


CLM-related model evaluations and improvements at the University of Arizona

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CLM4 improvement

- Skin temperature diurnal cycle over arid regions (Zeng et al.)

Land-atmosphere interaction

- Land-precipitation coupling strength (Zeng et al.)
- Impact of interannual climatic variabilities on vegetation (Shao et al.)

CLM evaluations

- Monthly river flow prediction (or simulation) (Zeng et al.)
- CMIP5 carbon cycle (Shao et al.)
- Steady state of Fractional Cover/carbon (Sakaguchi et al.)
- Dynamic root function (Christoffersen et al.)
- PFT distribution across Amazon (Moreno et al.)

Ongoing work (not discussed here)

- Global 1 km hybrid 3-D hydrological modeling
- B2 Landscape Earth Observatory (LEO)



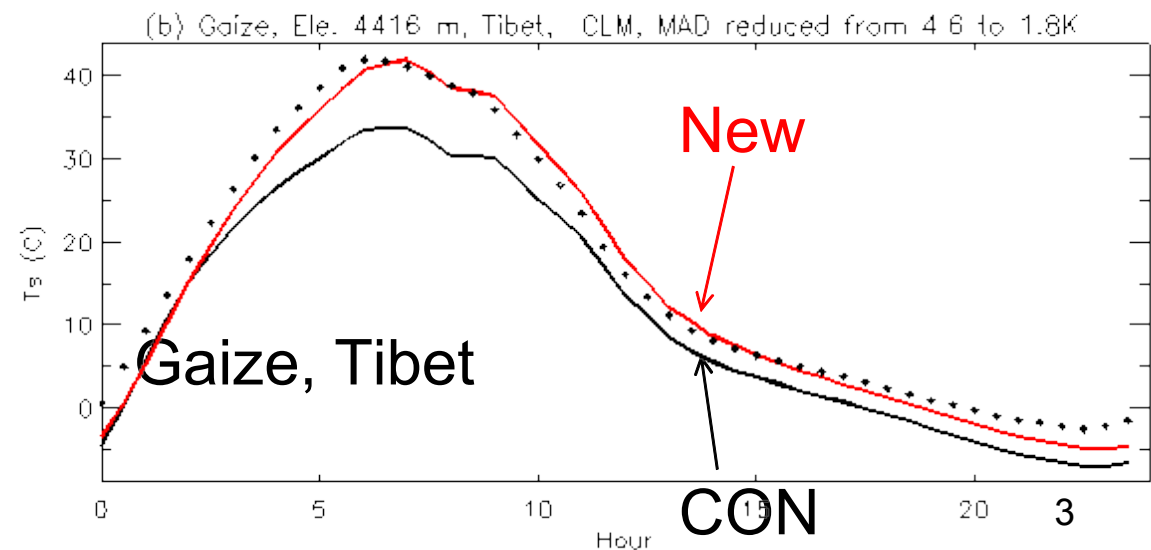
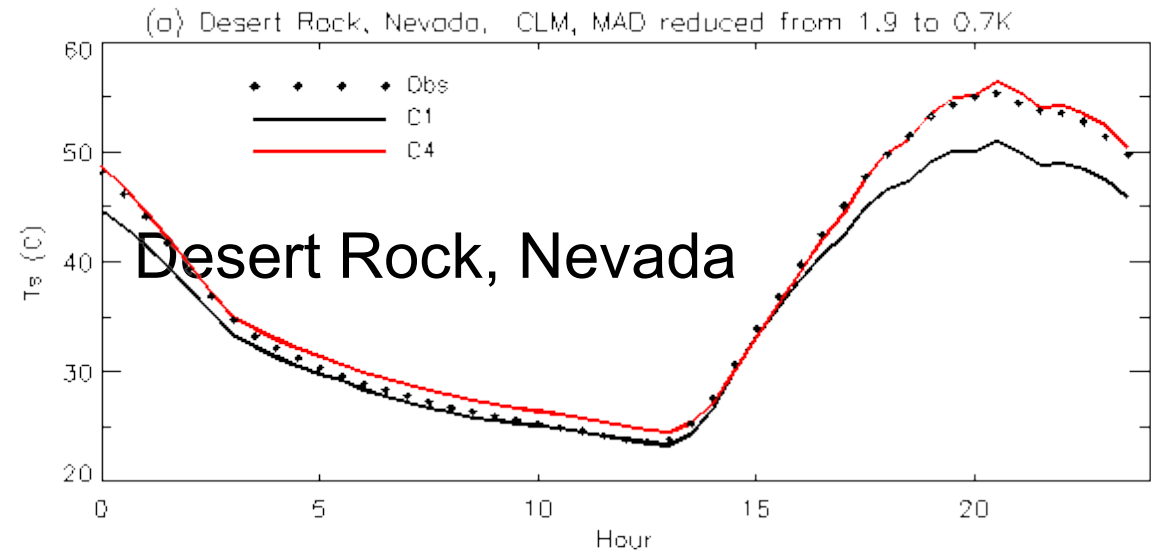
Improve the Skin Temperature Diurnal Cycle over Arid Regions (Zeng et al. 2012; in press)

$$\ln(z_{om}/z_{ot}) = 0.36 (u_* z_{om}/v)^{0.5}$$

$$u_{*min} = 0.07 \rho_o/\rho (z_{om}/z_{og})^{0.18}$$

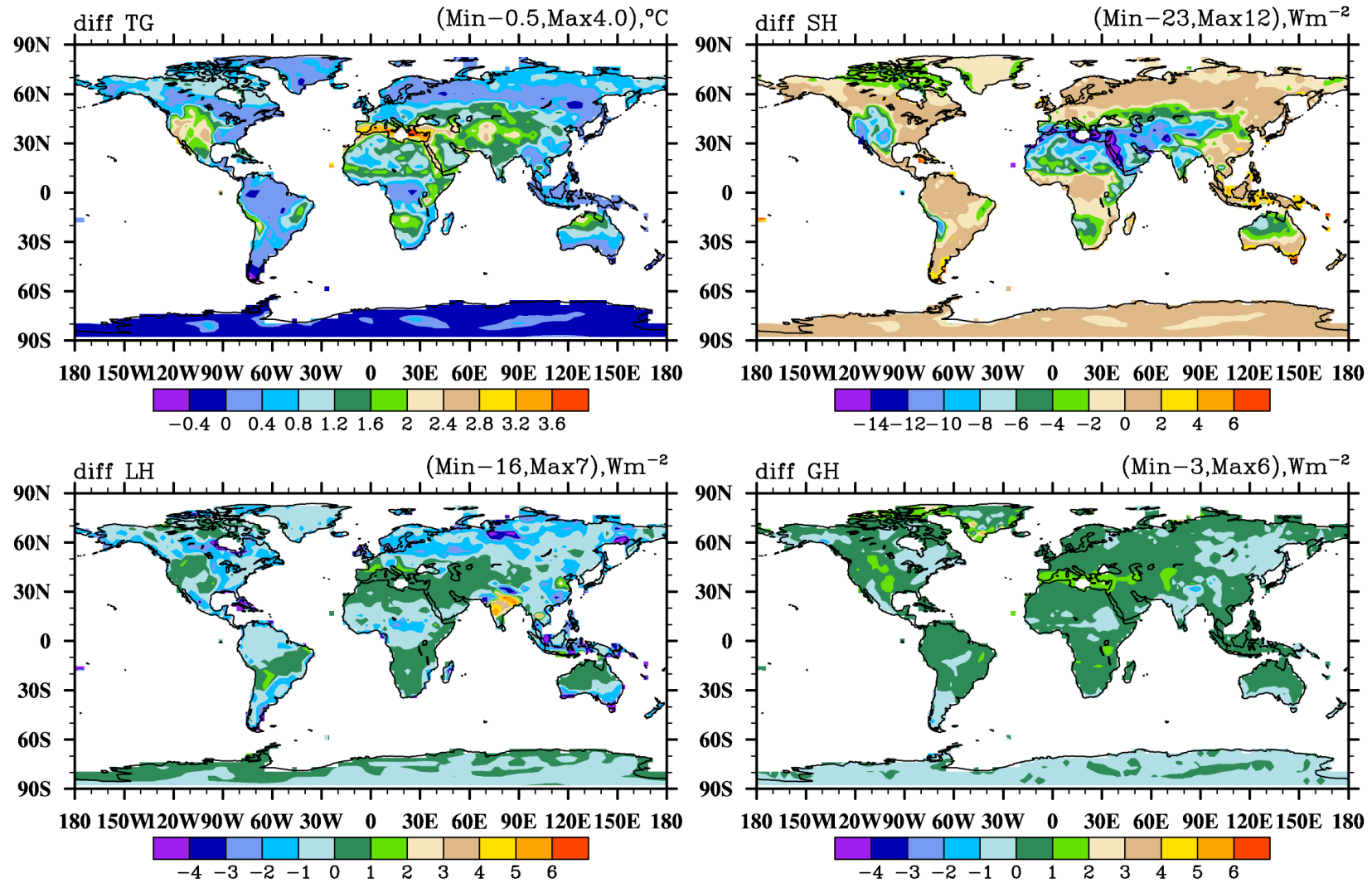
$$\text{minimum } K_{soil} = 0.75 \text{ Wm}^{-1}\text{K}^{-1}$$

