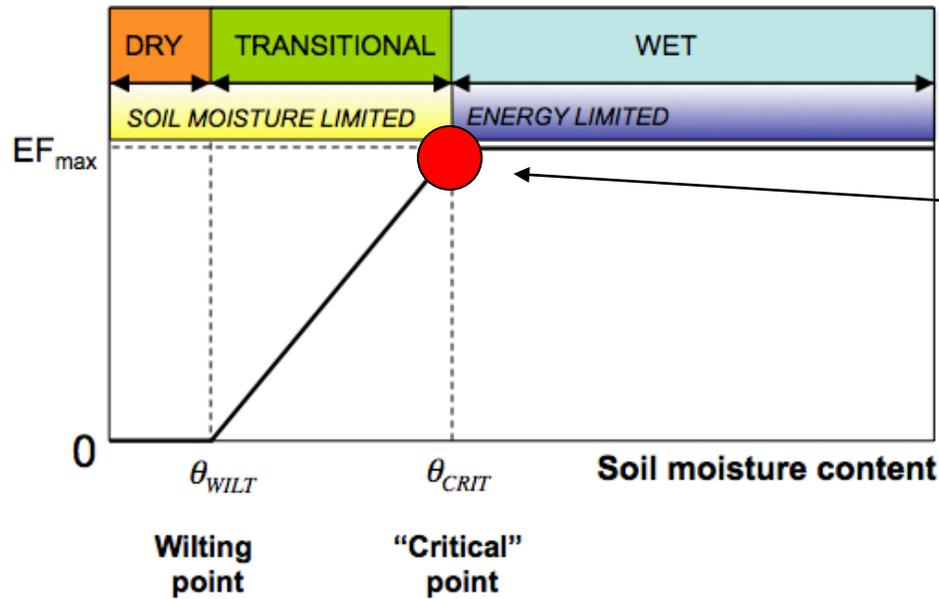
$\rho(E, T)$ , IPCC AR4, 1970-1989



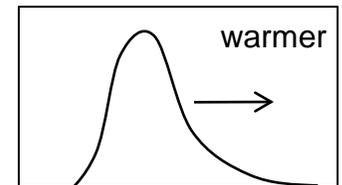
$\rho(E, T)$   
- blue: positive  $\rho(E, T)$   
- red: negative  $\rho(E, T)$

(Seneviratne et al. 2006, Nature;  
Teuling et al. 2009, GRL; Seneviratne et al. 2010, ESR)

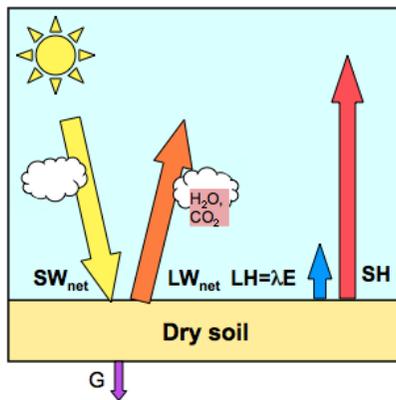
Evaporative fraction  $EF = \lambda E / R_n$



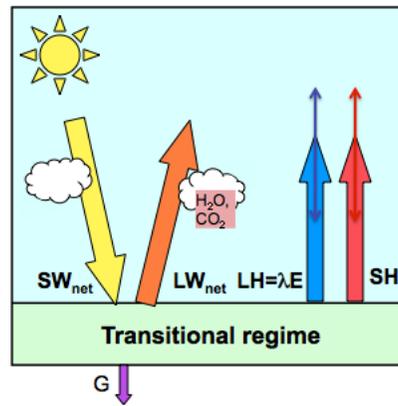
Discrete threshold implies non-linear effect & relevance for extreme events



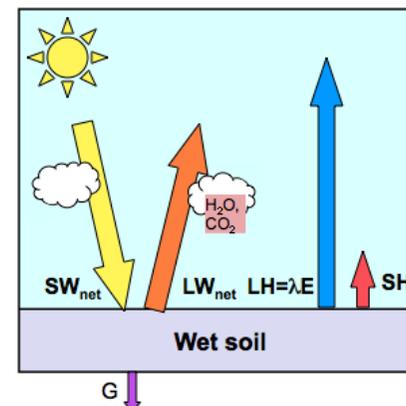
Dry climate regime



Transitional climate regime

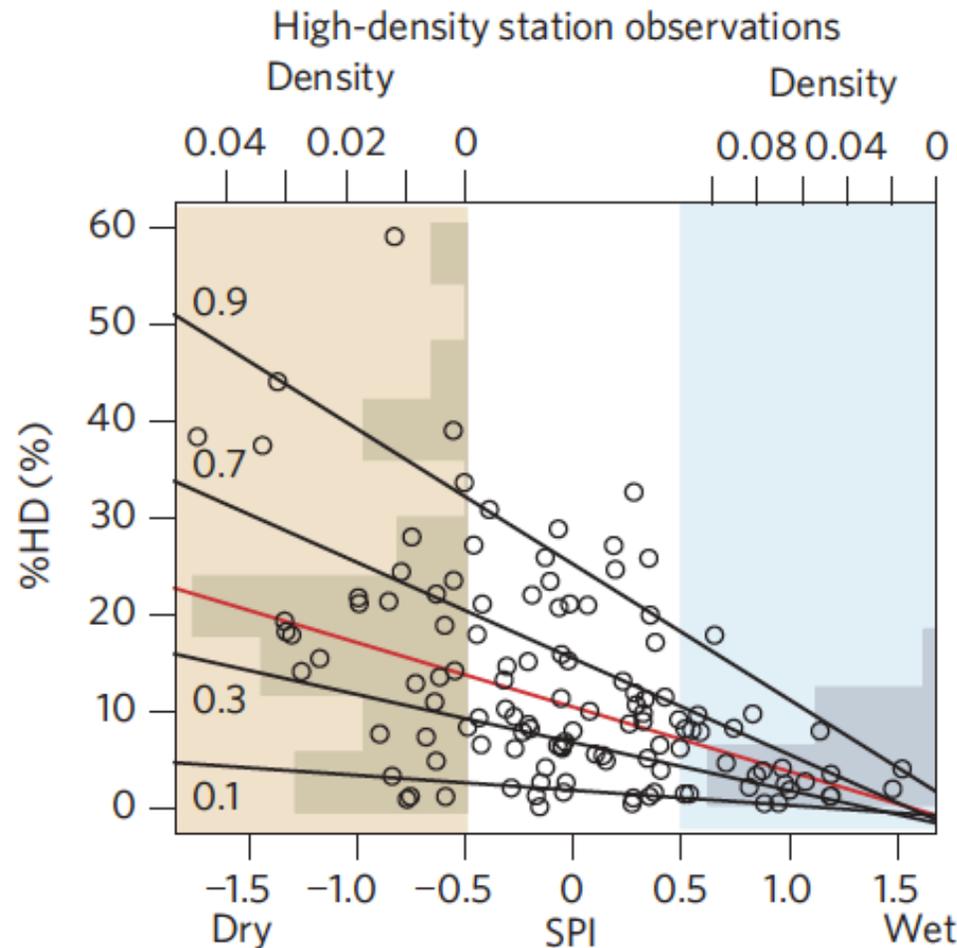


Wet climate regime



## Analysis for Southeastern Europe

Quantile regression of %HD with 6-month SPI



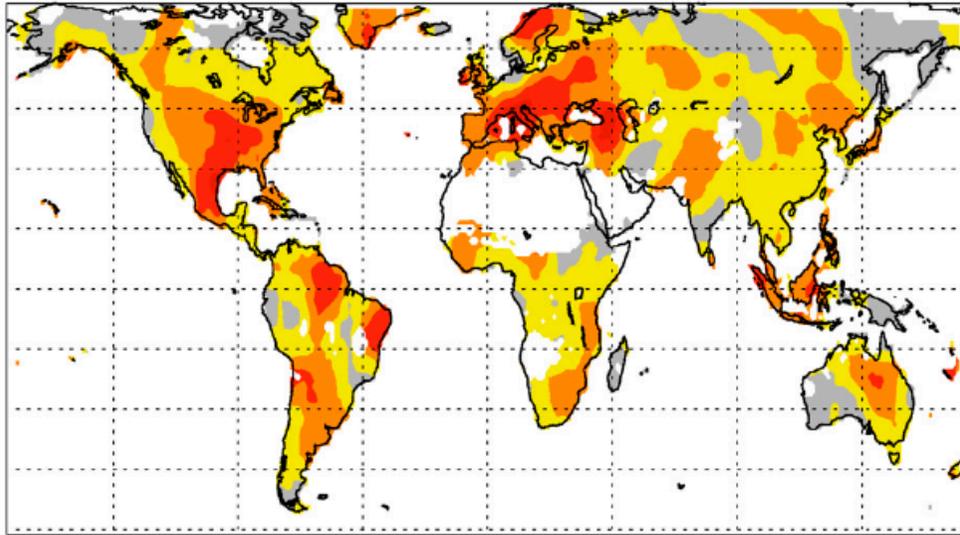
**Impact of soil moisture on hot extremes**

Regression lines: — 0.1, 0.3, 0.7, 0.9 %HD quantiles

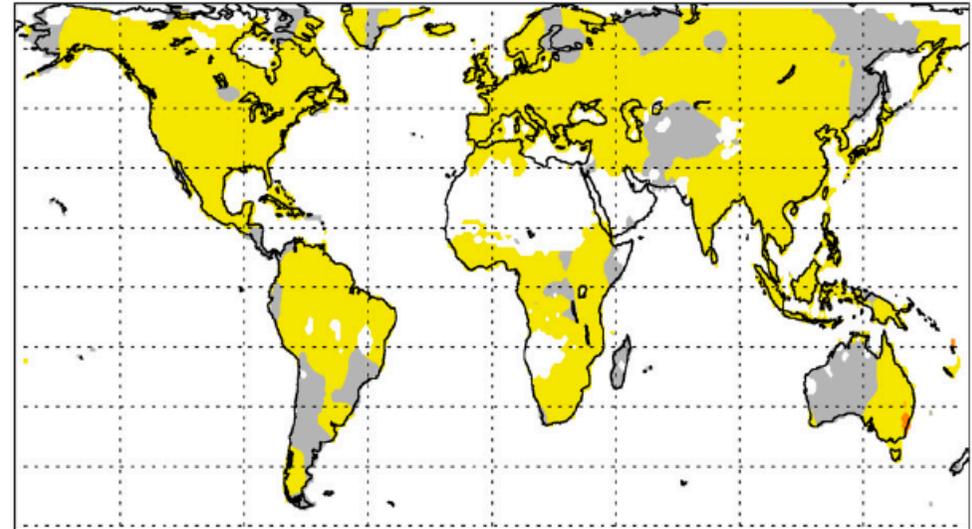
(Hirschi et al. 2011, Nature Geoscience)

## Quantile regression of NHD E-Int and preceding 3mn SPI CRU

90th percentile regression slope



10th percentile regression slope



NHD: # hot days

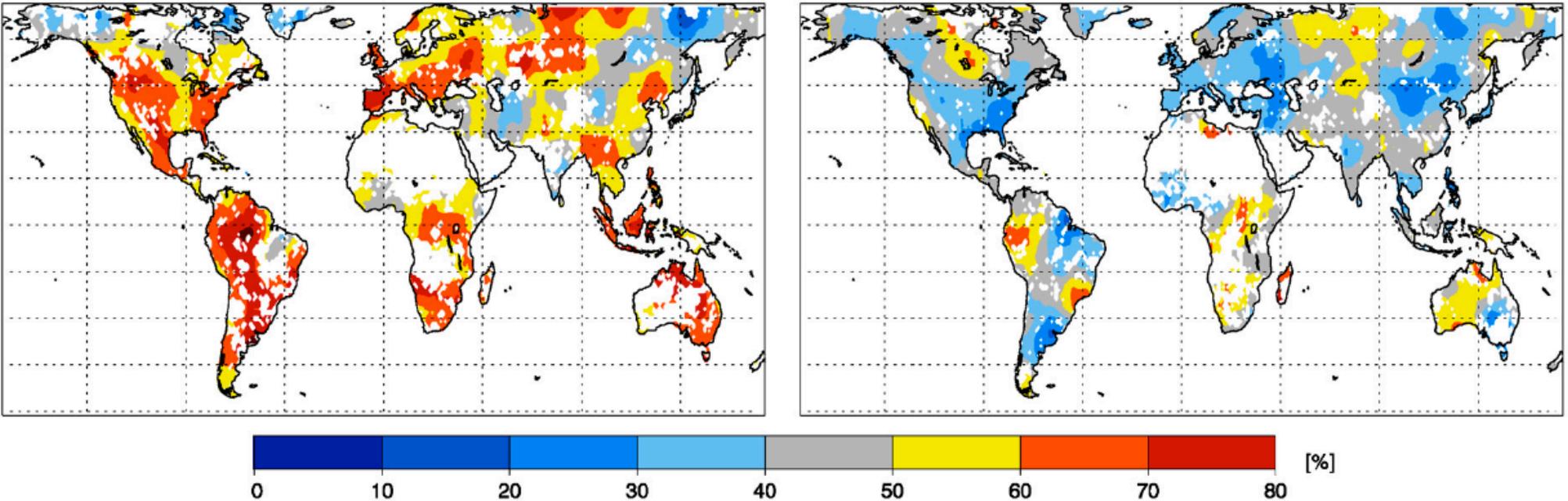
SPI: Standardized Precipitation Index

*(Mueller and Seneviratne 2012, PNAS)*

## Link to forecasting: conditional probability

Above avg. NHD after SPI < -0.8

Above avg. NHD after SPI > 0.8



NHD: # hot days

SPI: Standardized Precipitation Index

(Mueller and Seneviratne 2012, PNAS)

- Soil moisture affects evapotranspiration in transitional climate regions, thereby leading to impacts on temperature and precipitation
- Temperature: Widespread impacts, in particular for extreme events
- Soil moisture is a storage: → associated memory and forecasting potential

- Basic relevant land-climate feedbacks and observational evidence
- Diagnosing prediction potential from land surface initialization
  - GLACE-2: Forecasting of atmospheric variables
  - Drought forecasting
- Discussion
- Conclusions

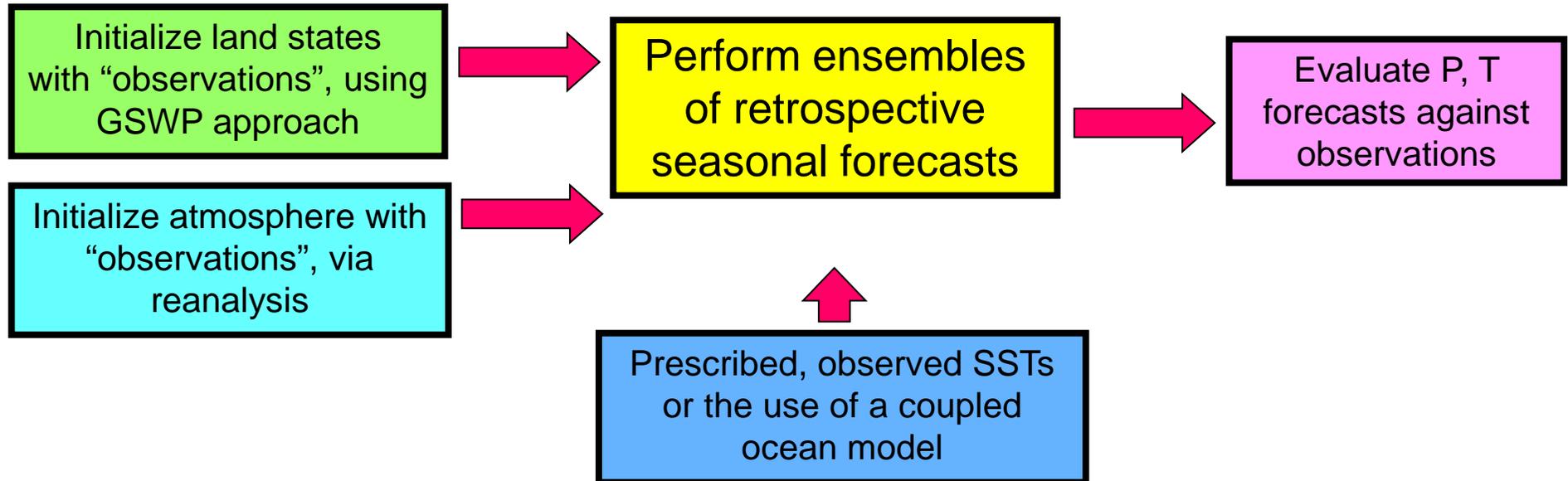
## **GLACE-2:** An international project aimed at quantifying soil moisture impacts on prediction skill.

