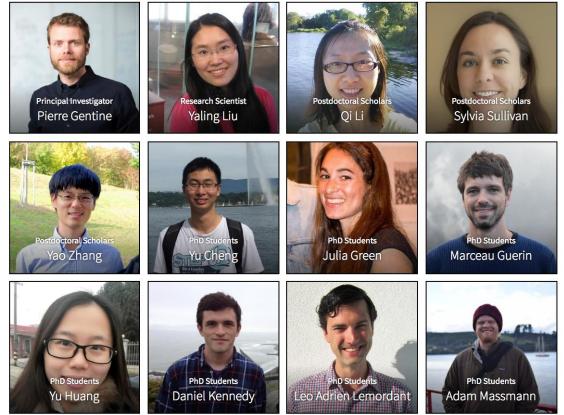
Implementing plant hydraulics in the NCAR Community Land Model (CLM) and the implications for drought vulnerability

# Daniel Kennedy, Pierre Gentine Columbia University

Sean Swenson, Keith Oleson, Rosie Fisher, Dave Lawrence, and the LMWG

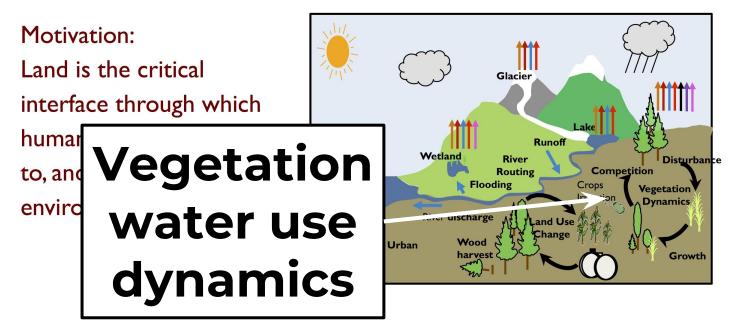
# COLUMBIA ENGINEERING Gentine Lab

"Investigating the terrestrial carbon and water cycles using multiscale modeling and observations"



### Community Land Model (CLM)

Consegue Conserve.

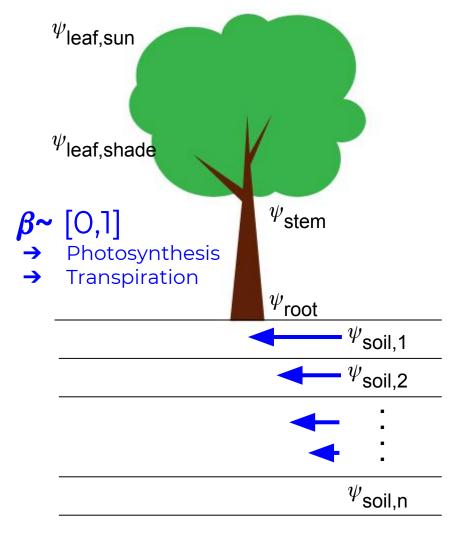


Comprehensive representations of land biogeophysics, hydrology, plant physiology, biogeochemistry, anthropogenic land use, and ecosystem dynamics

slide courtesy: Dave Lawrence

- vegetation water use dynamics derived from prognostic vegetation water potential
- CLM5 default
- 1. water stress ( $\beta$ )
  - function of leaf water potential
- 2. root water uptake
  - utilizes Darcy's Law approximation
  - gradient based on root water potential

$$\circ$$
 q = -k (  $\psi_{\text{root}}$  -  $\psi_{\text{soil}}$  )



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  - gradient based on root water potential

 $\circ \quad$  q = -k (  $\psi_{\rm root}$  -  $\psi_{\rm soil}$  )

### **Soil Moisture Stress**

- vegetation water use dynamics parameterized directly with soil potential
- CLM4.5 default
- 1. water stress ( $\beta$ )
  - function of soil water potential
- 2. root water uptake
  - partitioning heuristic
  - gradient based on constant parameter

$$ho$$
 q = -k (  $\psi_{\rm c}$  -  $\psi_{\rm soil}$  )