	Mean absolute deviation (K)	
	Desert Rock	Gaize
Con	1.9	4.6
New	0.7	1.8



New - Control

Global T42 July

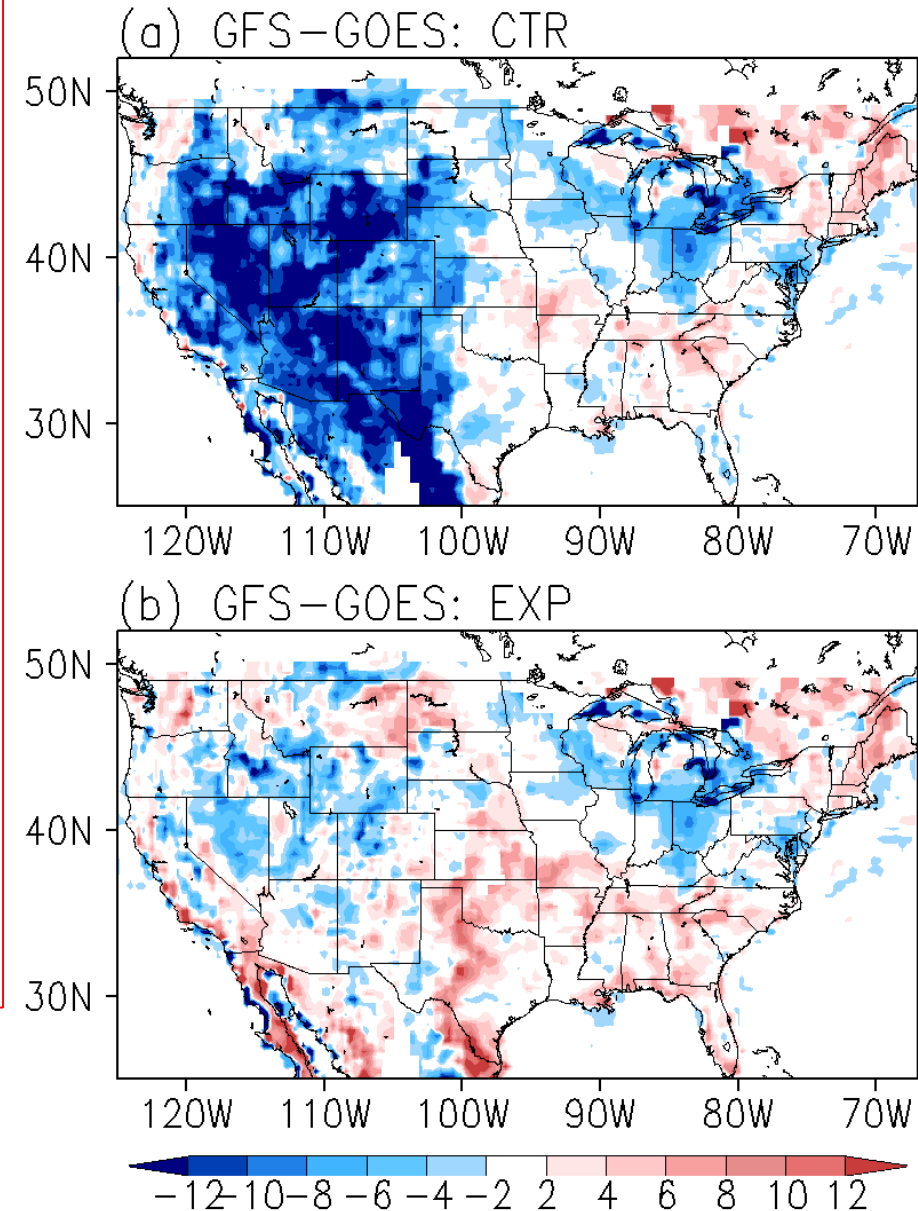


The same formulation for roughness length has been Implemented and tested in NCEP GFS operational Model (Zheng et al. 2012; in press)

Significant improvement over semi-arid regions

Significant increase in the number of surface-sensitive satellite brightness temperature data assimilated (not shown)

Skin temperature difference at 1800 UTC



Land-atmosphere interaction

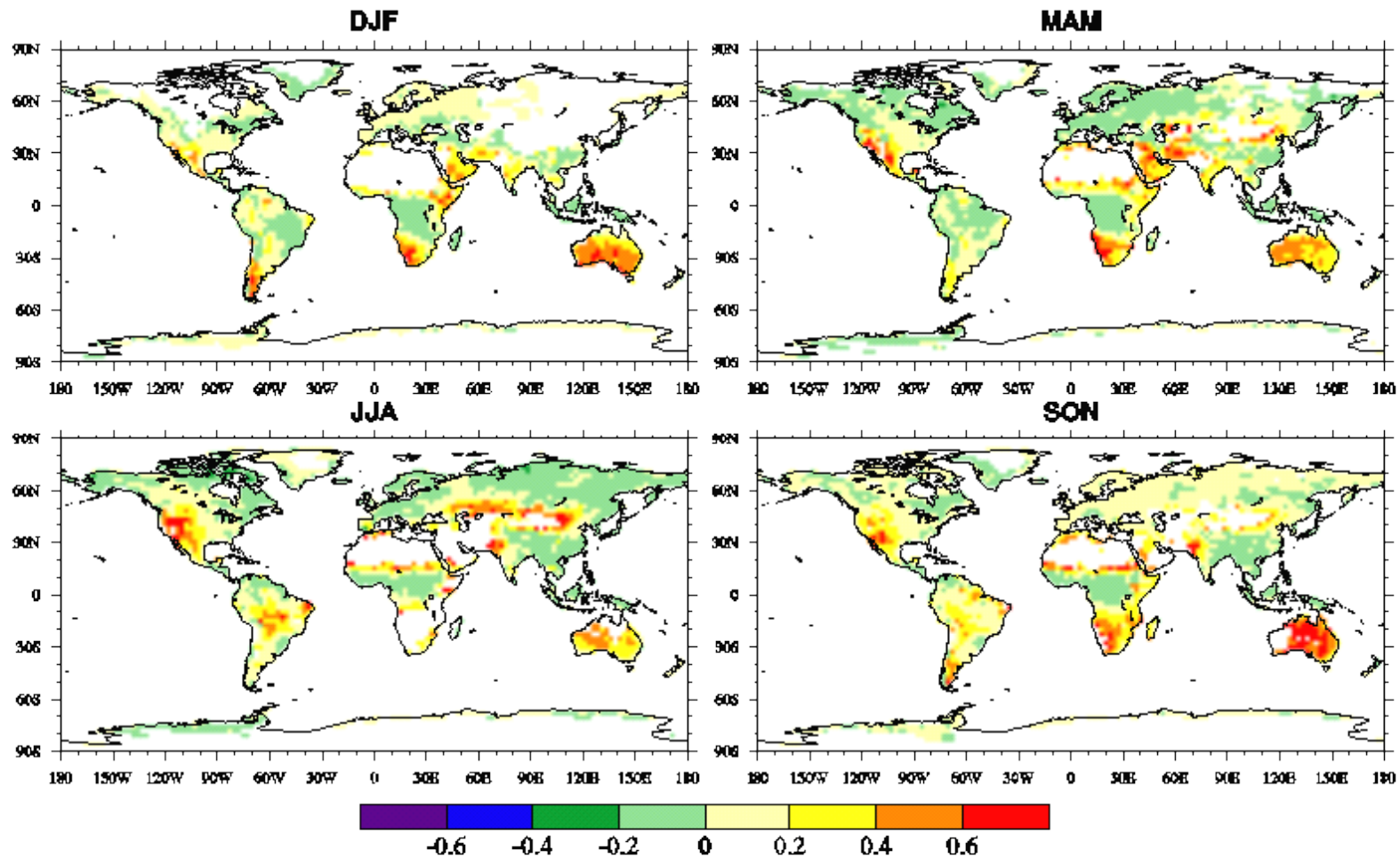
- *Land-precipitation coupling strength (Zeng et al.)*
- Influence of interannual climatic variabilities on vegetation(Shao et al.)

Land-Precipitation Coupling Strength (Zeng et al. 2010)