Overall goal of GLACE-2: Determine the degree to which realistic land surface (soil moisture) initialization contributes to forecast skill (rainfall, temperature) at 1-2 month leads, using a wide array of state-of-the-art forecast systems.

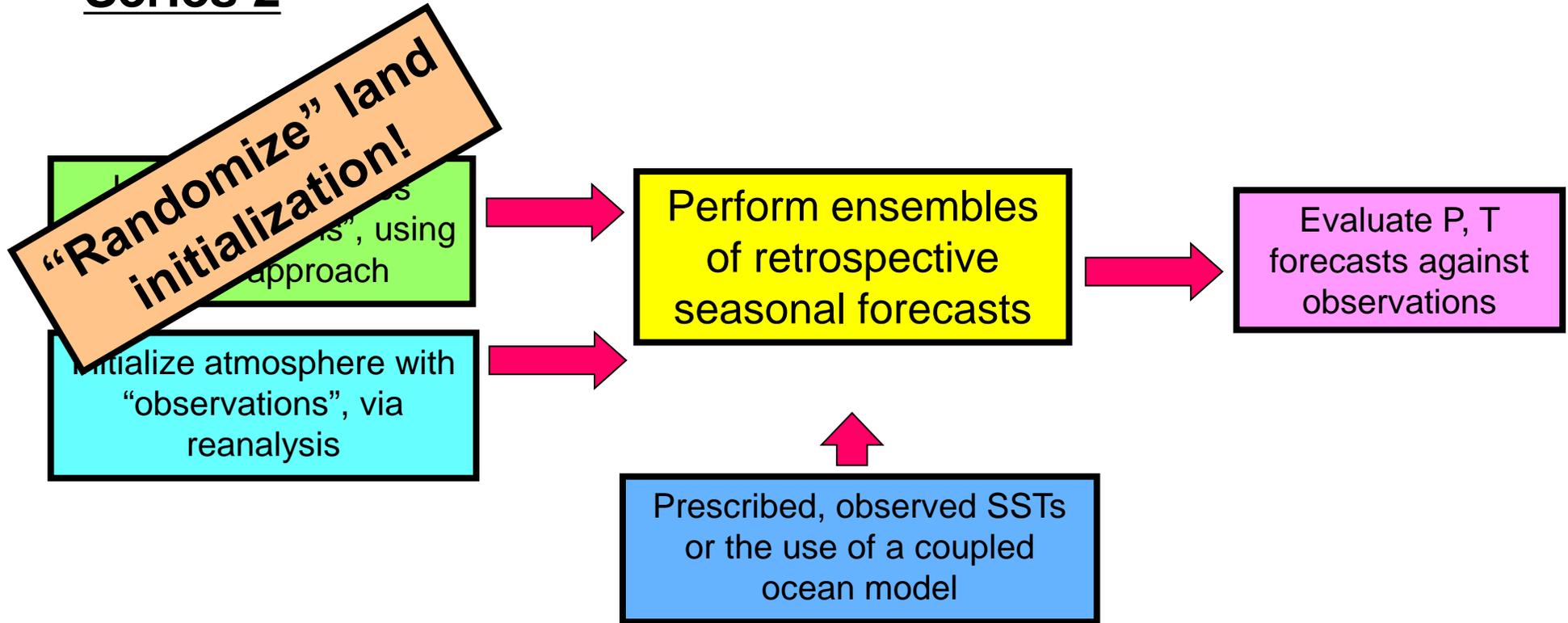


*(Koster et al. 2010, GRL; Koster et al. 2011, JHM)*

## Series 1



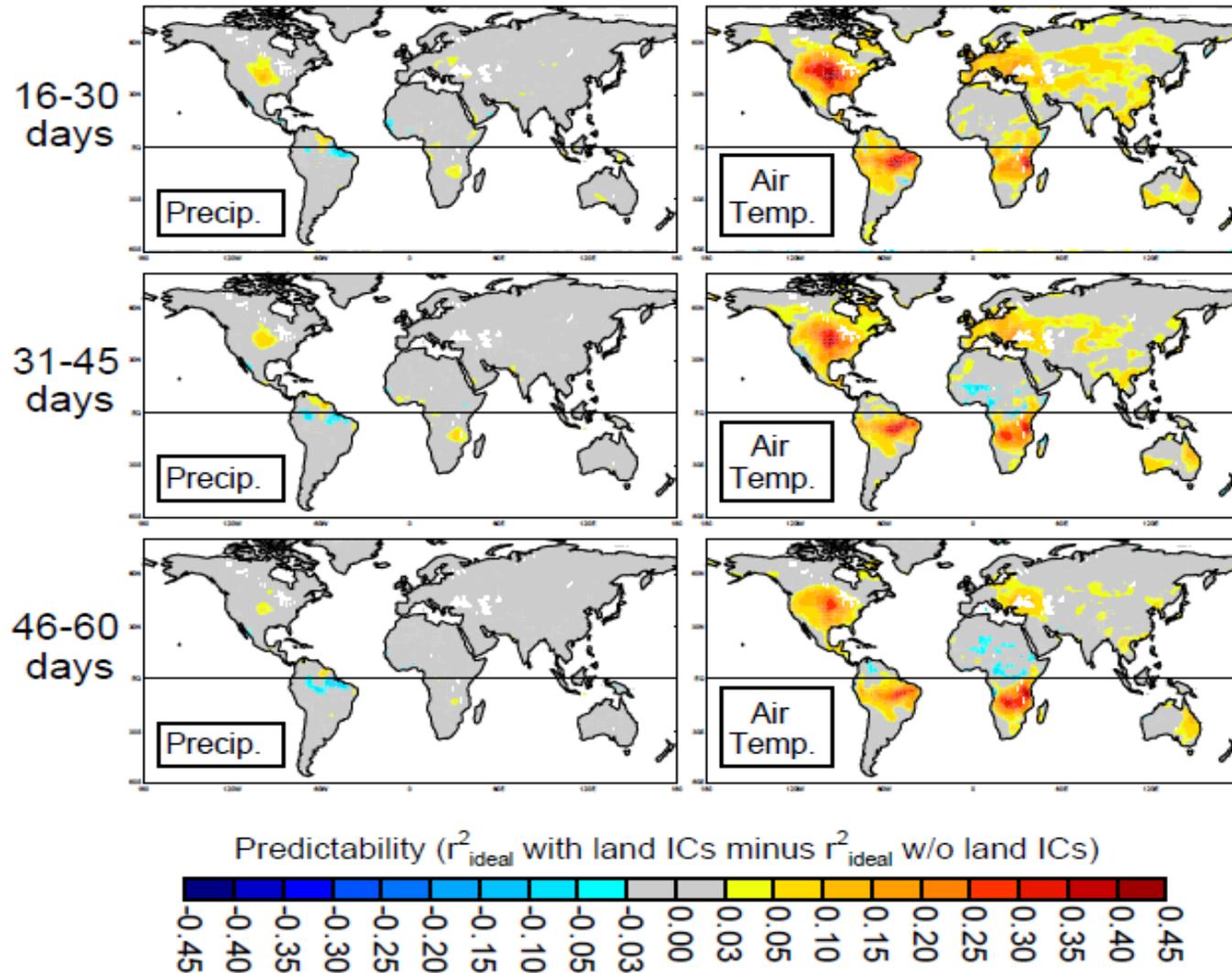
## Series 2



**Step 3:** Compare skill in two sets of forecasts; isolate contribution of realistic land initialization

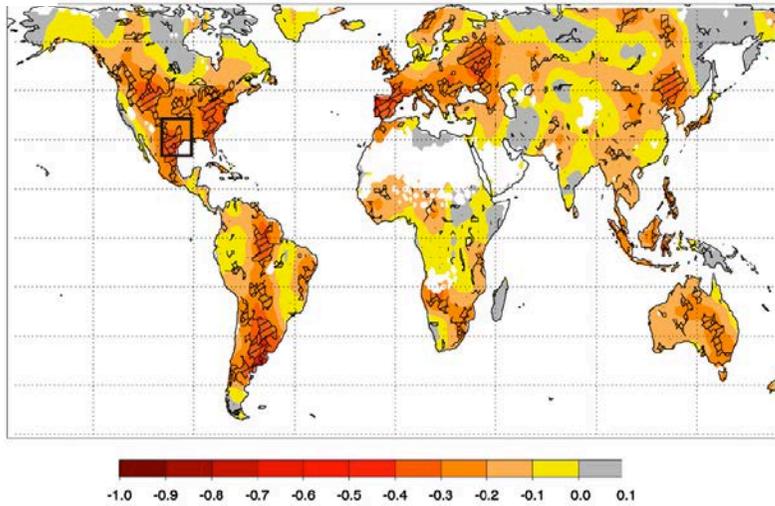


13 participating modeling systems

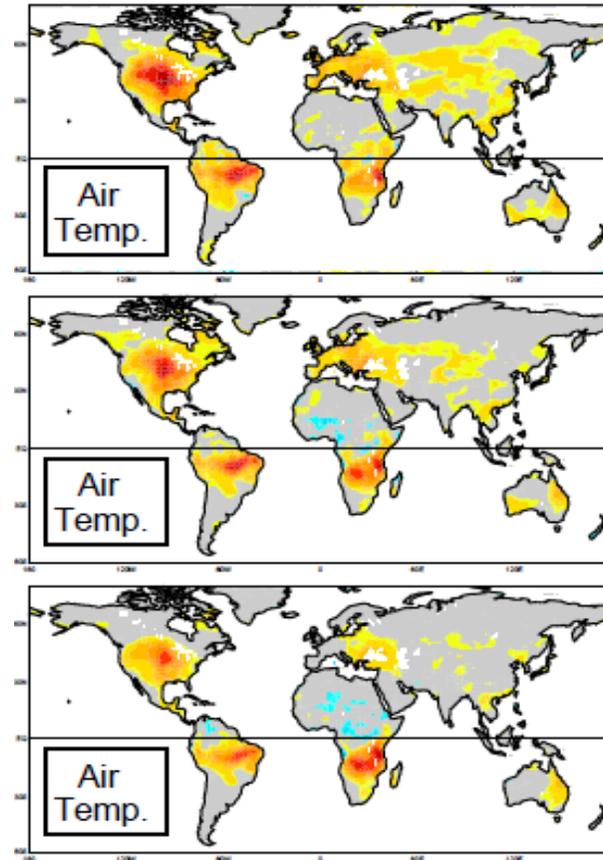


(Koster et al. 2011, JHM)

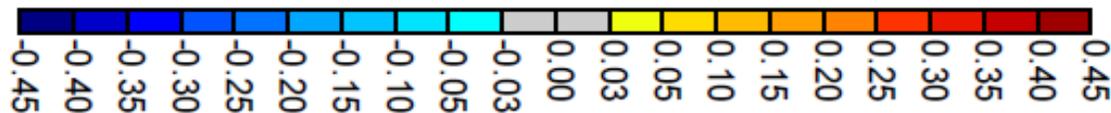
Correlation NHD E-Int and preceding 3mn SPI CRU



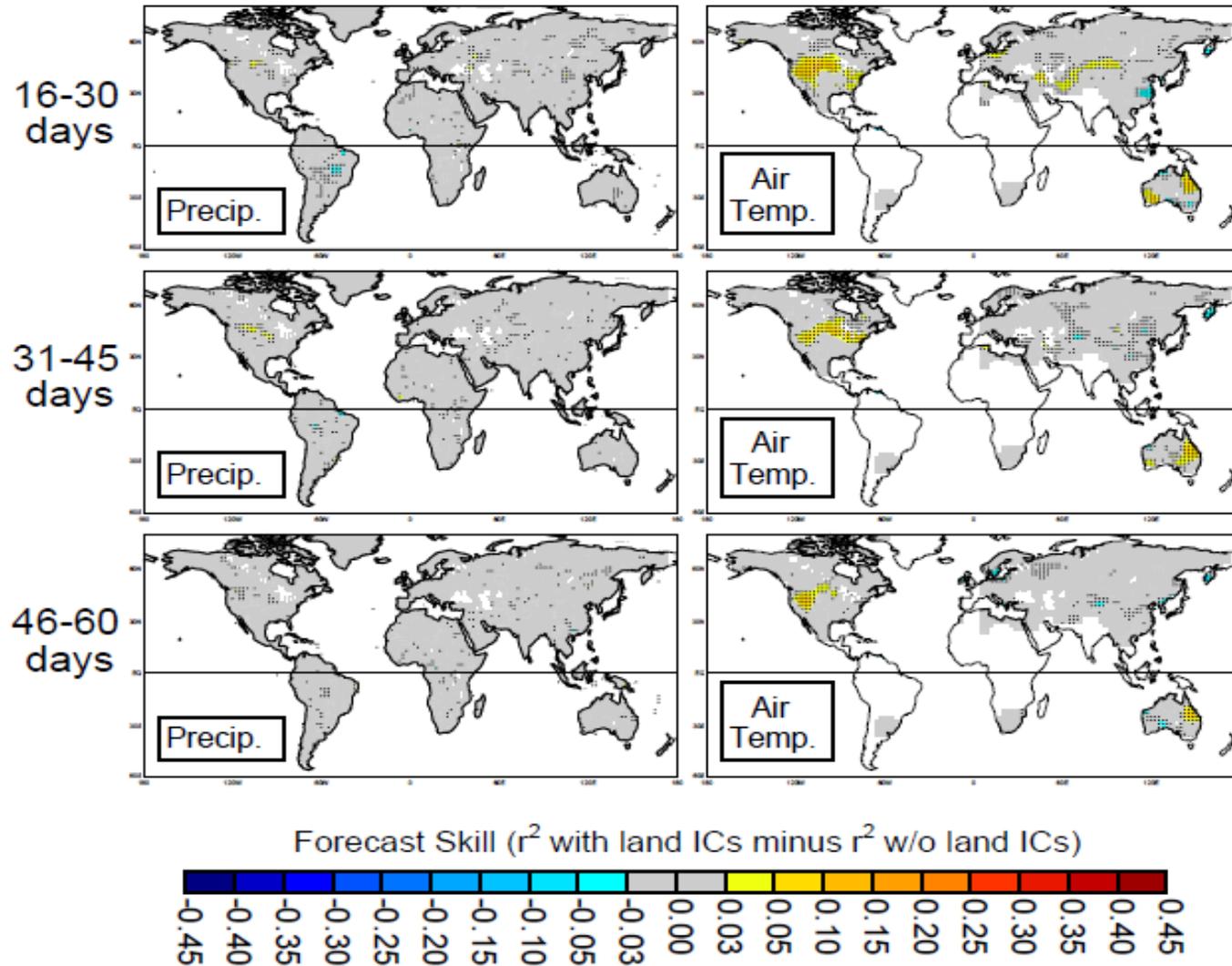
(Mueller and Seneviratne 2012, PNAS)