$$\Gamma = \Sigma P'E' / \Sigma P'P'$$

E' , P' are monthly deviations from climatology

ECMWF 45yr Reanalysis

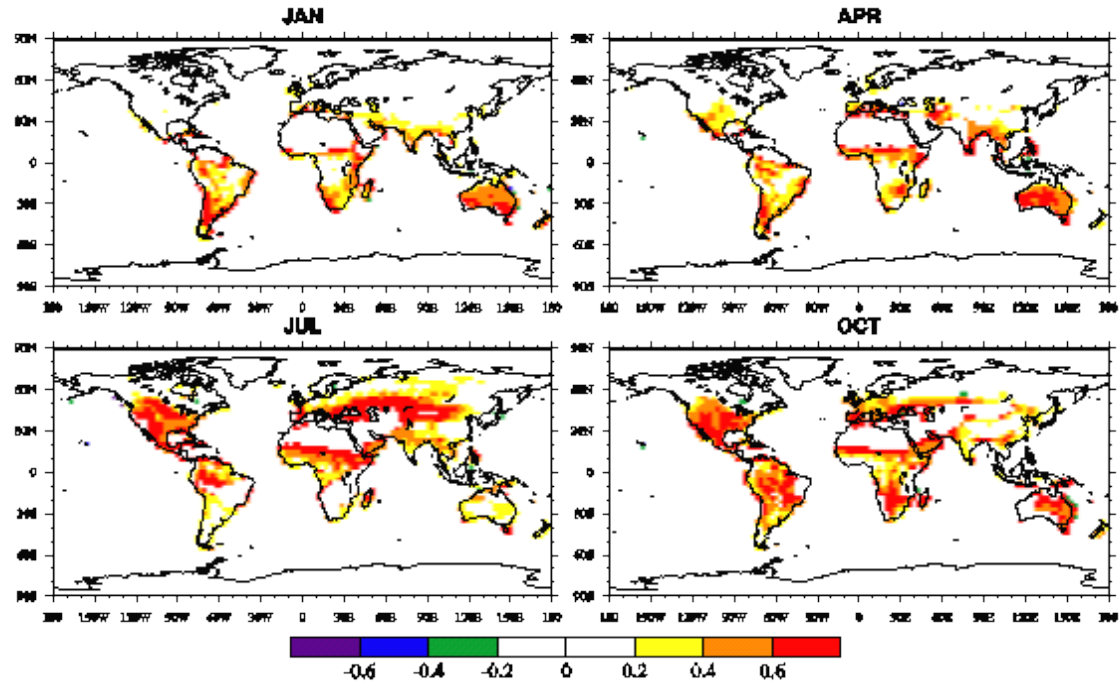


Γ provides a simple indicator to characterize a GCM's coupling strength

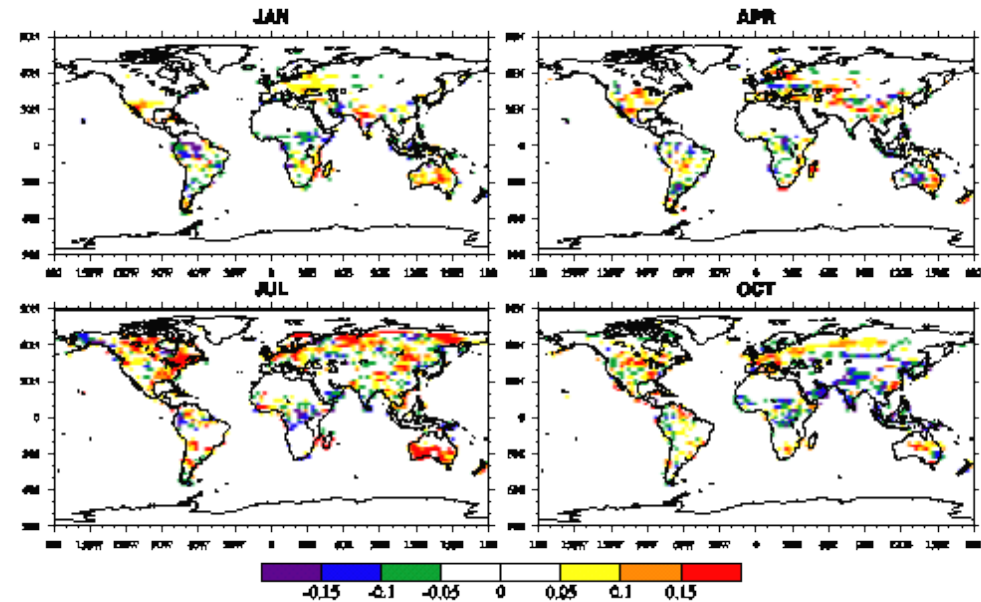
CCSM3 coupling is too strong

2*CO₂ increases the coupling strength over high latitudes in summer

CCSM3 50yr Control Run



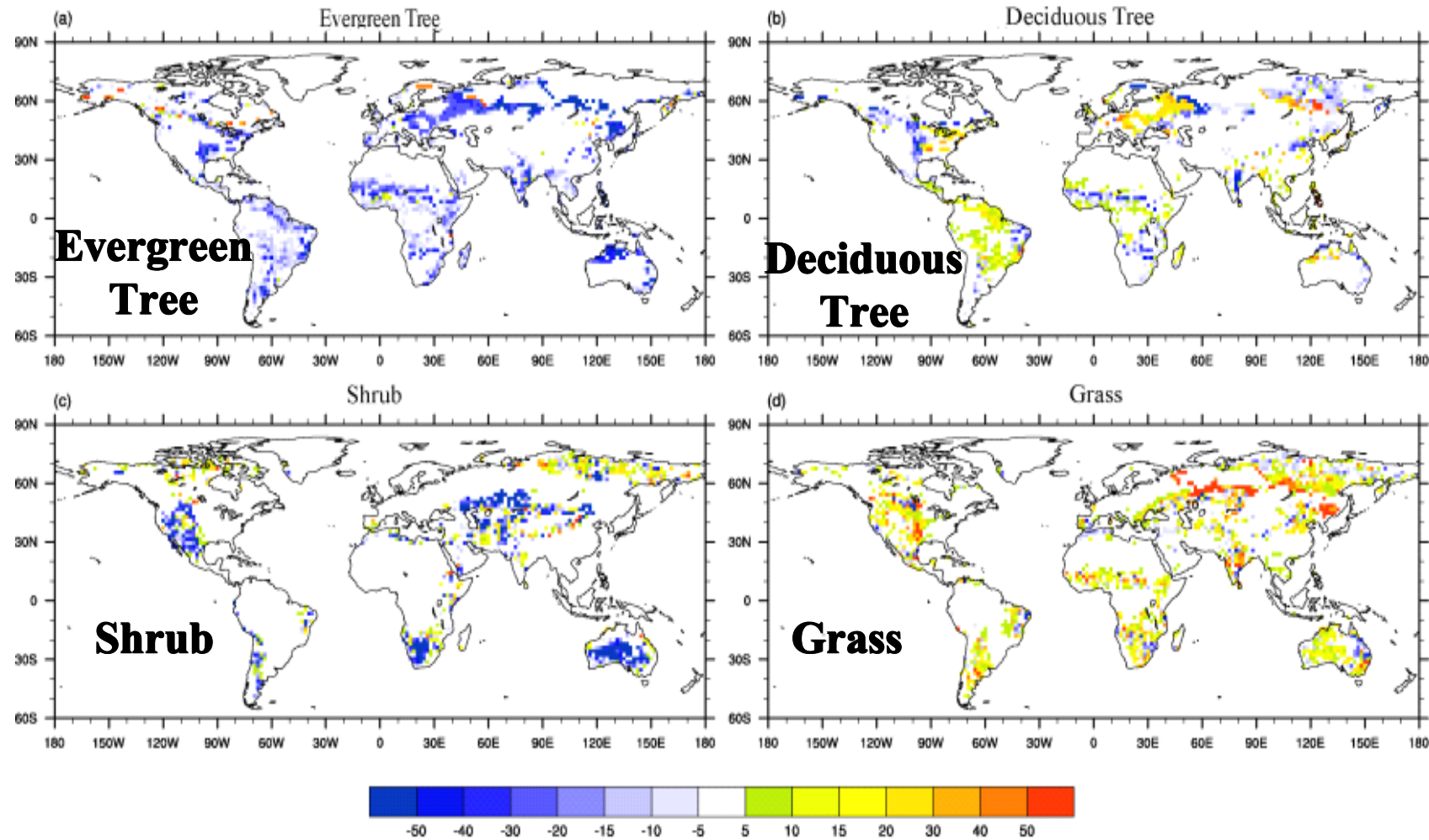
CCSM3 50yr 2X CO₂ - Control



Land-atmosphere interaction

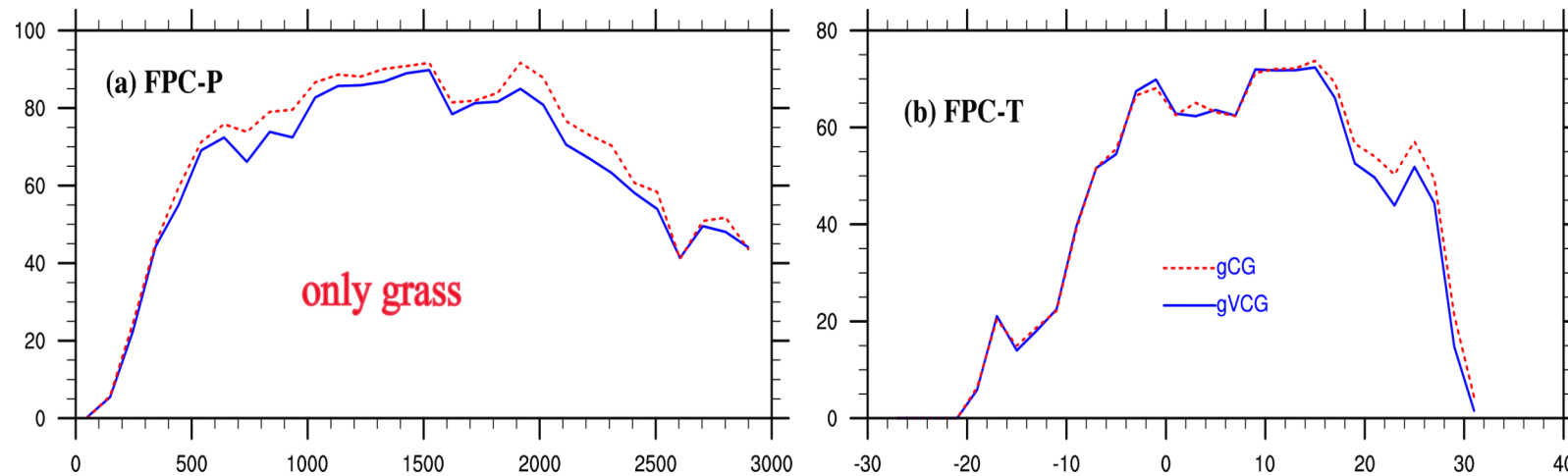
- Land-precipitation coupling strength (Zeng et al.)
- ***Impact of interannual climatic variabilities on vegetation***
(Shao et al.)
 - CLM/DGVM forced by observations from 1950-1999 versus from climatology.

Impact of Climatic Interannual Variabilities on Vegetation(Shao et al. 2011)



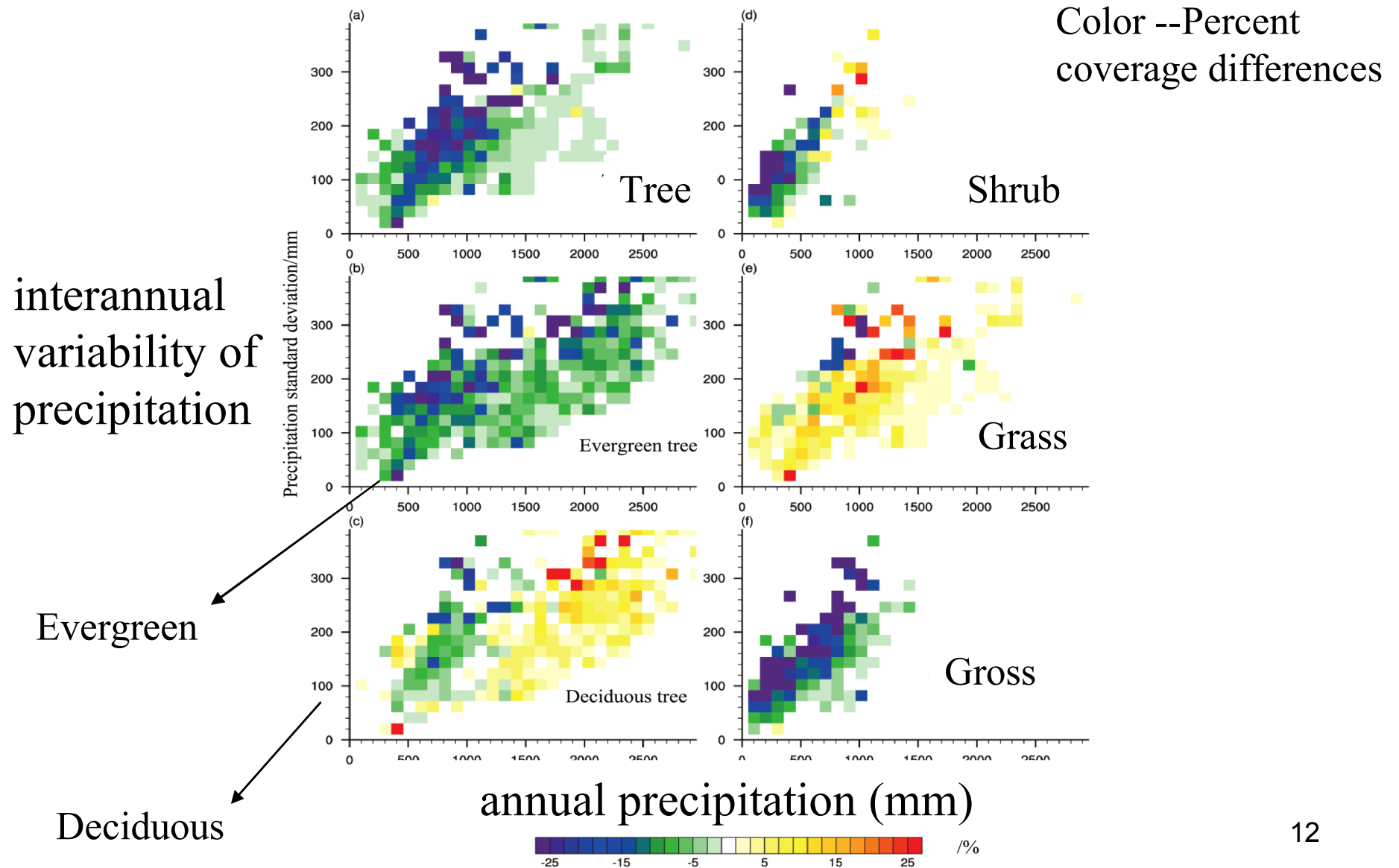
Is grass so sensitive to the climatic variability?

fractional cover distribution along the P&T



the expansion of grass is mainly due to the reduction of tree and shrub

Percent coverage differences in relation to mean and standard deviation of climatic factors

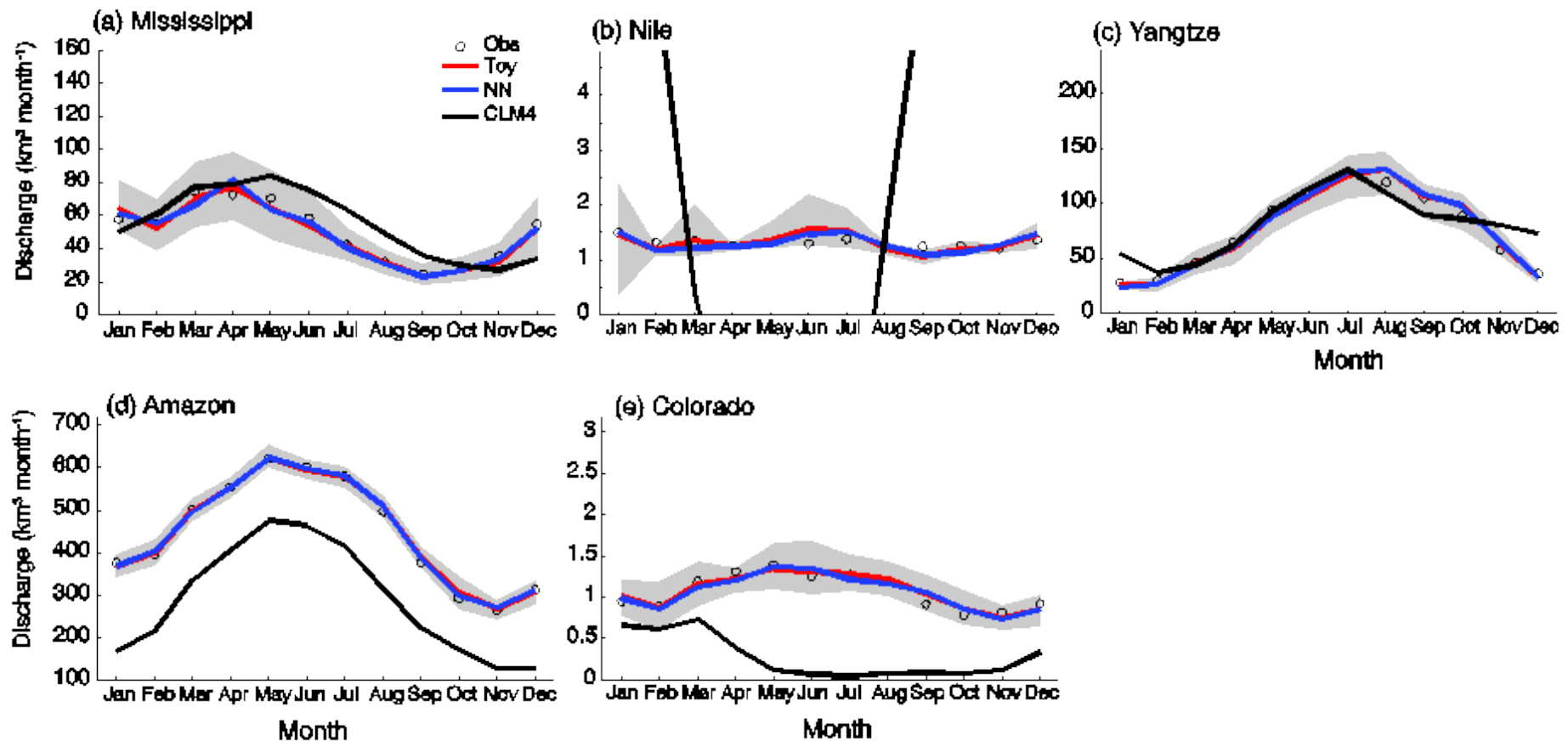


CLM evaluations

- *Monthly river flow prediction (or simulation) (Zeng et al.)*
- CMIP5 carbon cycle (Shao et al.)
- Steady state of Fractional Cover/carbon (Sakaguchi et al.)
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A water-balance based “toy” model (Zeng et al. 2012; in revision)

as good as a neural network for monthly river flow prediction, but the toy model is more robust. **They are both much better than CLM4 simulation.**



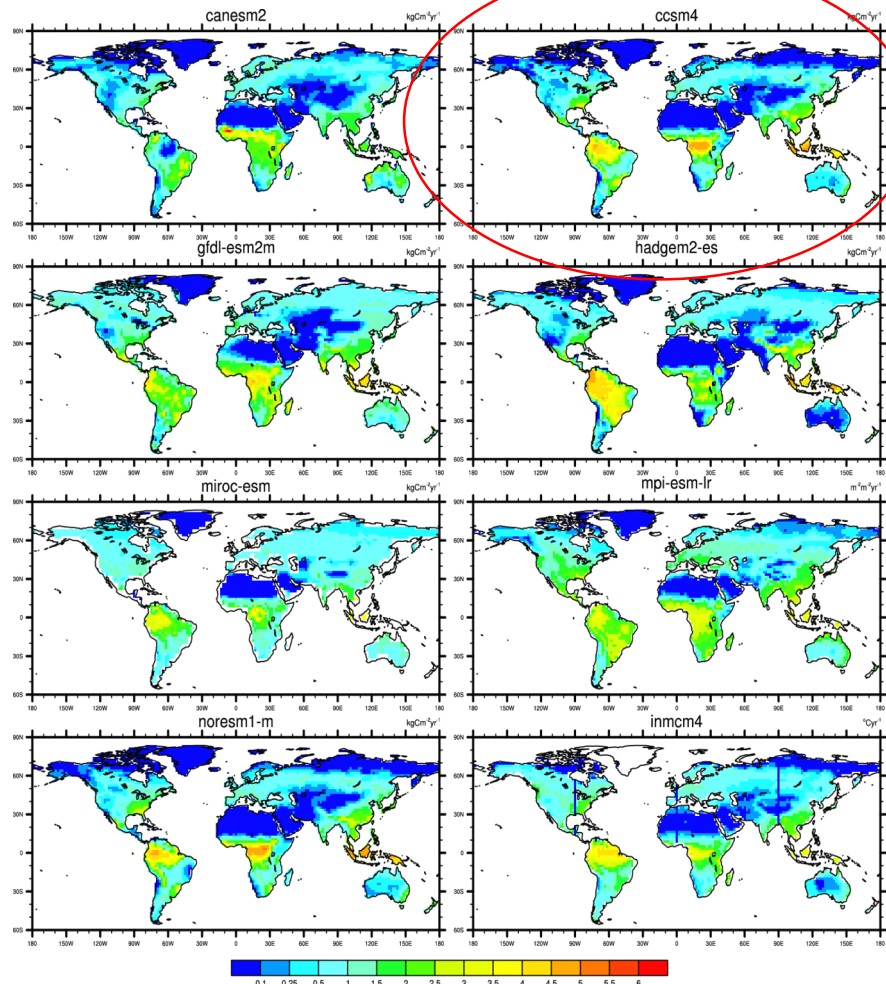
Modified coefficient of efficiency: **0.55-0.93 for Toy model**
<0.1 for CLM4