Predictability ( $r^2_{ideal}$  with land ICs minus  $r^2_{ideal}$  w/o land ICs)



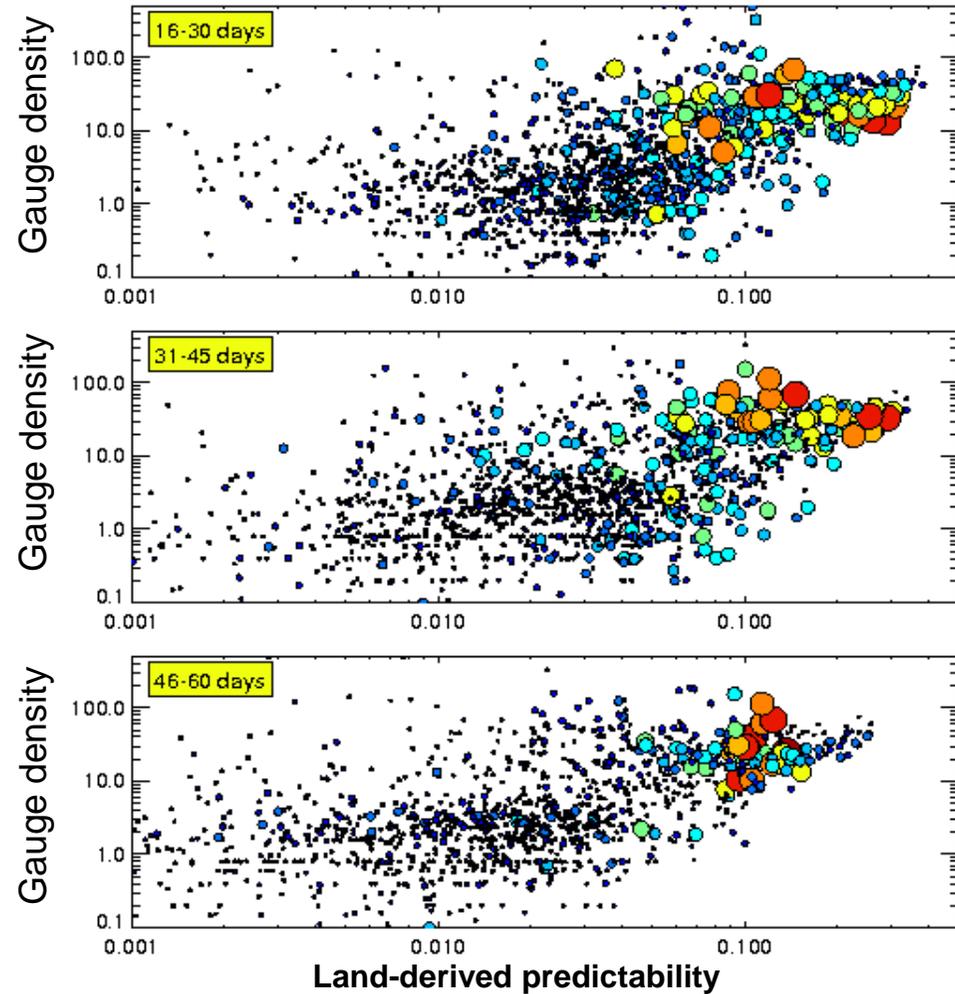
(Koster et al. 2011, JHM)



(Koster et al. 2011, JHM)

Forecast skill levels are highest in regions with both:

- a) some inherent model “predictability”, and
- b) an adequate observational network for accurate initialization

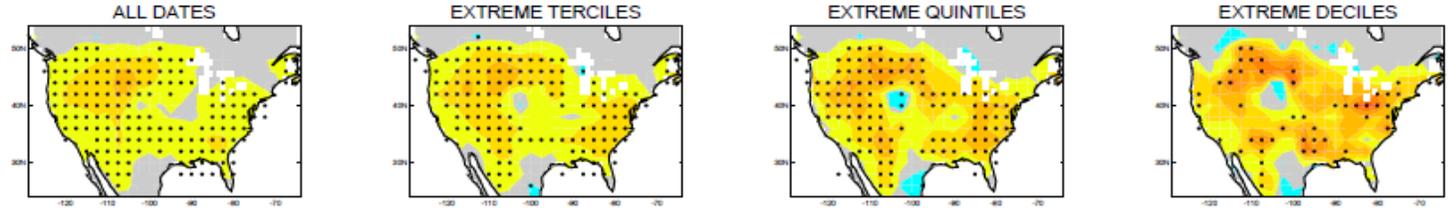


(Koster et al. 2011, JHM)

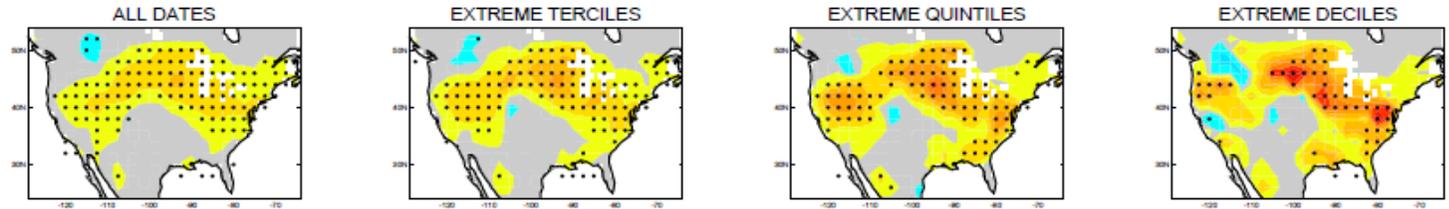
(conditioned on strength of local initial soil moisture anomaly)

**all points**      **Extreme terciles**      **Extreme quintiles**      **Extreme deciles**

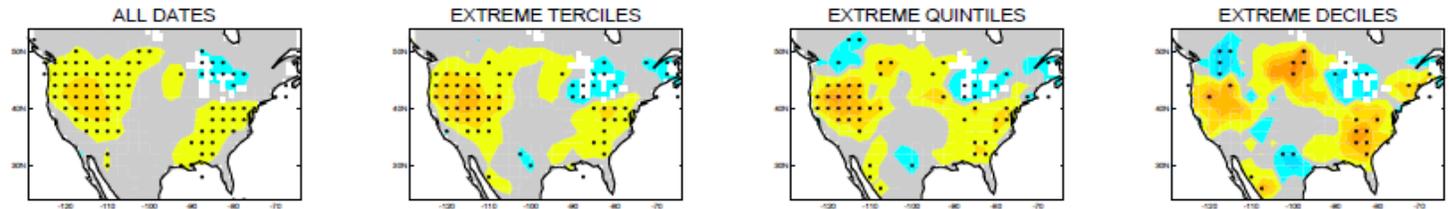
**16-30 days**



**31-45 days**



**46-60 days**



Dates for conditioning vary w/location

Forecast skill:  $r^2$  with land ICs vs  $r^2$  w/o land ICs



(Koster et al. 2010, GRL)

(conditioned on strength of local initial soil moisture anomaly)

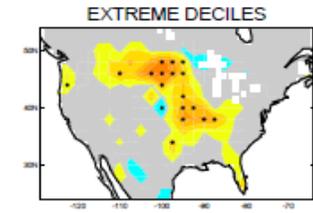
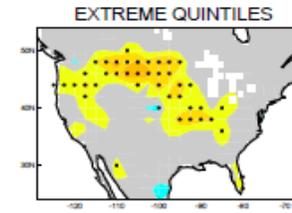
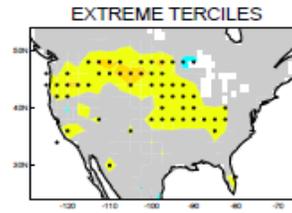
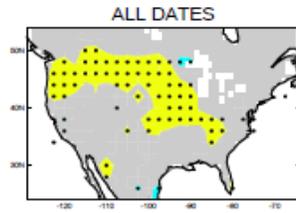
**all points**

**Extreme  
terciles**

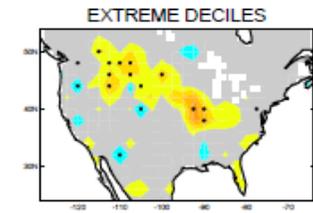
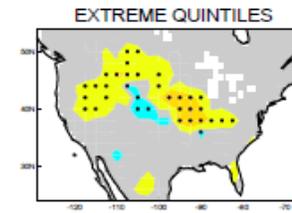
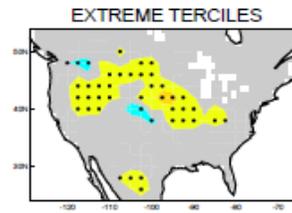
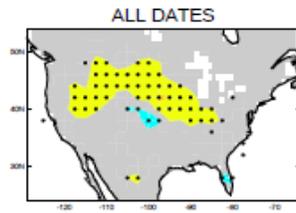
**Extreme  
quintiles**

**Extreme  
deciles**

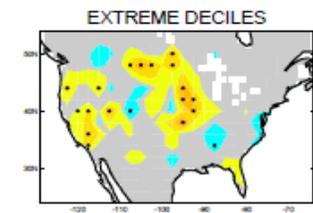
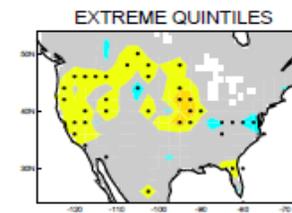
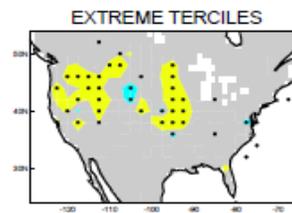
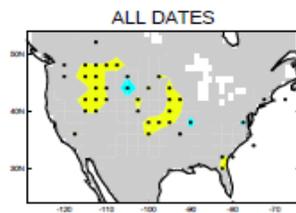
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Dates for conditioning vary w/location

Forecast skill:  $r^2$  with land ICs vs  $r^2$  w/o land ICs



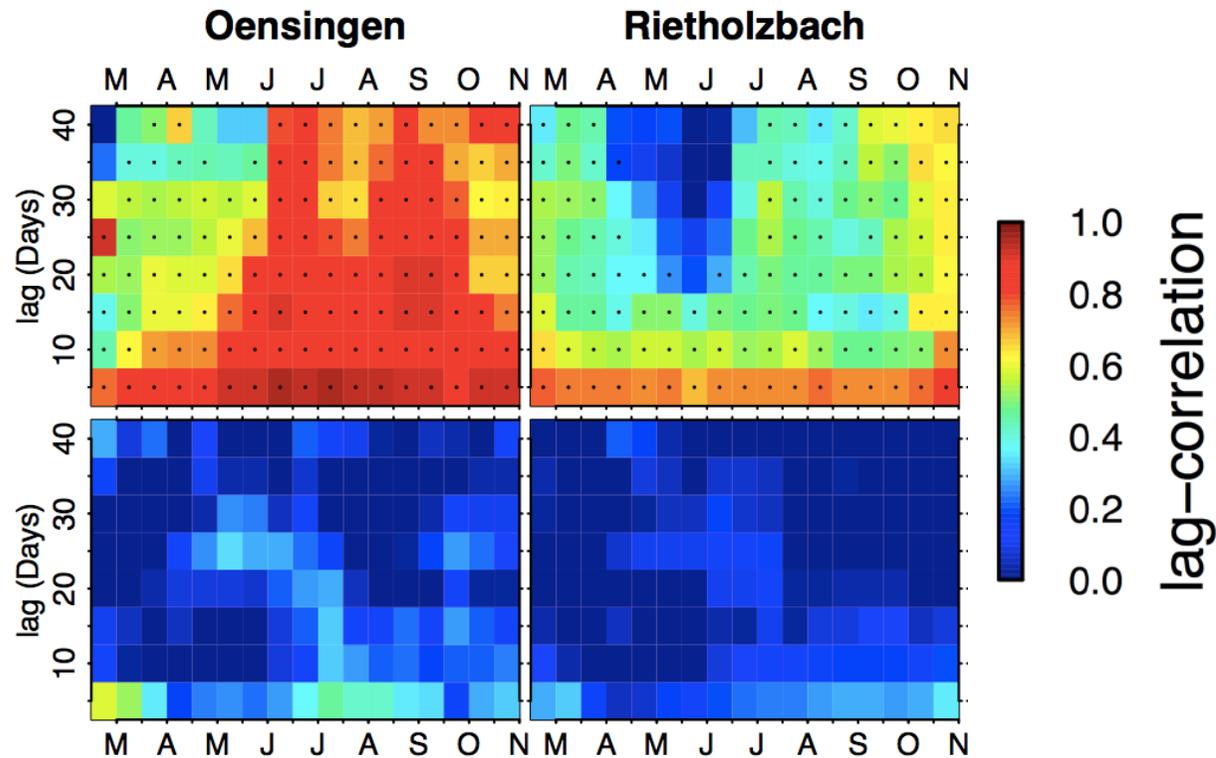
(Koster et al. 2010, GRL)

- Moderate skill for temperature and precipitation (in some regions) **up to 2 months** → source of skill in mid-latitude regions where ocean teleconnections play a rather minor role!
- Highest potential for **temperature** and for **extreme conditions**
- **Ideal skill much higher than actual skill:**
  - Issue with quality of initial data: poor coverage in most regions
  - Models overestimate skill?
  - Interestingly, observational data (for extremes) suggests similar map as ideal skill: Problem in translation of information?
- Area of current research development: **Suggests substantial potential that could be tapped in operational applications**

- Soil moisture is characterized by long persistence
- This implies high potential for improved **early warning and subseasonal forecasting of drought based on land surface information alone** (from several weeks to several months)

Soil moisture persistence (“memory”)