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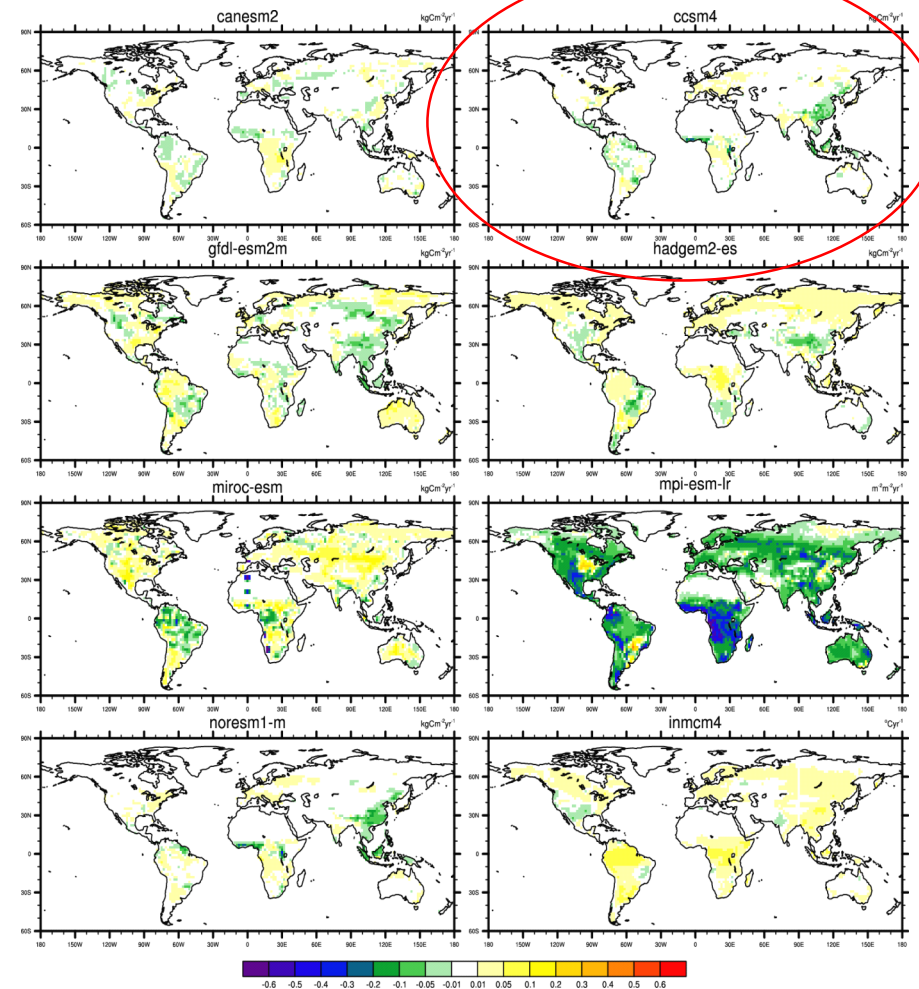
CMIP5: GPP /NBP in historical and RCP4.5 exp (Shao et al.)

GPP (Gross Primary Productivity)



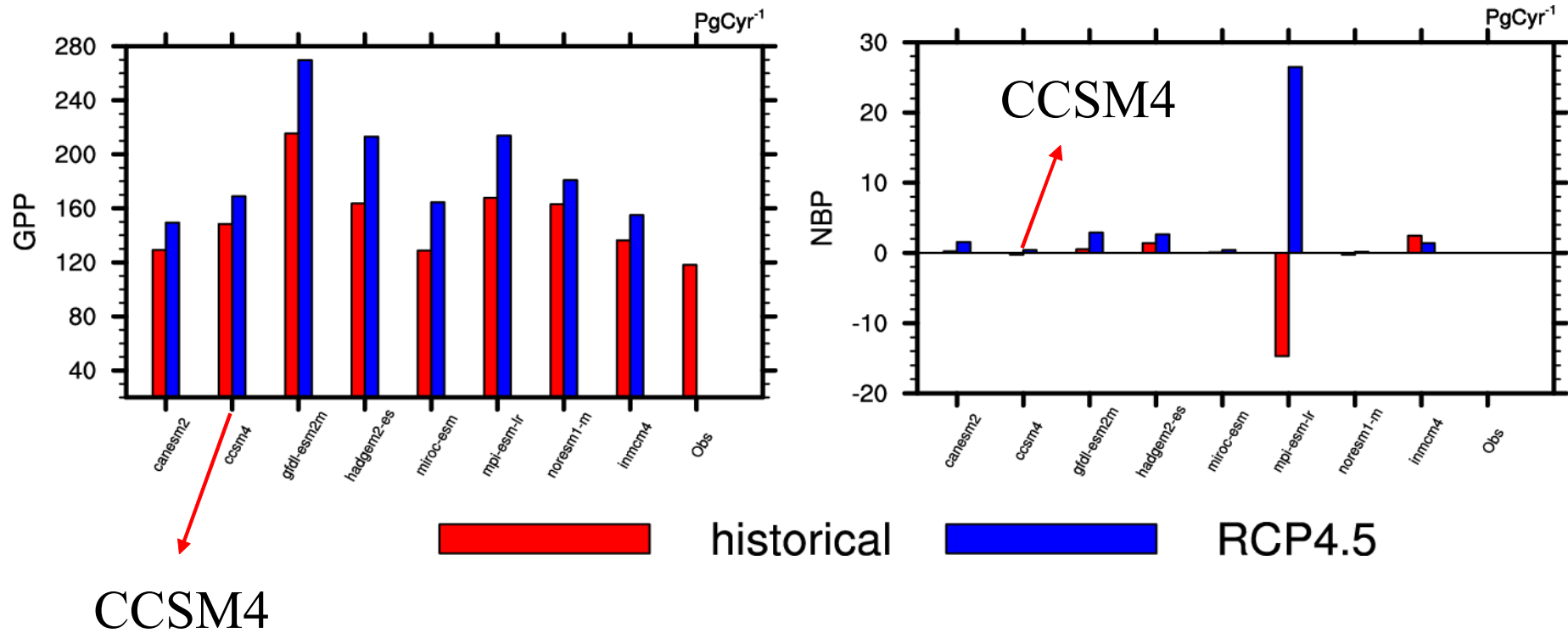
general pattern: similar
magnitude: very different

NBP (Net Biosphere Productivity)



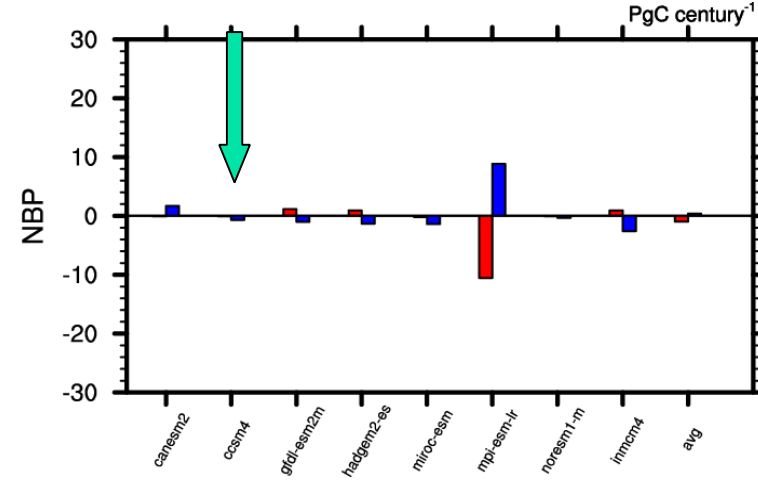
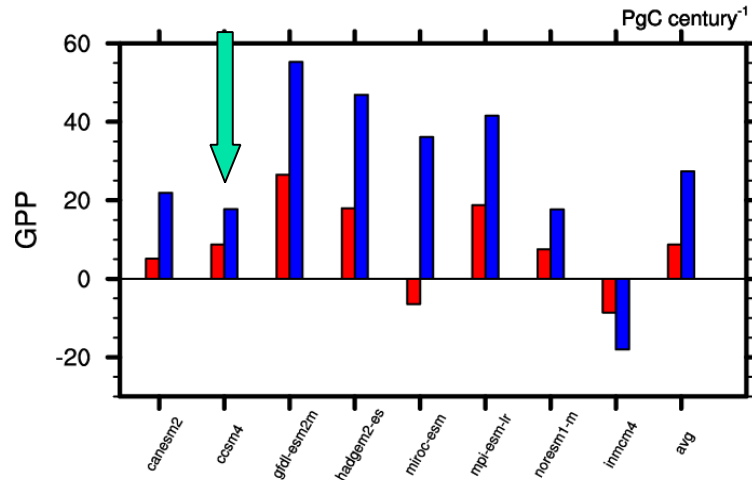
discrepancies exist

last 30 year average of global GPP and NBP



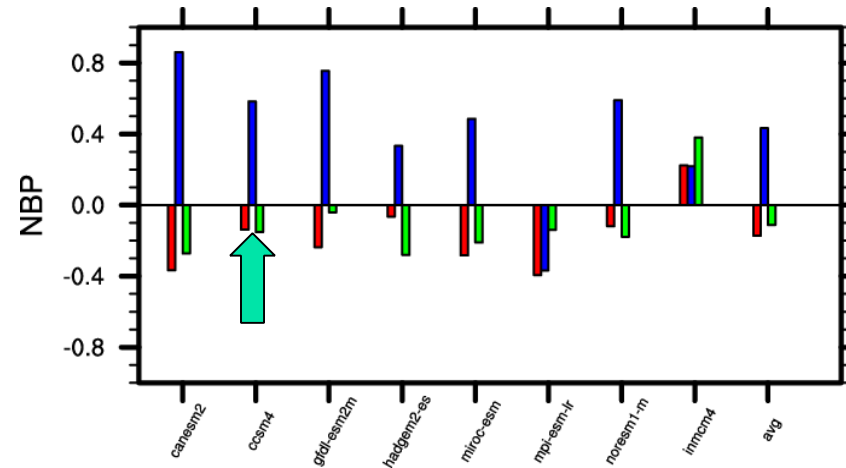
GPP : increased in every model; NBP: close to 0 for balance

Trend



historical RCP4.5

correlations between global historical NBP and climatic variables



T SM P

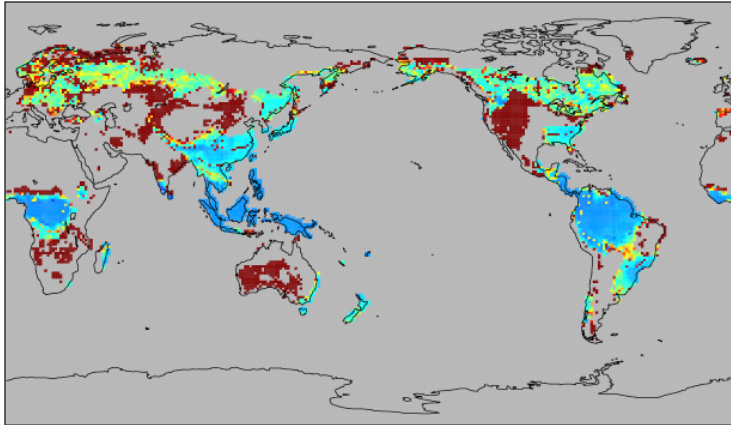
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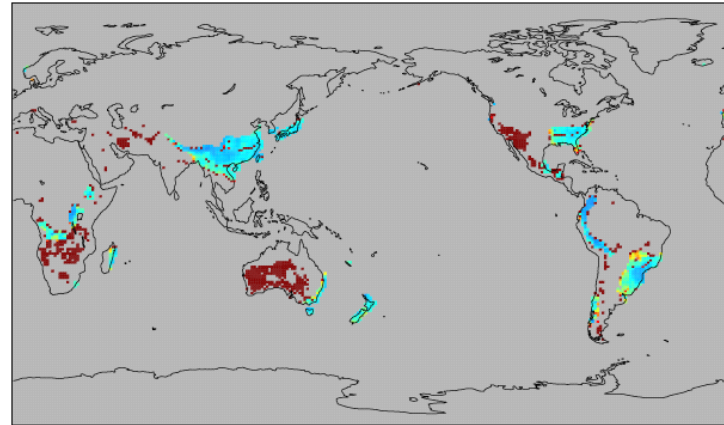
Years to Reach the Steady State of Fractional Cover (Sakaguchi et al.)

Evergreen Tree PFTs

Average of the three evergreen PFTs

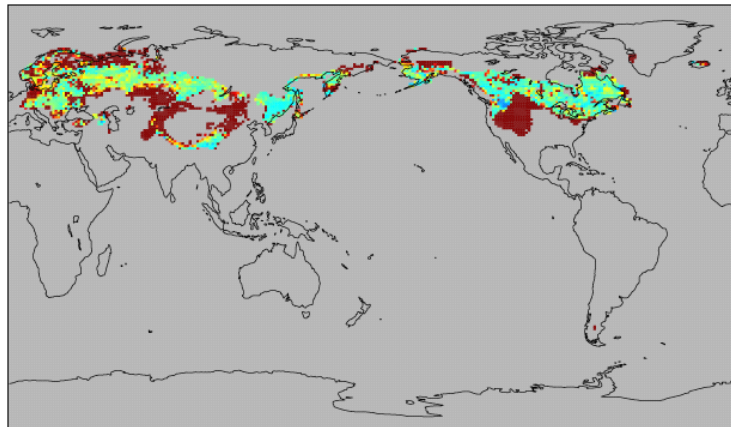


NET Temperate

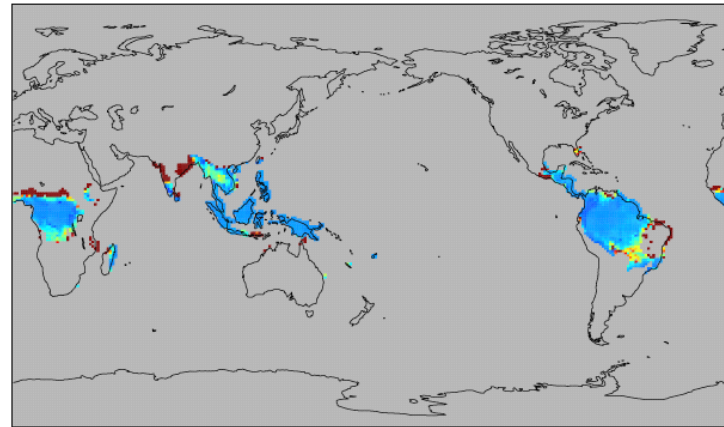


FPC < 5% is not included

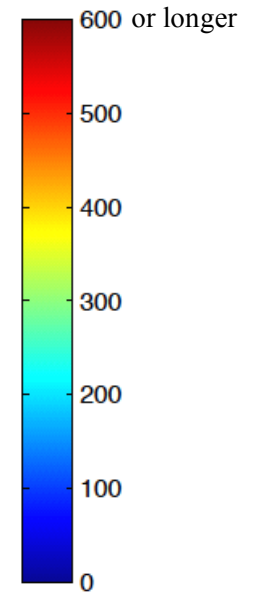
NET Boreal



BET Tropical



(year)



- Longer years for dry regions and for NET Boreal

Steady State of Fractional Cover

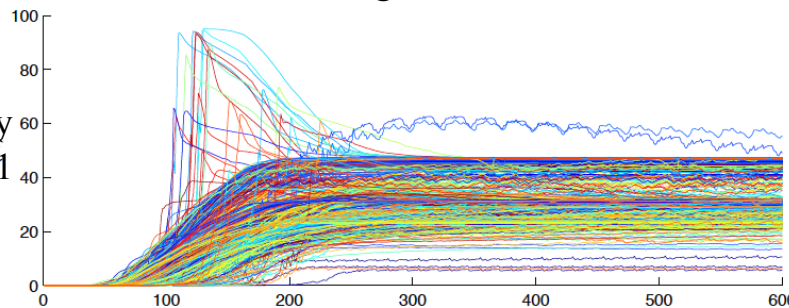
% of global grid boxes to reach steady state for tree PFTs.

	1 - 100	101 - 200	201 - 300	301 - 400	401 - 500	501 - 599	> 600 or unstable
NET temp	0	17	37	5	2	3	37
NET boreal	0	1	36	16	5	5	37
BET tropical	0	69	12	5	1	1	11
BDT tropical	0	77	14	5	2	1	2
BDT temp	1	30	36	8	5	5	15
BDT boreal	0	9	30	19	6	5	30

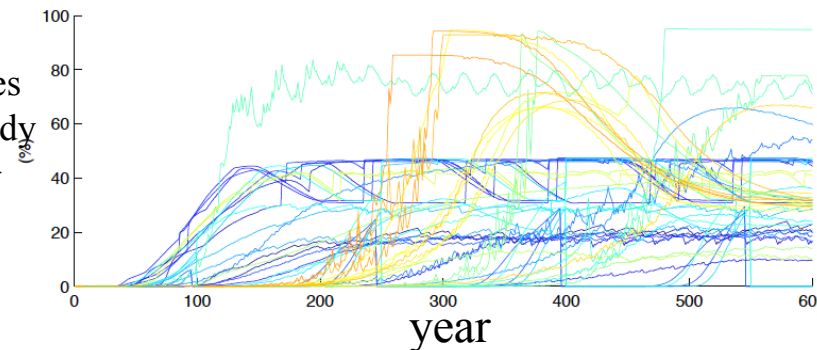
Example: NET temperate

Fractional Vegetation Cover

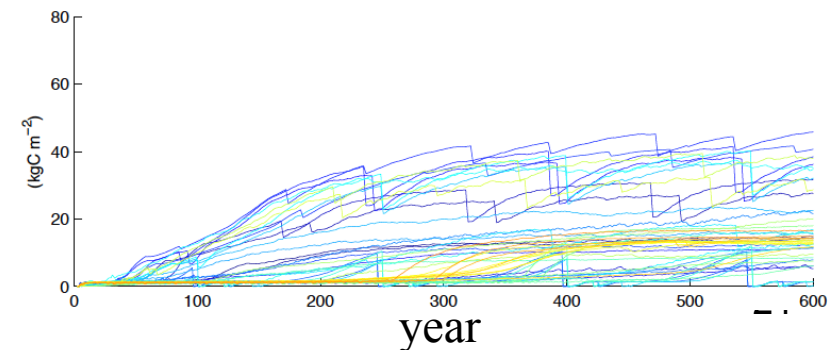
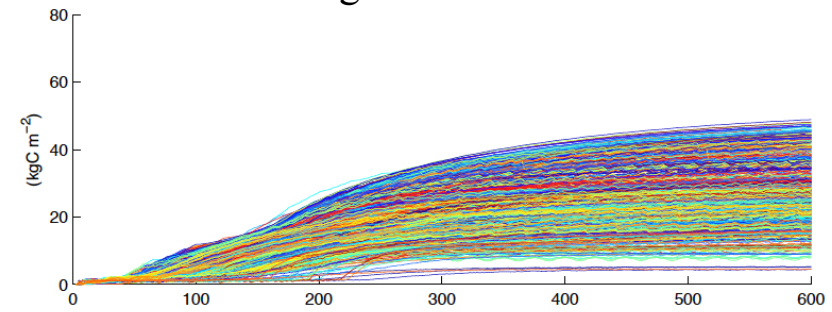
grid boxes with steady state in 201 ~ 300 yrs