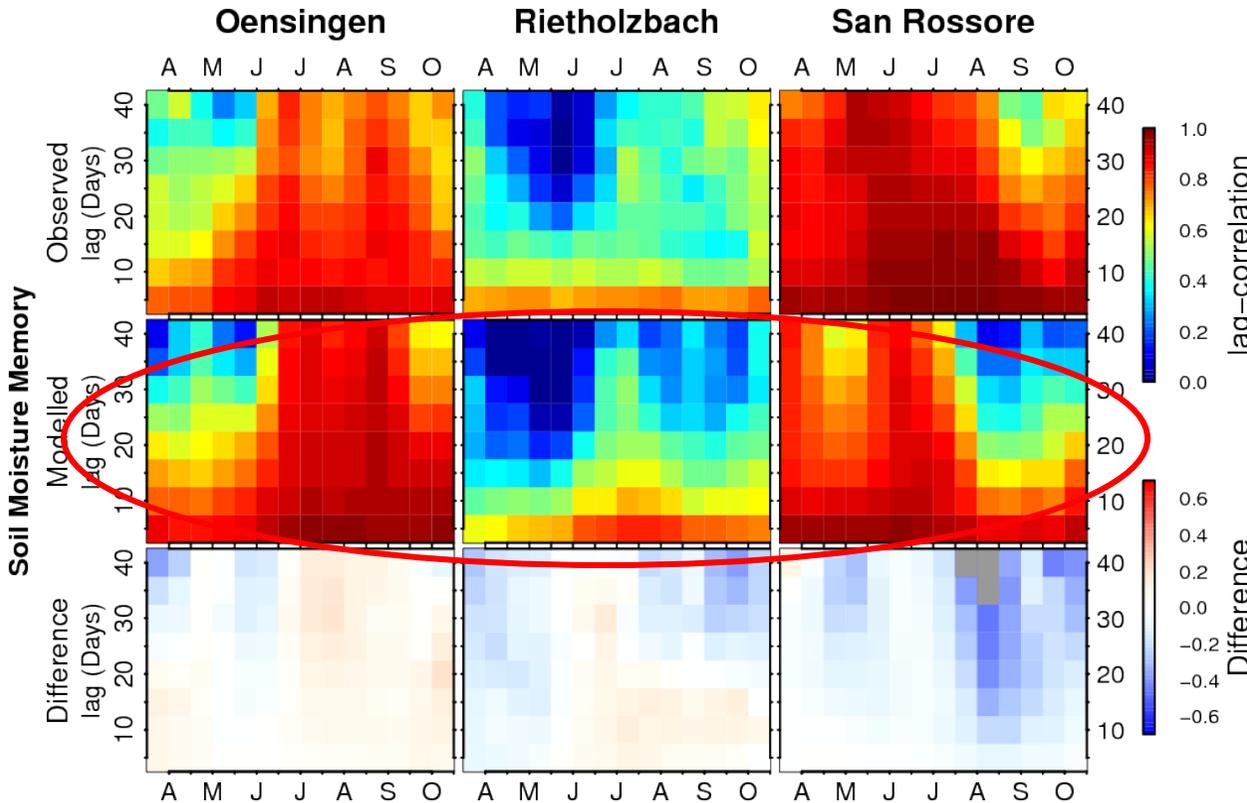
Temperature persistence



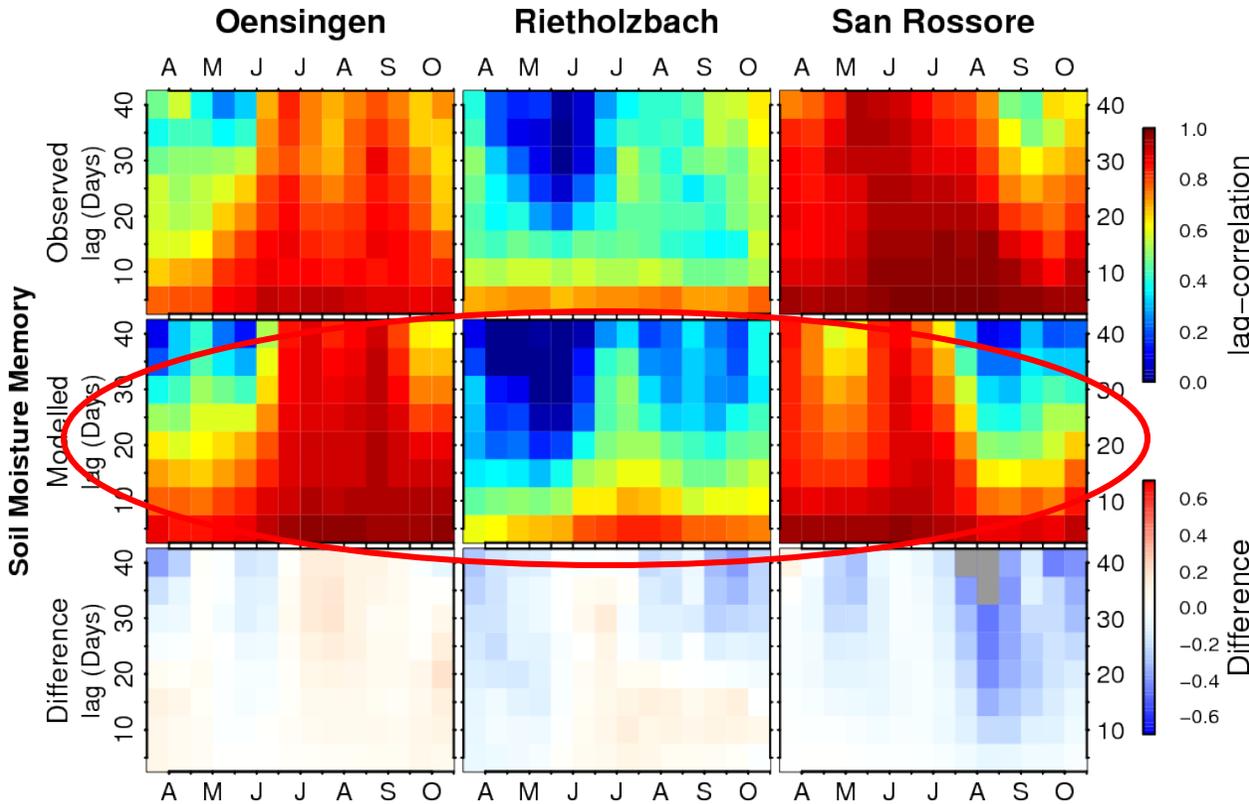
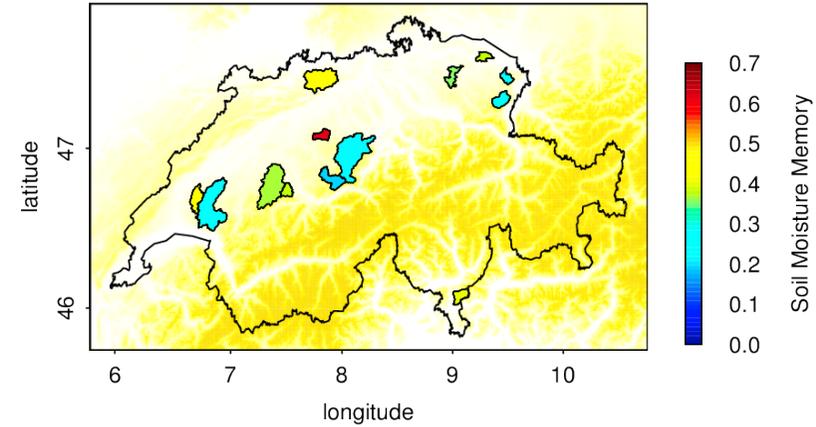
(Orth and Seneviratne 2012, JGR)



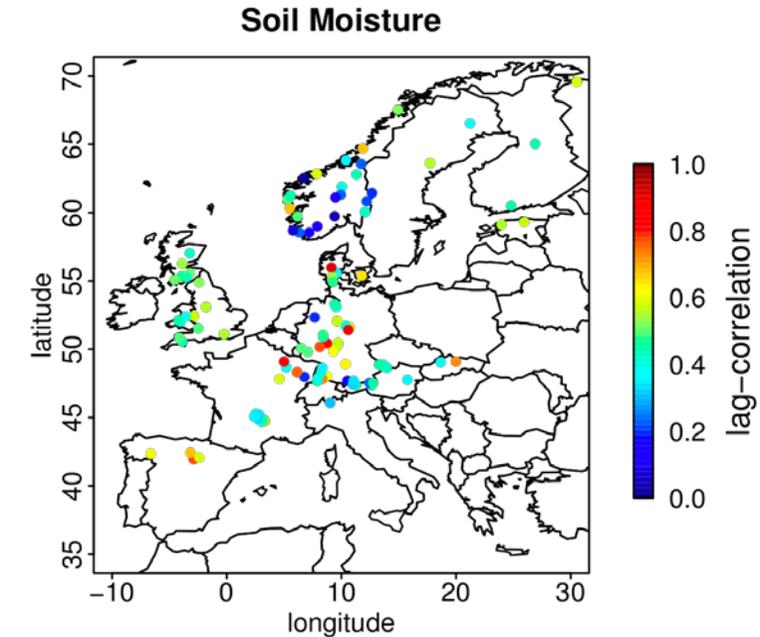
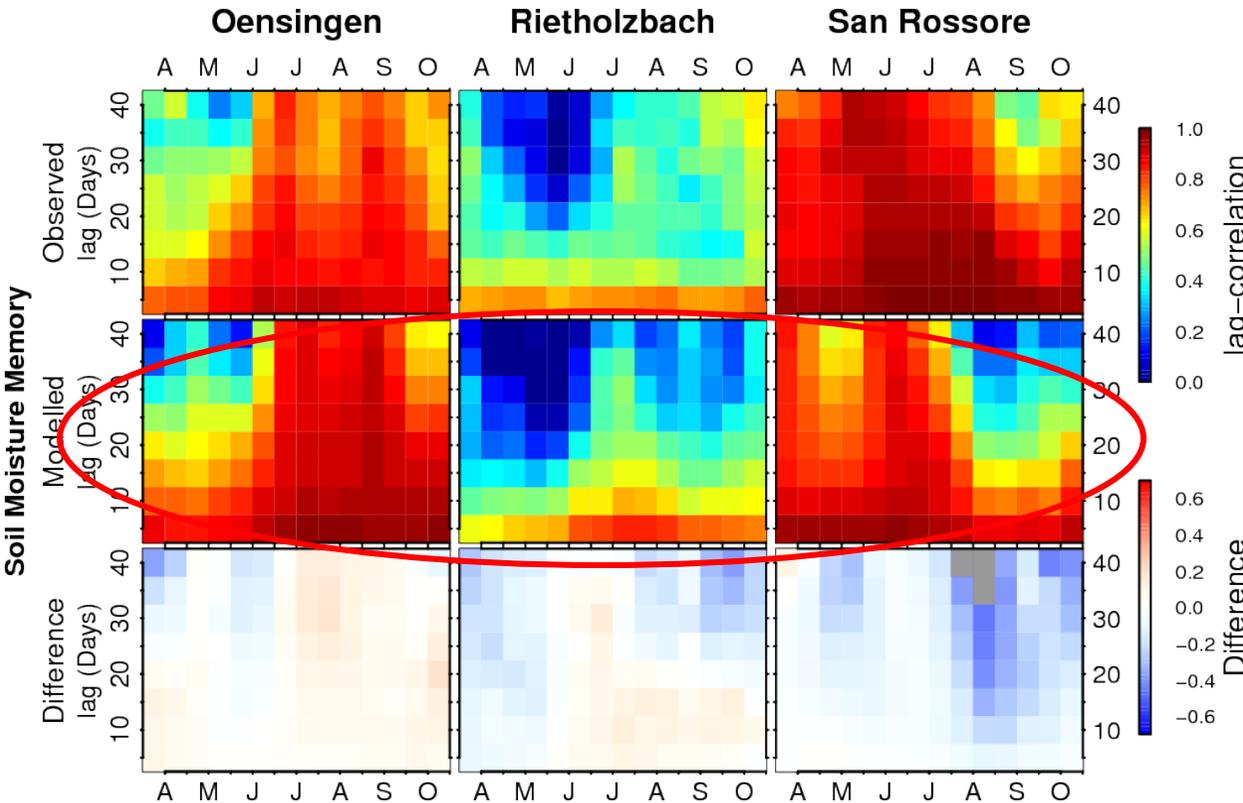
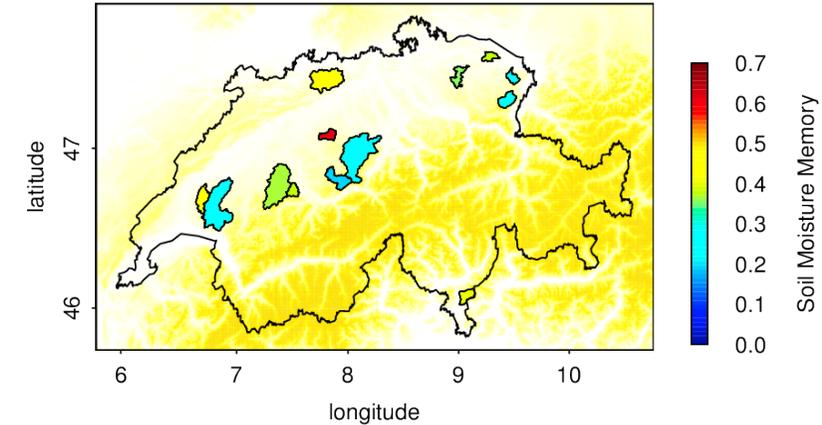
Good approximation of memory using simple water-balance model only calibrated with runoff observations (no soil moisture and ET information)



(Orth et al. submitted, JHM)



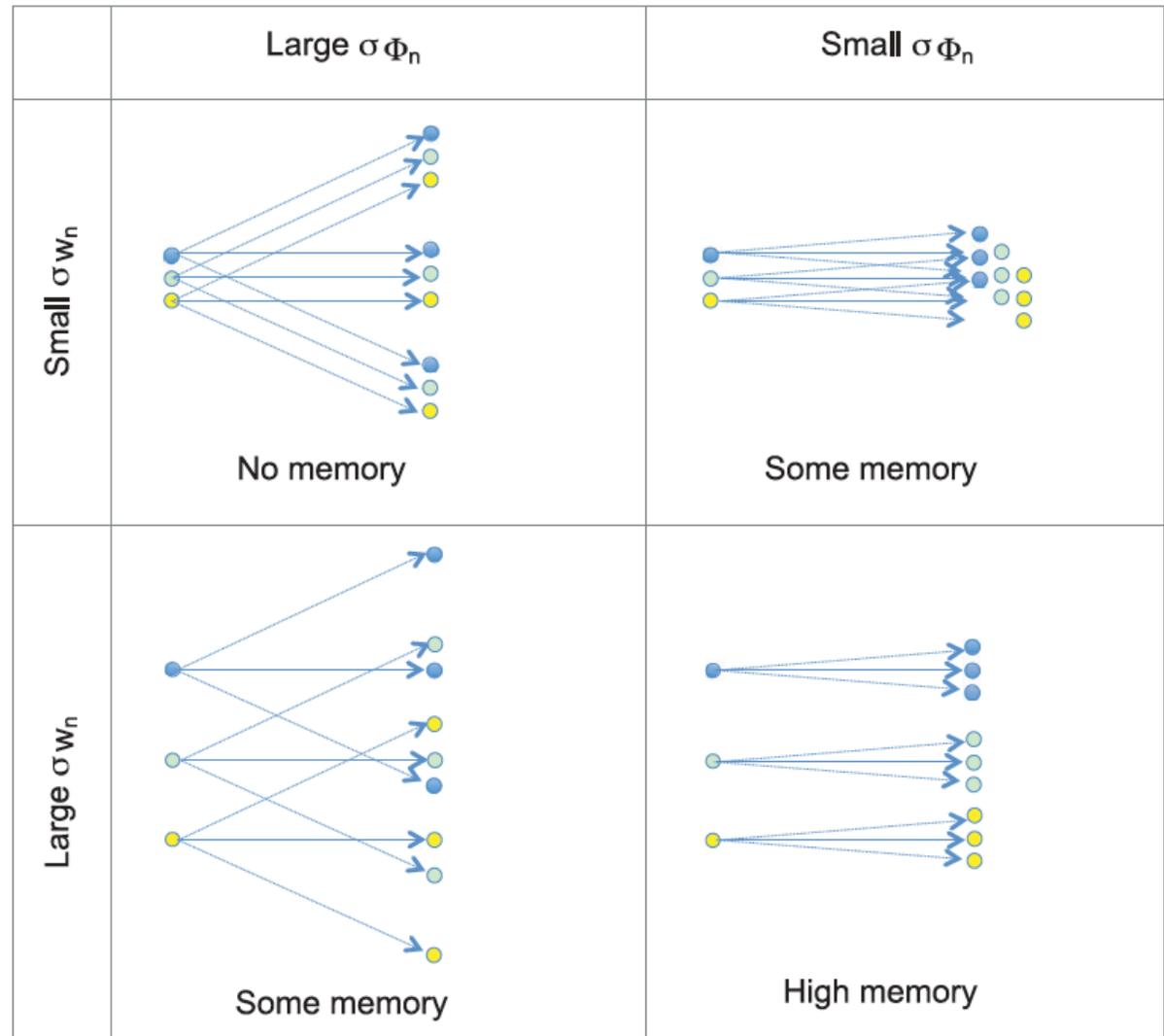
(Orth et al. submitted, JHM; Orth and Seneviratne, in prep.)



(Orth et al. submitted, JHM; Orth and Seneviratne, in prep.)

Analytical decomposition:  
4 main terms controlling  
memory (Koster and Suarez 2001,  
Seneviratne and Koster 2012)

- Spread of initial anomalies ( $\sigma_{w_n}$ )
- Spread of forcing ( $\sigma_{\Phi_n}$ )
- Damping factors (sensitivity of ET and runoff to soil moisture)
- Enhancing factors (correlation with subsequent forcing)



(Seneviratne and Koster 2012, JHM)

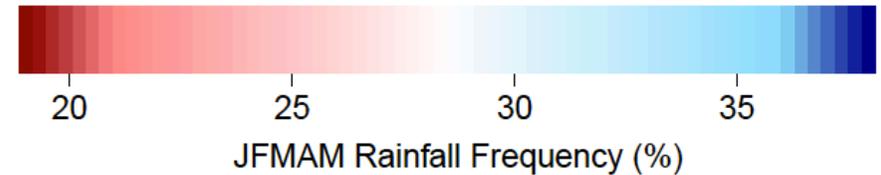
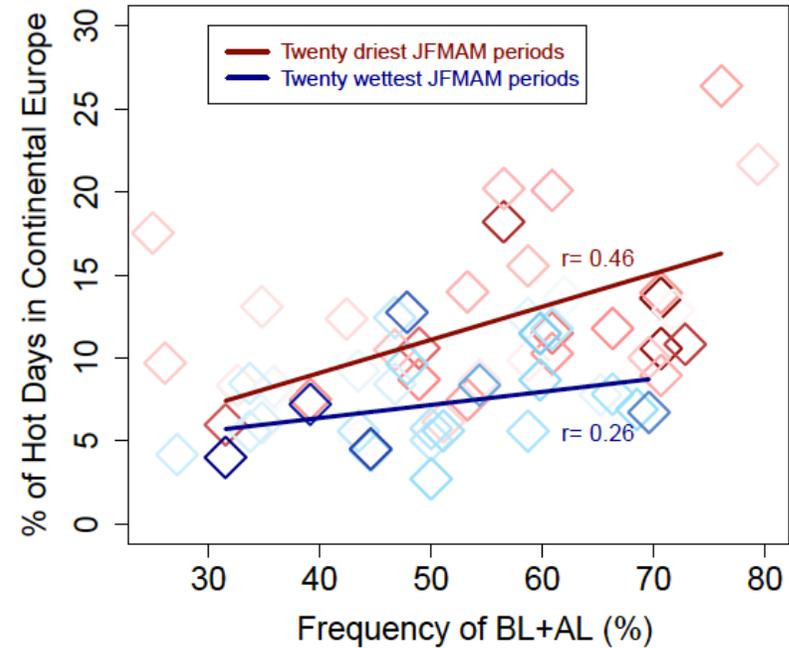
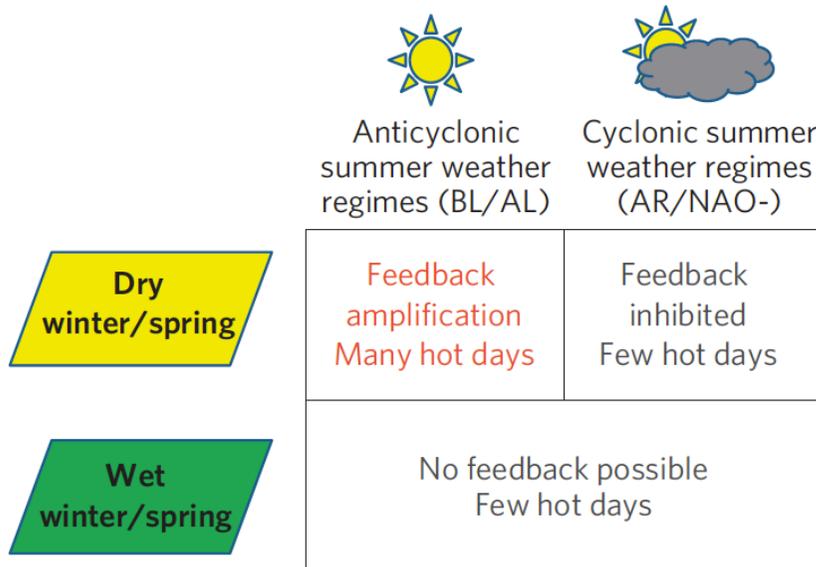
- Independently of possible feedbacks to the atmosphere, the **soil moisture (and groundwater) persistence implies potential skill for land hydrology** (soil moisture, runoff):  
Could be used for drought early warning and forecasting
- Memory patterns can be approximated with a simple water balance model calibrated with runoff observations → estimation of persistence characteristics for wide areas

- Basic relevant land-climate feedbacks and observational evidence
- Diagnosing prediction potential from land surface initialization
  - GLACE-2: Forecasting of atmospheric variables
  - Drought forecasting
- Discussion
  - Combined effects of large-scale circulation forcing and soil moisture feedbacks
  - Soil moisture initialization
  - Land-climate feedbacks in the context of climate change
- Conclusions

**European analysis:** High percentage of hot days found for combination of  
1) dry springs and 2) anticyclonic summer weather regimes

*(Quesada et al. 2012, Nature Climate Change, published online)*

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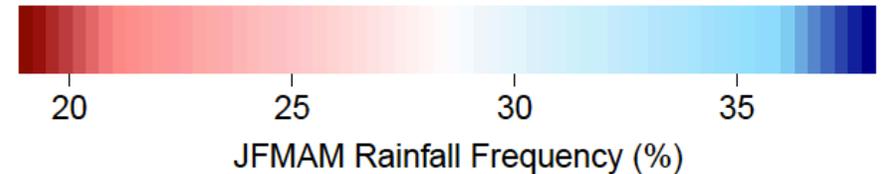
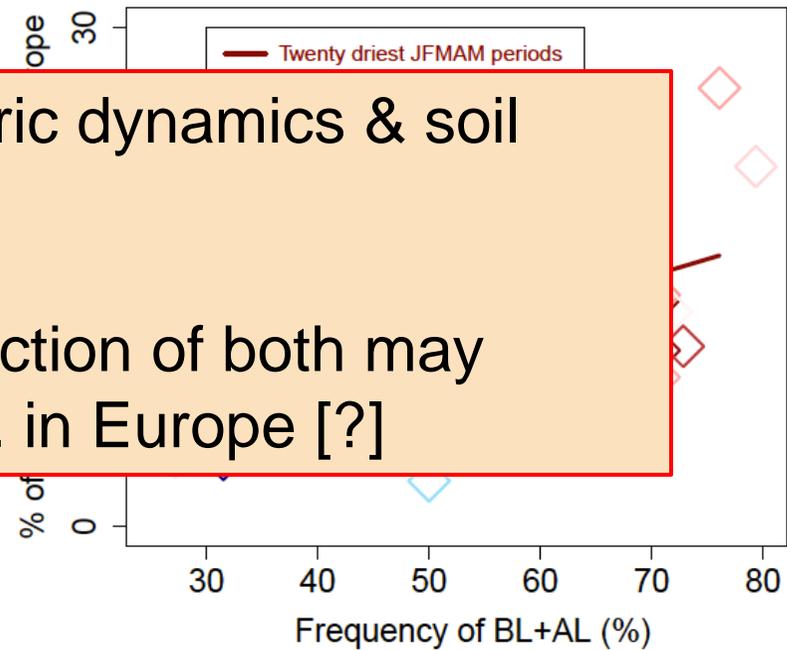
Information on both atmospheric dynamics & soil moisture state are critical

Modest improvements in prediction of both may substantially improve skill, e.g. in Europe [?]

Dry winter/spring

Wet winter/spring

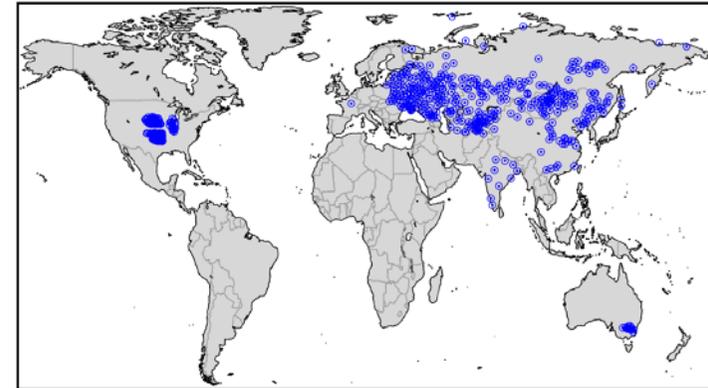
No feedback possible  
Few hot days

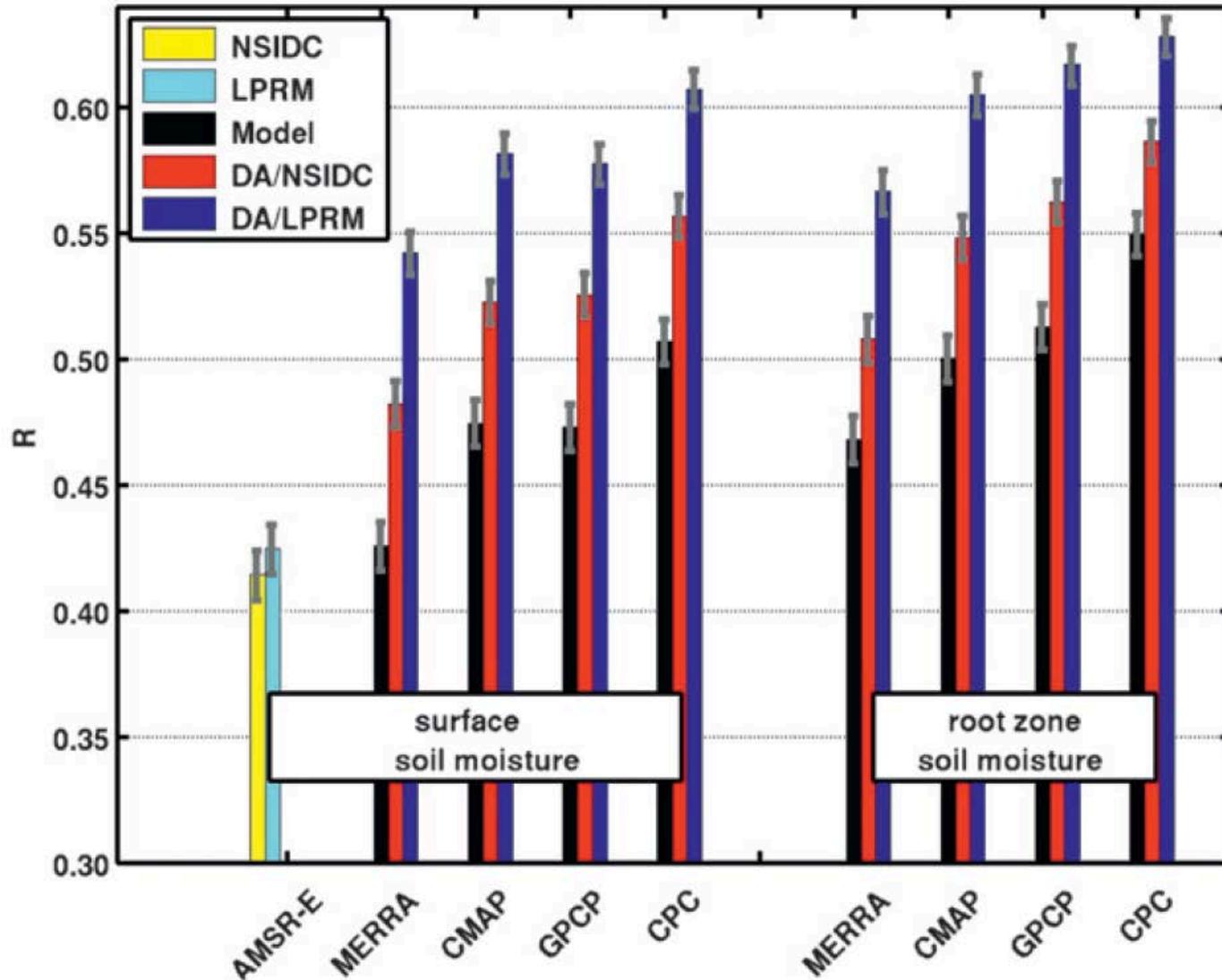


(Quesada et al. 2012, Nature Climate Change, published online)

- A major issue for any forecast using land surface information is the derivation of suitable initial conditions:
  - Few ground soil moisture networks (but soil moisture is newly characterized as essential climate variable, GCOS 2010)
  - Standard approach is to derive land surface model estimates using observation-based forcing (nonetheless also in this case data availability is an issue)
  - In development: Also use of remote-sensing based surface soil moisture estimates (e.g. microwave) using data assimilation, e.g. at NASA and ECMWF

Global Soil Moisture Data Bank

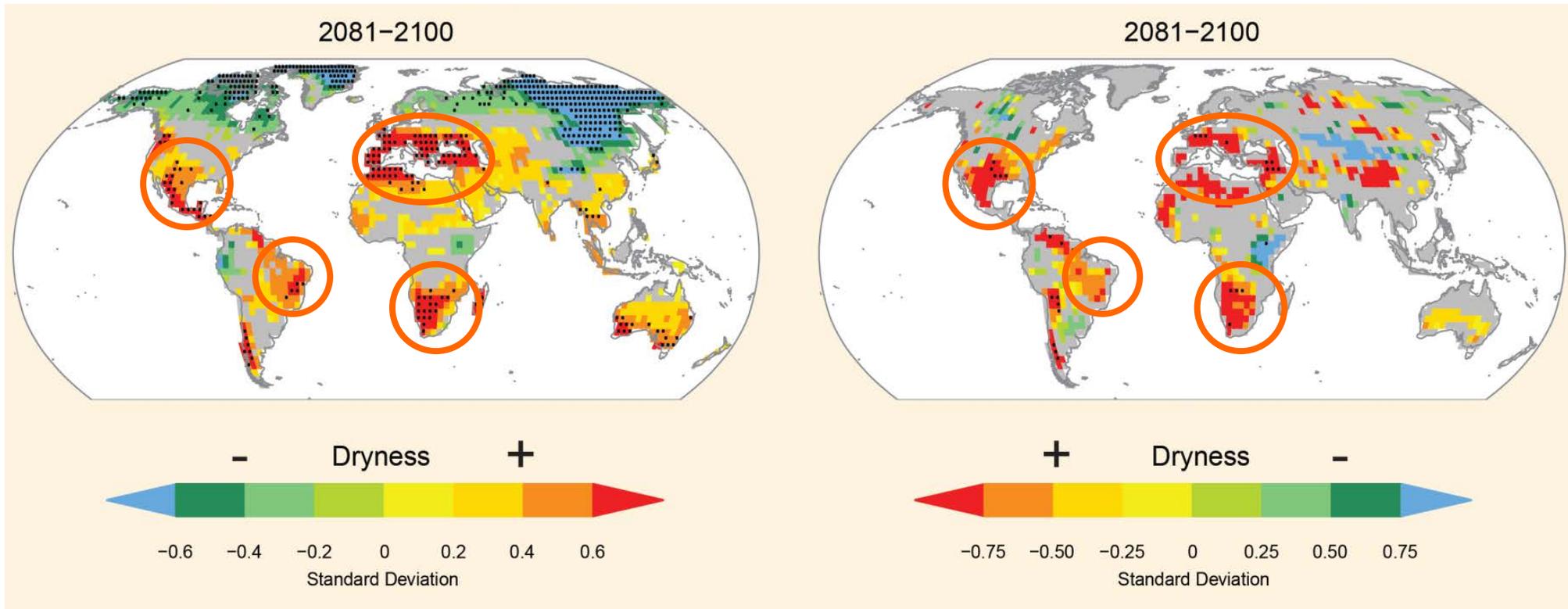




Correlation with surface soil moisture data (SCAN network, US)