grid boxes with steady state in > 500 yrs



Total Vegetation Carbon



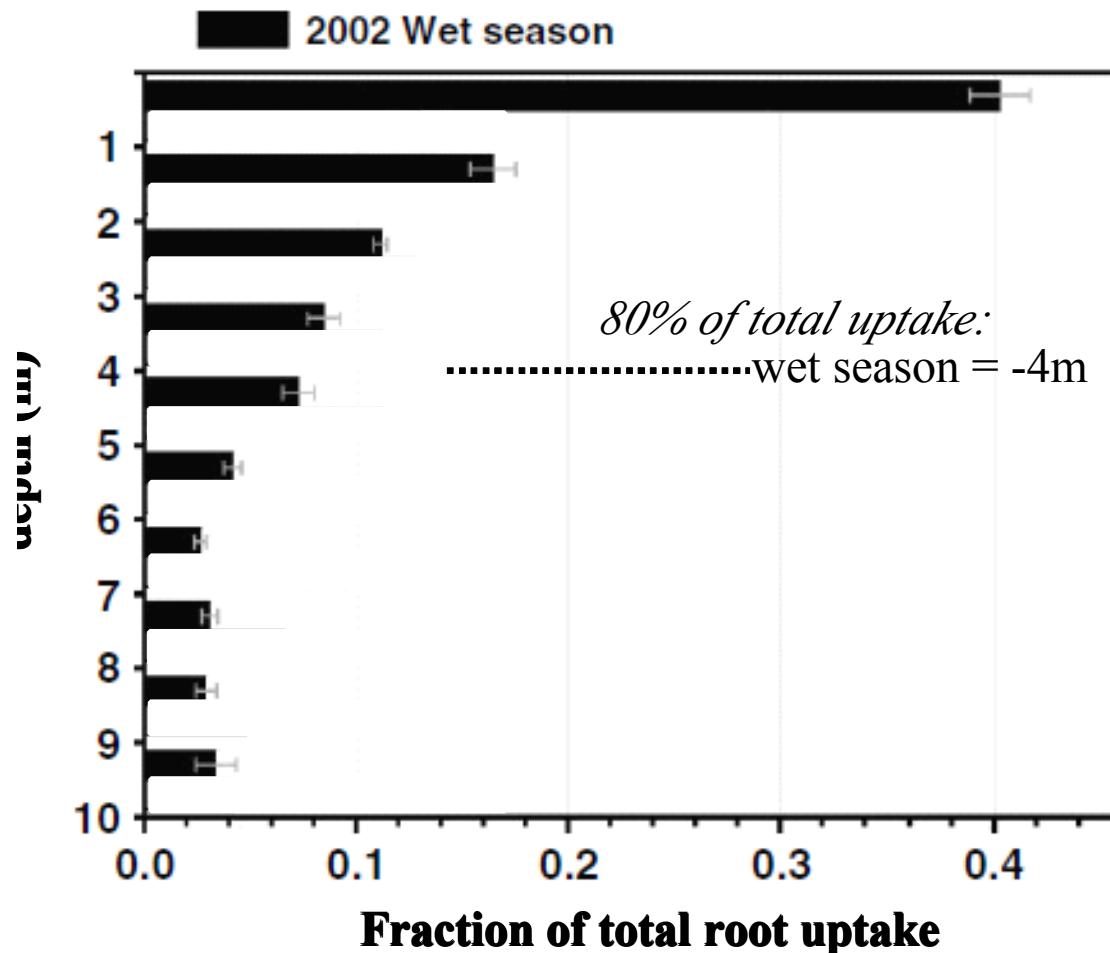
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Dynamic root function (Christoffersen et al., in prep)

Amazon: Observations indicate root uptake shifts to deeper layers during dry season

Tapajos site, observed changes in soil moisture attributed to root uptake



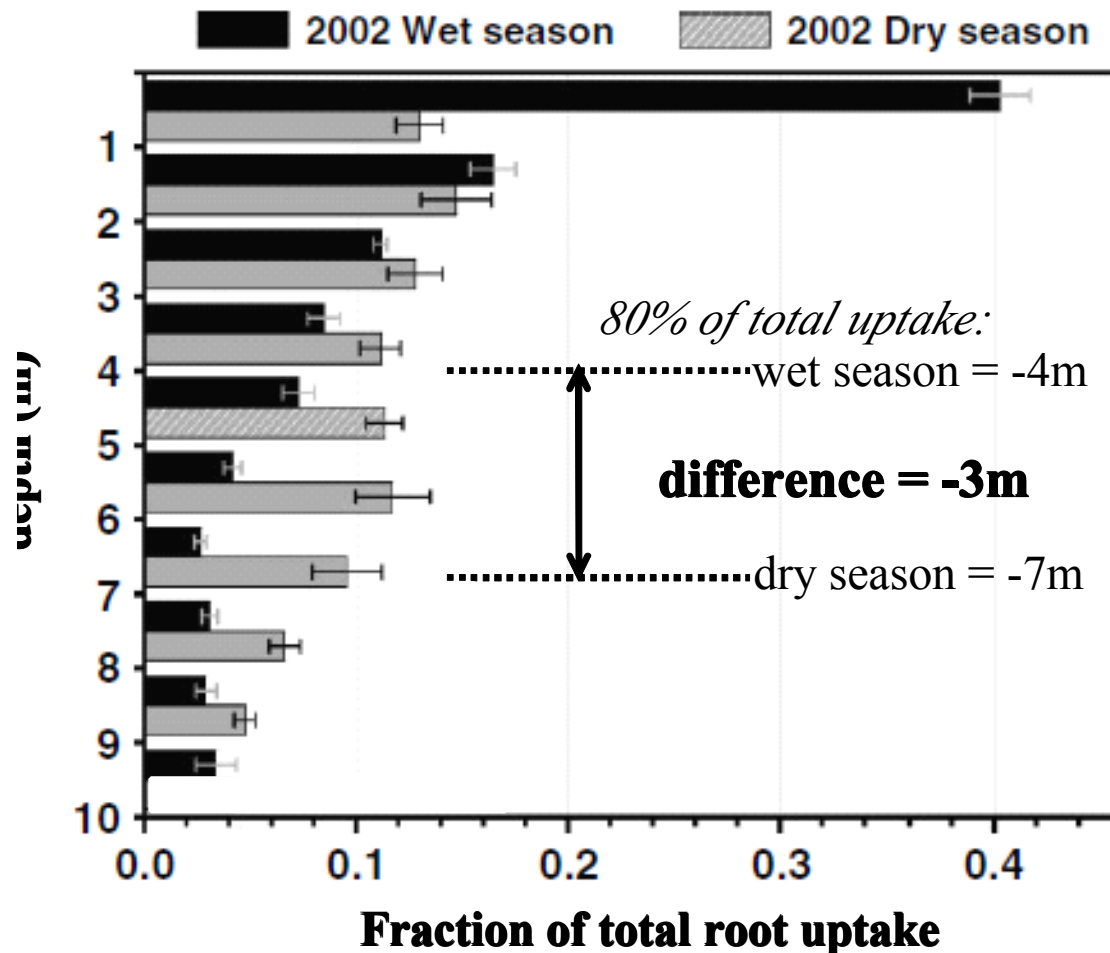
Taken from Bruno et al., 2006

- *Can CLM and other models capture this dynamic aspect of root function?*
- Use a suite of models:
 - CLM3.5-DGVM, IBIS, JULES, ED2, SiB3, SPA
 - standardized soil physics
 - span range of complexity in treatment of root function

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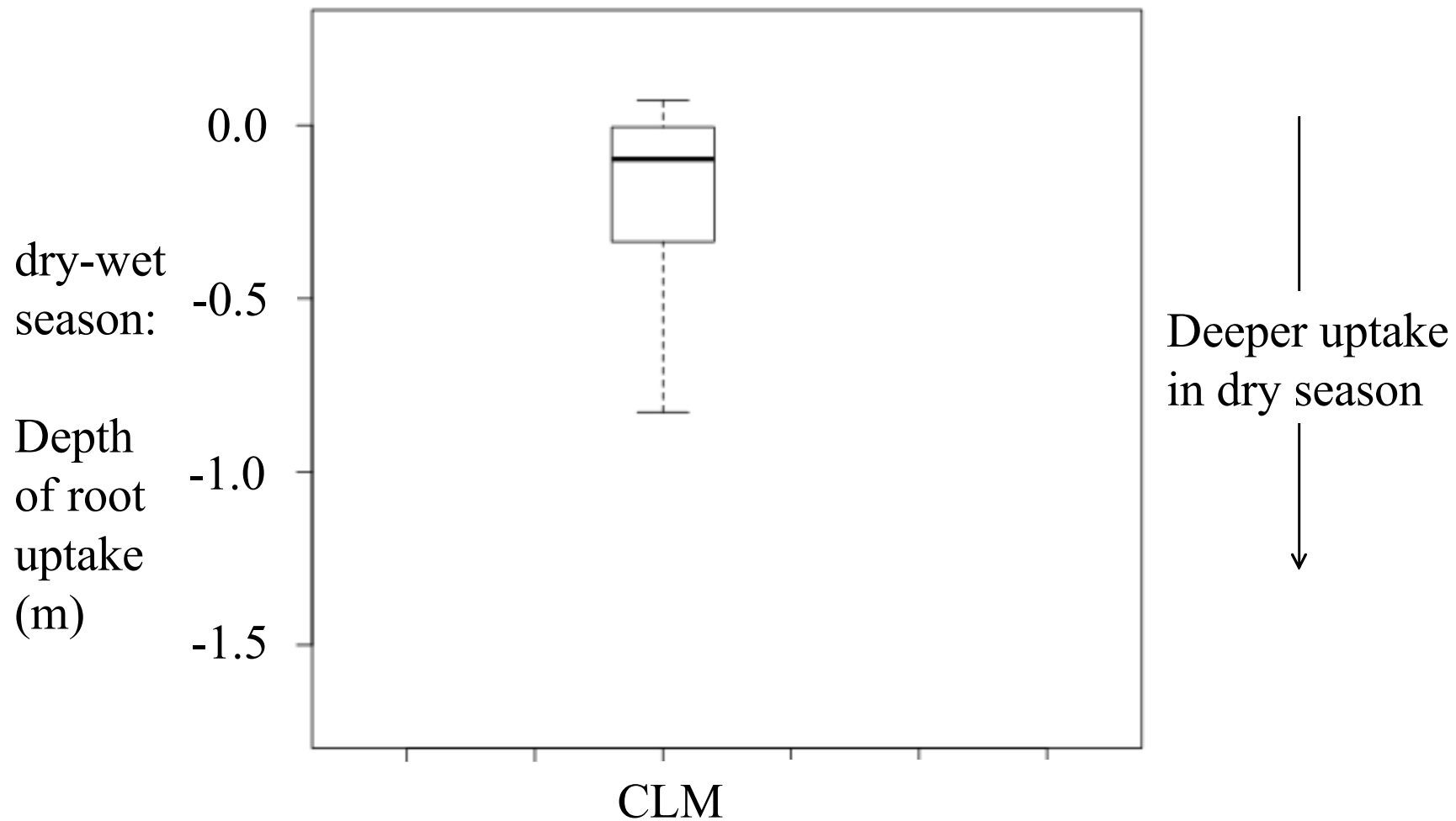