(Liu et al. 2011, JHM)

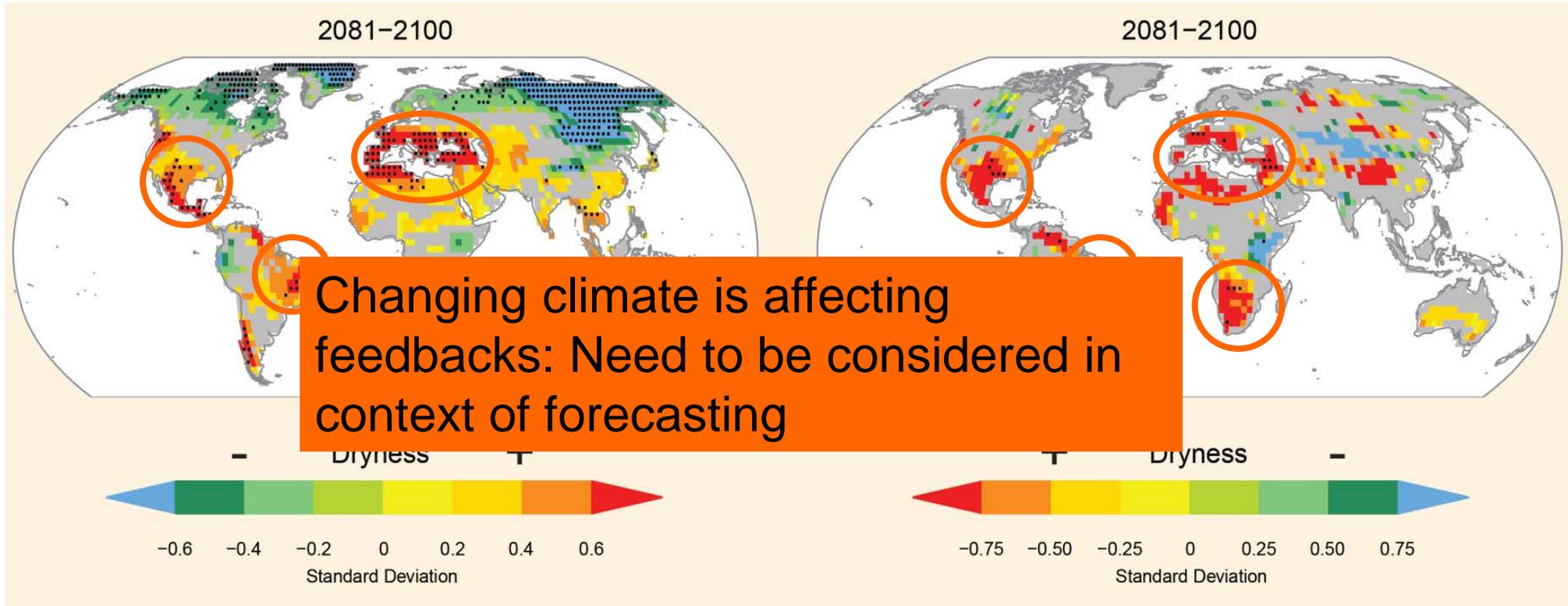
IPCC SREX (2012): <http://ipcc-wg2.gov/SREX/>



(IPCC 2012; Summary for Policymakers & Chapter 3)

Consistent projections of increased dryness in the Mediterranean region, central Europe, central North America, Central America and Mexico, northeast Brazil, and southern Africa

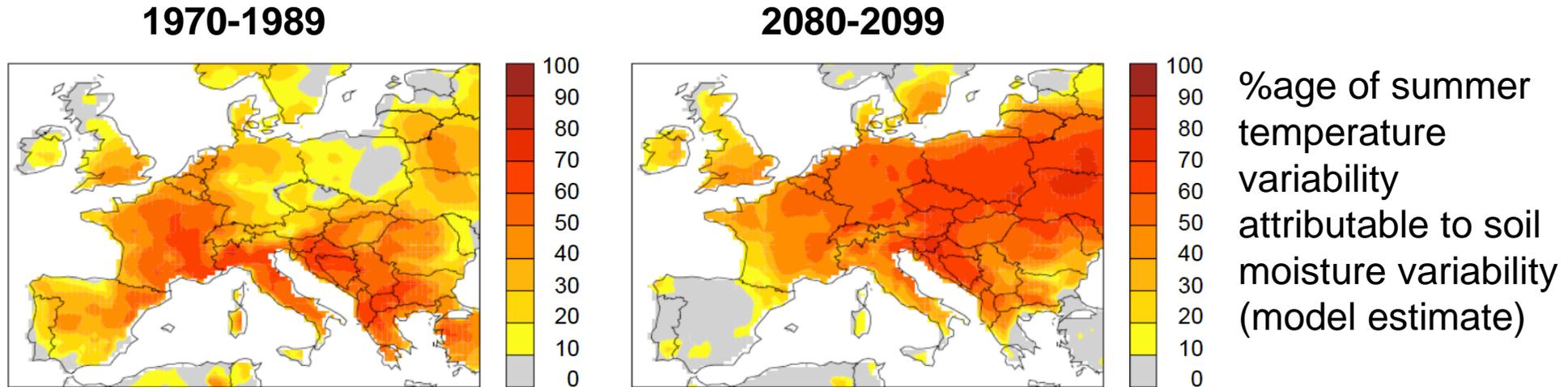
IPCC SREX (2012): <http://ipcc-wg2.gov/SREX/>



(IPCC 2012; Summary for Policymakers & Chapter 3)

Consistent projections of increased dryness in the Mediterranean region, central Europe, central North America, Central America and Mexico, northeast Brazil, and southern Africa

## Hot spots of soil moisture-climate coupling are expected to shift with changing climate (and changing soil moisture regimes)

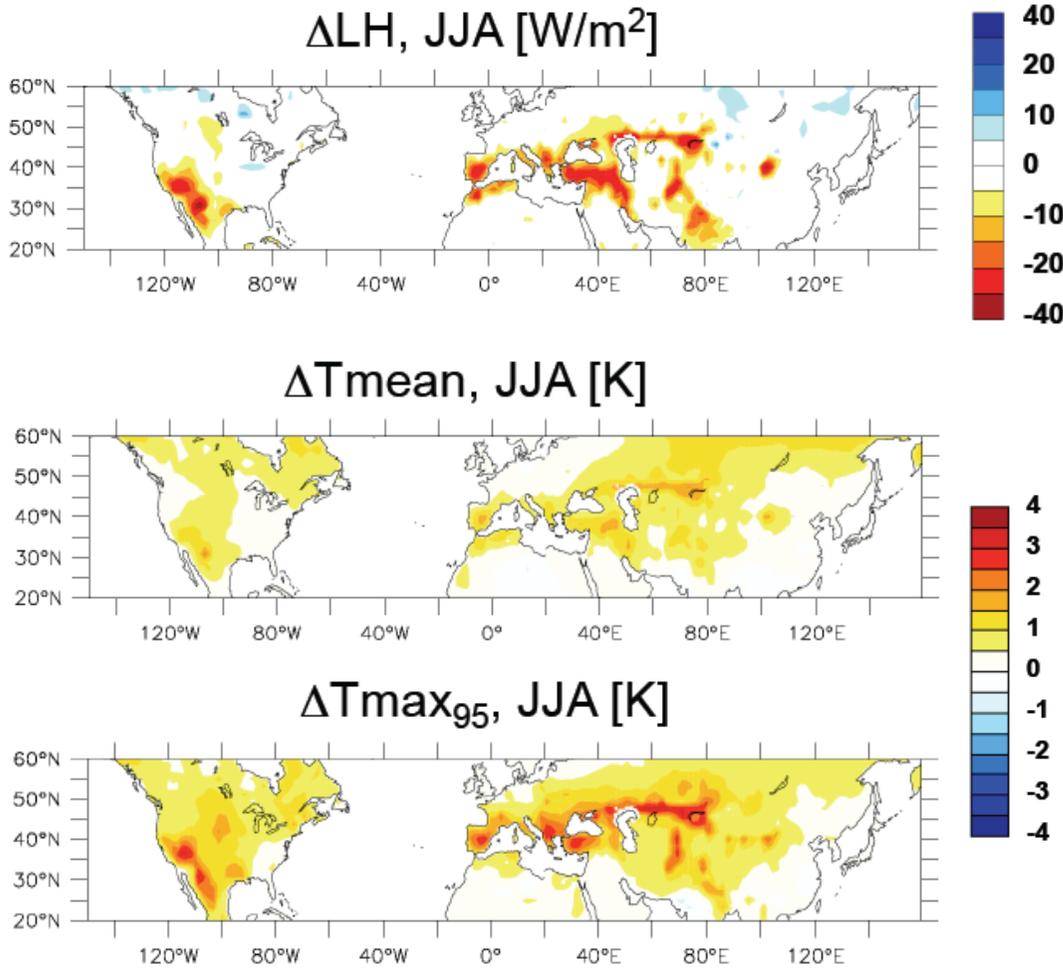


RCM simulations: Up to **60%** of *summer temperature variability* is induced by soil moisture feedbacks:

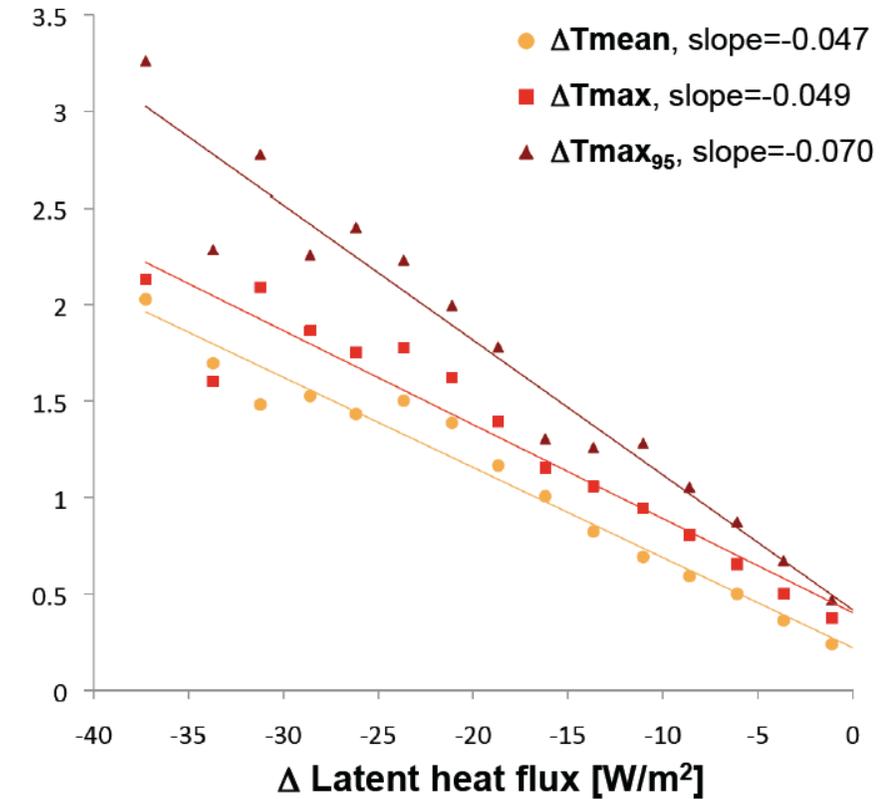
- In the Mediterranean area in late 20<sup>th</sup> century climate
- In Central and Eastern Europe in late 21<sup>st</sup> century climate

(Seneviratne et al. 2006, Nature)

## GLACE-CMIP5: Impact of changes in soil moisture regimes for projected climate (2071-2100 vs 1971-2000)



$\Delta$  Temperature [K], JJA



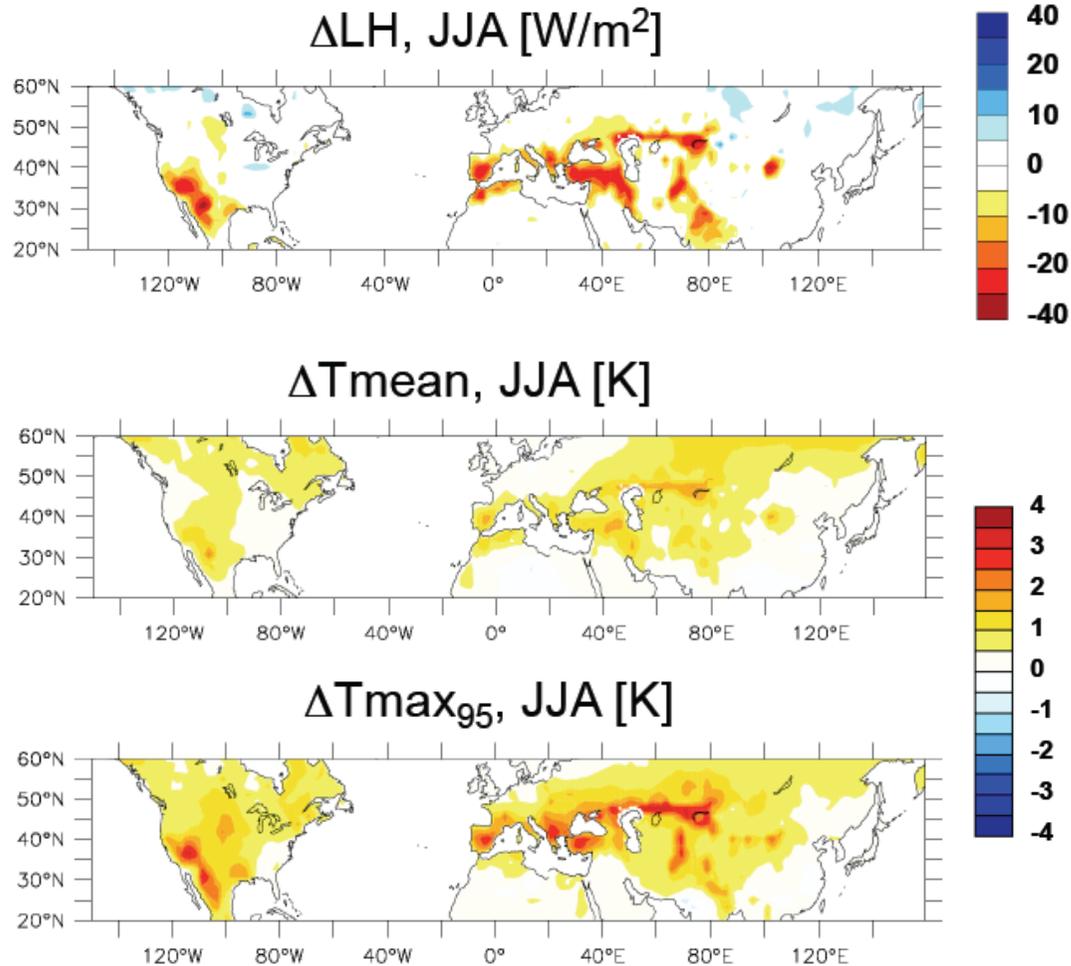
(Seneviratne et al. 2012, submitted)

- Basic relevant land-climate feedbacks and observational evidence
- Diagnosing prediction potential from land surface initialization
  - GLACE-2: Forecasting of atmospheric variables
  - Drought forecasting
- Discussion
  - Combined effects of large-scale circulation forcing and soil moisture feedbacks
  - Soil moisture initialization
  - Land-climate feedbacks in the context of climate change
- **Conclusions**

- High relevance of land-climate feedbacks for **climate variability over continents, in particular hot extremes**
- Soil moisture persistence implies **potential for improved monthly-subseasonal predictability** of temperature, precipitation, and agricultural and hydrological droughts: → not used at present!
- Key areas of research:
  - Respective role of large-circulation anomalies and soil moisture feedbacks for climate variability
  - Improved observational estimates and quantification of feedbacks based on observations
  - Feedbacks in the context of climate change (also for attribution)



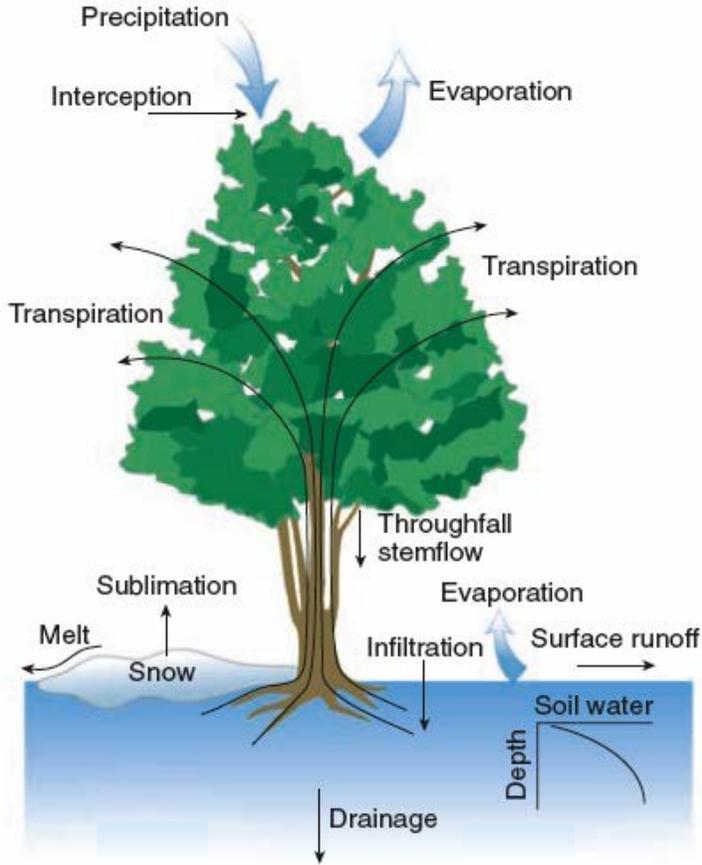
## GLACE-CMIP5: Impact of changes in soil moisture regimes for projected climate (2071-2100 vs 1971-2000)



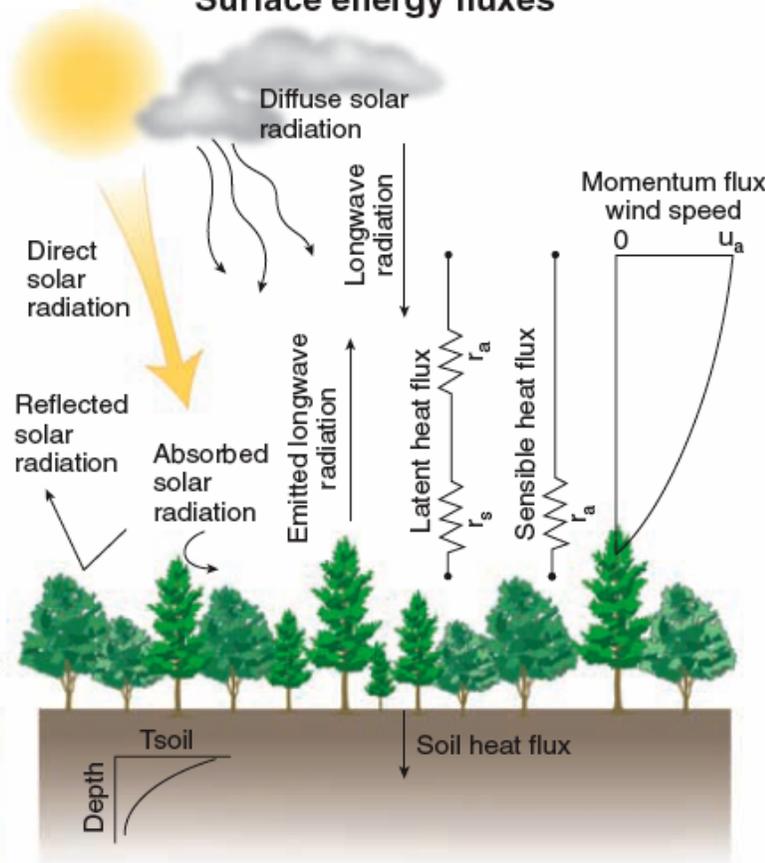
(Seneviratne et al. 2012, submitted)

Group/Model	# models	Points of Contact
1. NASA/GSFC (USA): GMAO seasonal forecast system (old and new)	2	R. Koster, S. Mahanama
2. COLA (USA): COLA GCM, NCAR/CAM GCM	2	P. Dirmeyer, Z. Guo
3. Princeton (USA): NCEP GCM	1	E. Wood, L. Luo
4. ETH Zurich (Switzerland): ECHAM GCM	1	S. Seneviratne, T. Stanelle
5. KNMI (Netherlands): ECMWF	1	B. van den Hurk
6. ECMWF	1	G. Balsamo, F. Doblas-Reyes
7. GFDL (USA): GFDL system	1	T. Gordon
8. U. Gothenburg (Sweden): NCAR	1	J.-H. Jeong
9. CCSR/NIES/FRCGC (Japan): CCSR GCM	1	T. Yamada
10. FSU/COAPS	1	M. Boisserie
11. CCCma	1	B. Merryfield
	<hr/> 13 models	

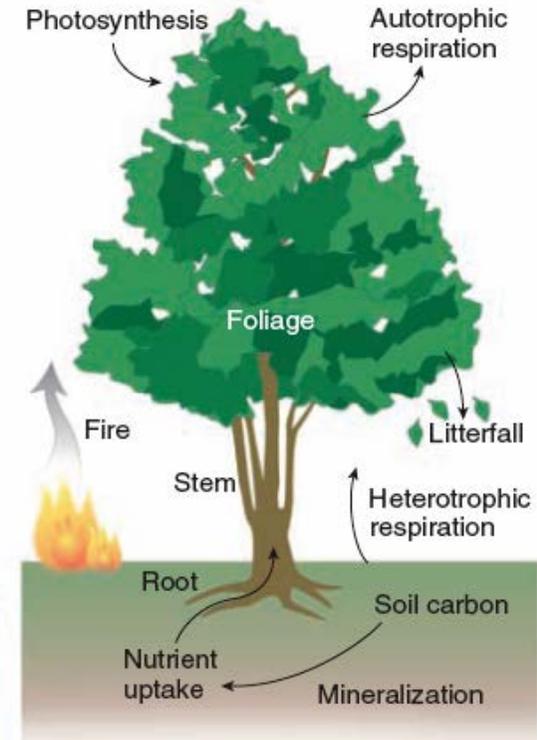
## Hydrology



## Surface energy fluxes



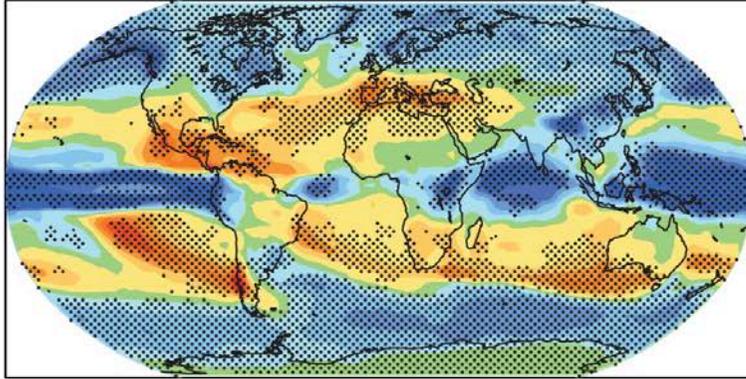
## Carbon Cycle



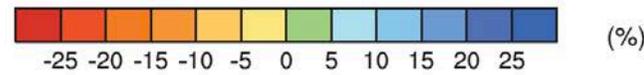
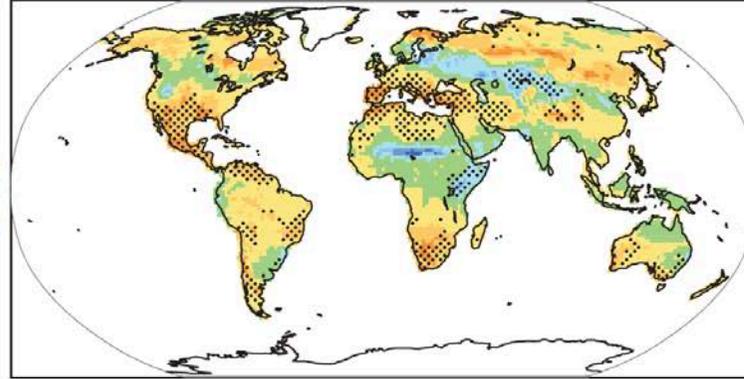
(Bonan, Science 2008)

## Changes in hydrological cycle

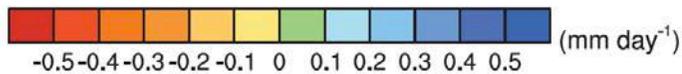
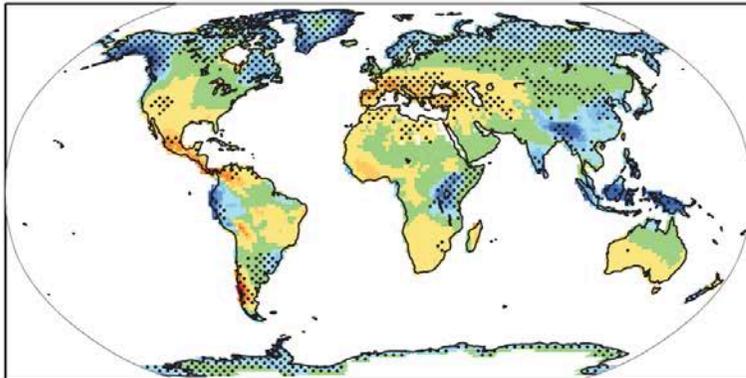
a) Precipitation



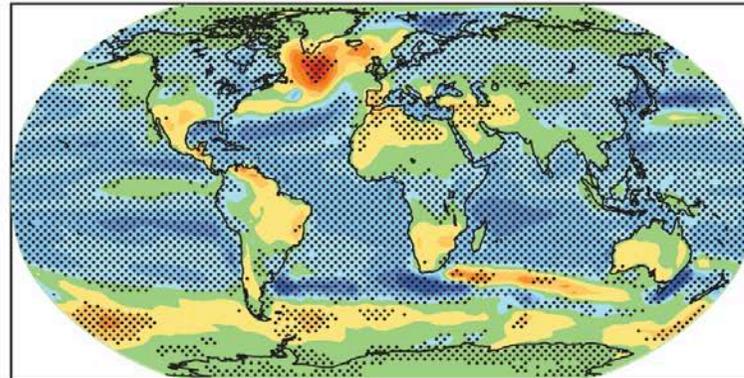
b) Soil moisture



c) Runoff



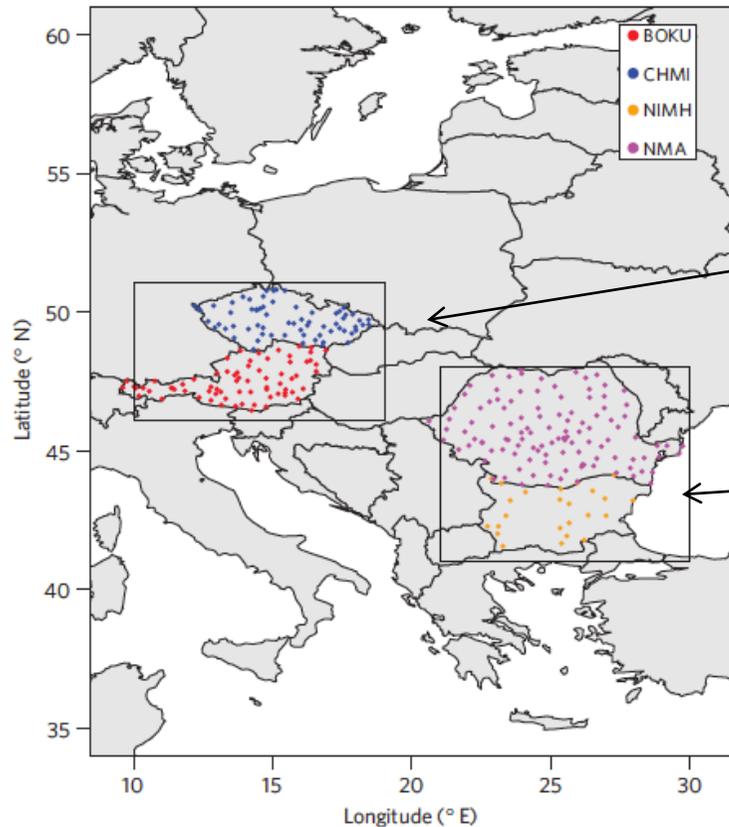
d) Evaporation



(IPCC 2007)

## Observational evidence for soil-moisture impact on hot extremes in southeastern Europe

Martin Hirschi<sup>1,2\*</sup>, Sonia I. Seneviratne<sup>1\*</sup>, Vesselin Alexandrov<sup>3</sup>, Fredrik Boberg<sup>4</sup>,  
Constanta Boroneant<sup>5</sup>, Ole B. Christensen<sup>4</sup>, Herbert Formayer<sup>6</sup>, Boris Orlowsky<sup>1</sup> and Petr Stepanek<sup>7</sup>