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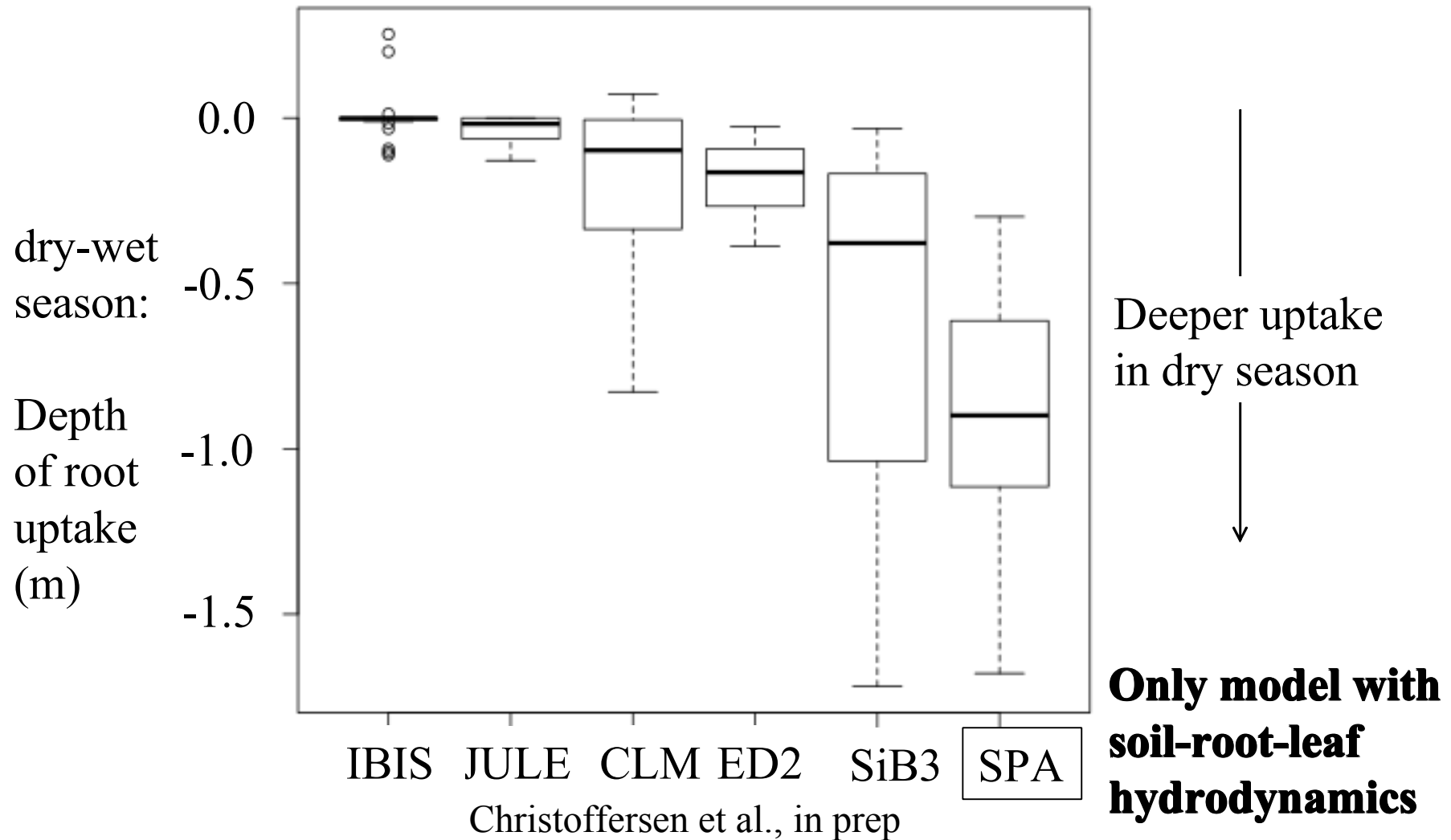
Which model best captures dynamic root behavior?

Difference between wet & dry season depth of root uptake across 4 forest sites



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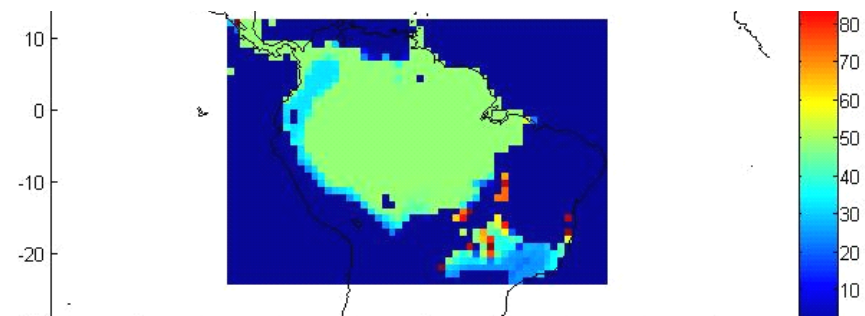
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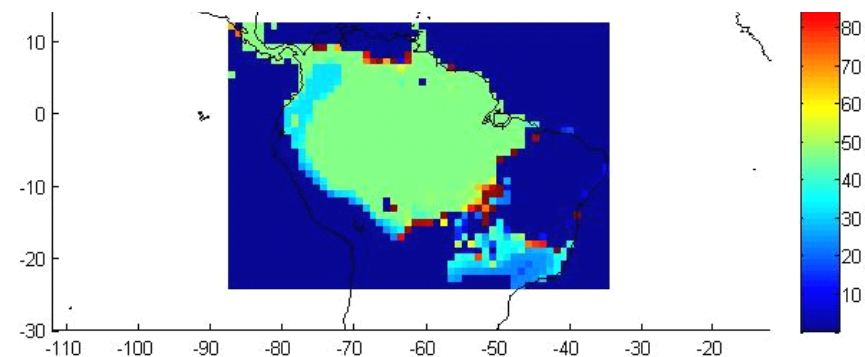
PFT Distribution across Amazon in CLM4 (Moreno et al.)

- PFT distribution, after 200 years, shows coverage of both tropical evergreen and deciduous trees.
- CLM4 over-represents deciduous tree cover in Amazonia.

Broadleaf Deciduous Tropical Tree (%)



Broadleaf Evergreen Tropical Tree (%)

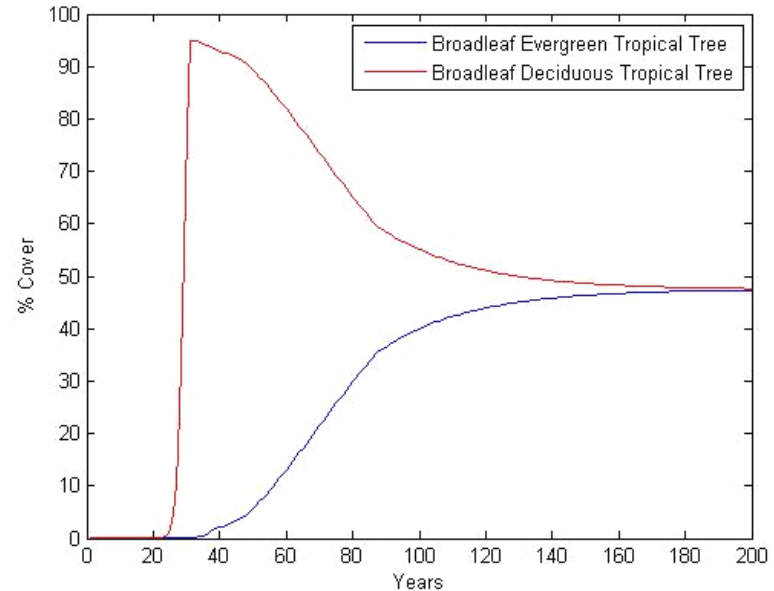


PFT Establishment

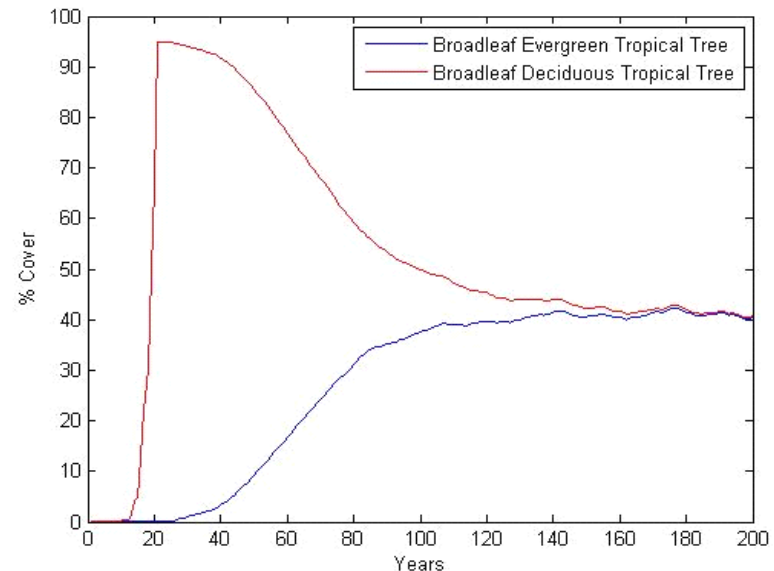
- Initially CLM populates the forest composition with the deciduous tropical trees
- Tropical evergreen trees are slow to establish and do not overtake the deciduous composition.

— deciduous
— evergreen

CAX site (%)



K67 site (%)



CLM4 improvement

- Developed formulations to improve skin temperature over arid regions

Land-atmosphere interaction

- Proposed a simple index for land-precipitation coupling strength
- Demonstrated the impact of interannual climate variability on plant distribution in CLM/DGVM

CLM evaluations

- Identified CLM deficiencies in monthly river flow simulation
- Analyzed the CMIP5 carbon cycle
- analyzed the spinup time in carbon/biomass in CLM-CNDV
- Demonstrated the need for dynamic root function
- identified the PFT distribution deficiency across Amazon in CLM-CNDV