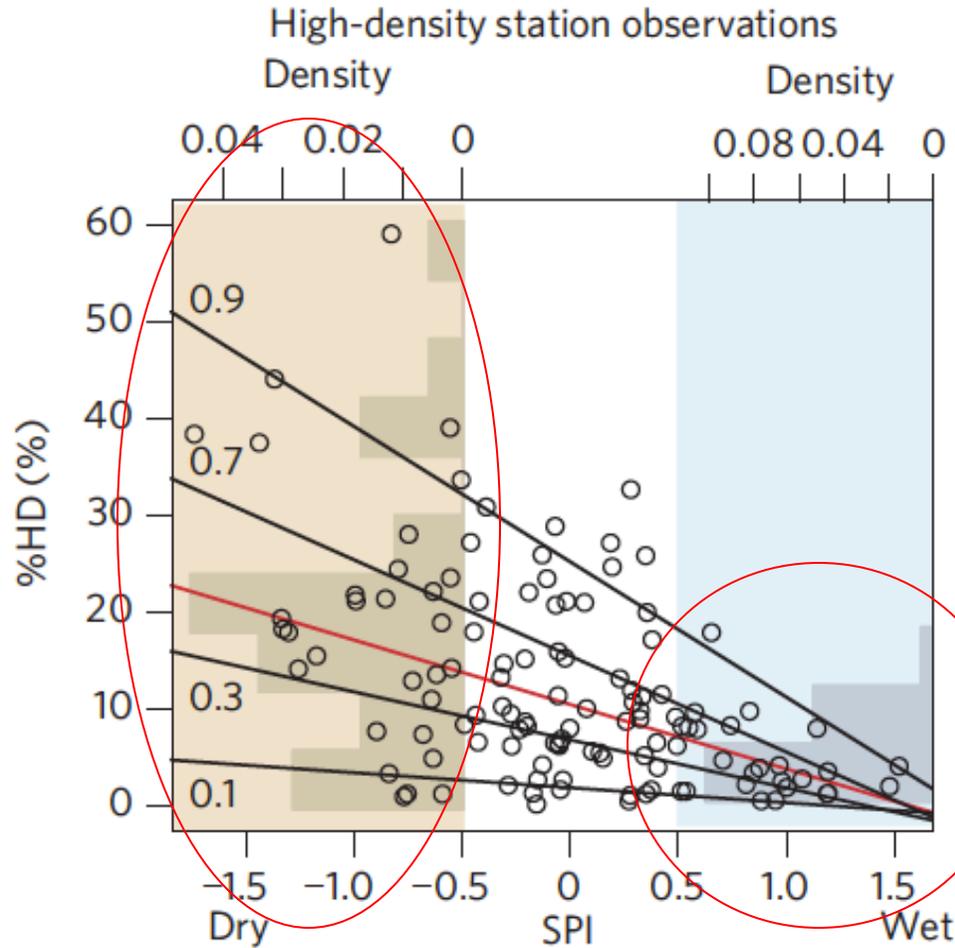


Radiation-limited  
evapotranspiration regime

Soil moisture-limited  
evapotranspiration regime

(Hirschi et al. 2011, Nature Geoscience)

**Possibly more skill for hot extremes after wet vs dry conditions (dry soil necessary but not sufficient condition)**

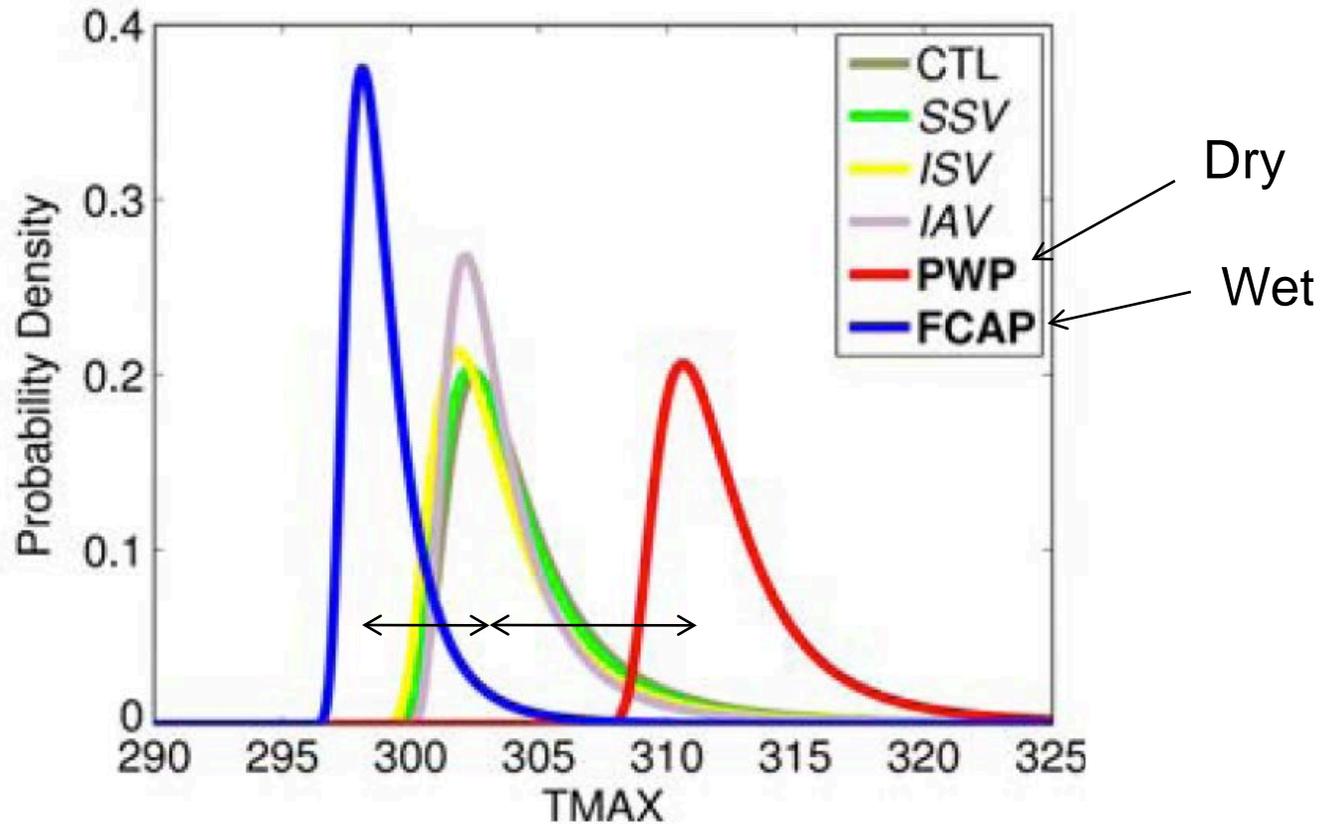


Regression lines: — 0.1, 0.3, 0.7, 0.9 %HD quantiles

(Hirschi et al. 2011, Nature Geoscience)

## Distribution of summer Tmax block maxima

RCM simulation with COSMO/CCLM (France, 1959-2006)



(Jaeger and Seneviratne,  
*Climate Dynamics*, 2011)

1. The individual models vary in their ability to extract forecast skill from land initialization (not shown). In general,
  - Low skill for precipitation
  - Moderate skill (in places) for temperature, even out to two months.
2. Land initialization impacts on skill increase dramatically when conditioned on the size of the initial local soil moisture anomaly.

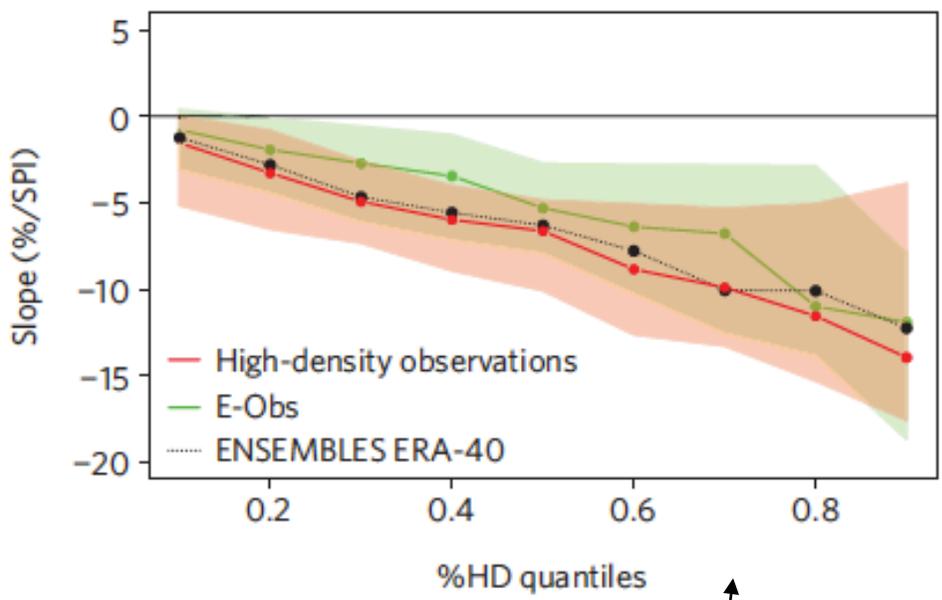


If you know the local soil moisture anomaly at time 0 is large, you can expect (in places) that initializing the land correctly will improve your temperature forecast significantly, and your precipitation forecast slightly, even out to 2 months.

2. The results highlight the potential usefulness of improved observational networks for prediction.

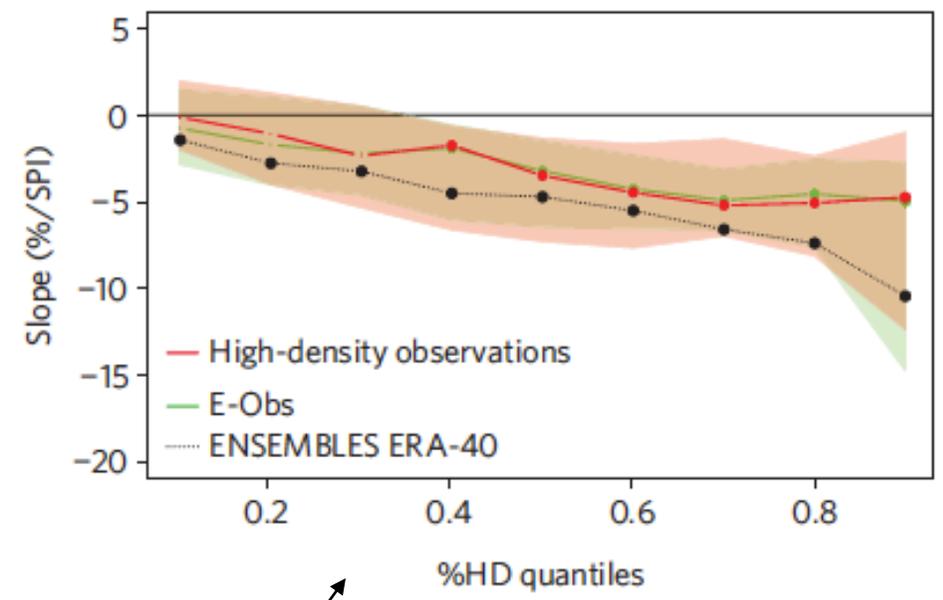
## Southeastern Europe

%HD-SPI



## Central Europe

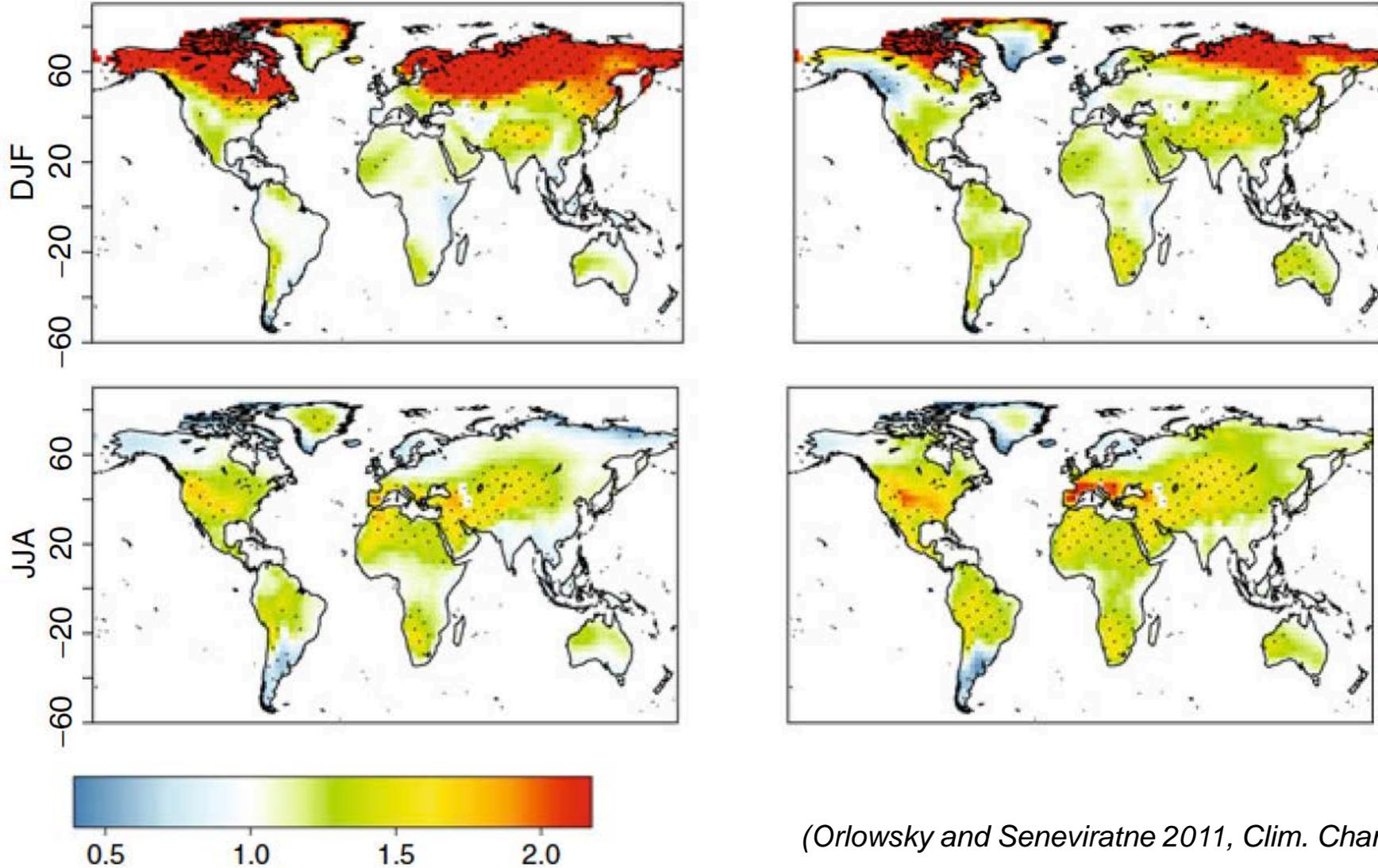
%HD-SPI



Quantile regressions: RCMs from ENSEMBLES perform fairly well but display an overestimation of feedback strength in C. Europe

(Hirschi et al. 2011, Nature Geoscience)

## Scaling $\Delta T_{\text{max,local,seas}} / \Delta T_{\text{max,global}}$ for 10% (left) and 90%ile (right)



(Orlowsky and Seneviratne 2011, *Clim. Change*, publ. online)