- $\psi_{\rm c}$  = soil water potential with stomates fully closed
  - varies by PFT
  - typically around -2.5 MPa

- vegetation water use dynamics derived from prognostic vegetation water potential
- CLM5 default

### **Soil Moisture Stress**

- vegetation water use dynamics parameterized directly with soil potential
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### JAMES Journal of Advances in Modeling Earth Systems



#### **RESEARCH ARTICLE**

10.1029/2018MS001500

#### **Special Section:**

Community Earth System Model version 2 (CESM2) Special Collection

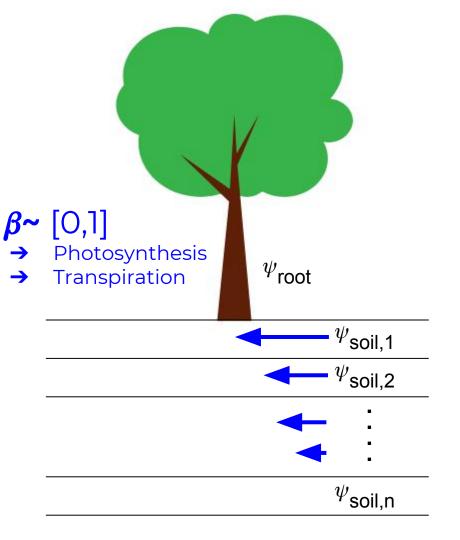
### Implementing Plant Hydraulics in the Community Land Model, Version 5

Daniel Kennedy<sup>1</sup>, Sean Swenson<sup>2</sup>, Keith W. Oleson<sup>2</sup>, David M. Lawrence<sup>2</sup>, Rosie Fisher<sup>2</sup>, Antonio Carlos Lola de Costa<sup>3</sup>, and Pierre Gentine<sup>1</sup>

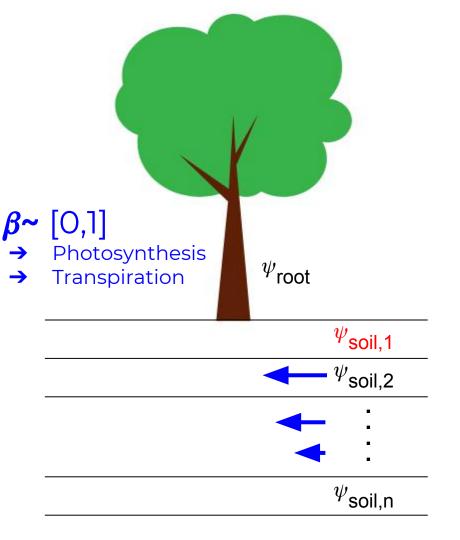
<sup>1</sup>Columbia University, New York, NY, USA, <sup>2</sup>Nation<sup>1</sup> Center for Atmospheric Research, Boulder, CO, USA, <sup>3</sup>Centro de

Features output from a point simulation in Caxiuana, Brazil

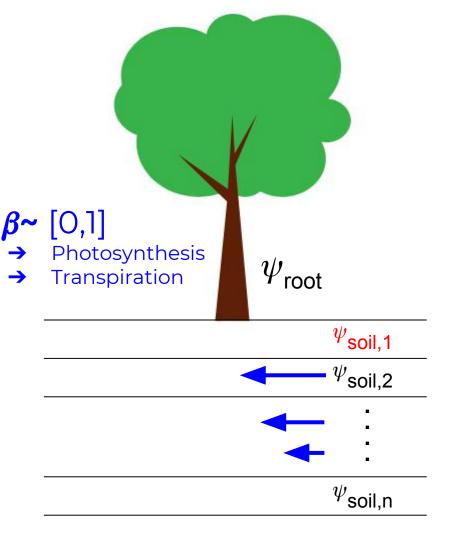
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   prefers to use water near the surface, but has the flexibility to look deeper



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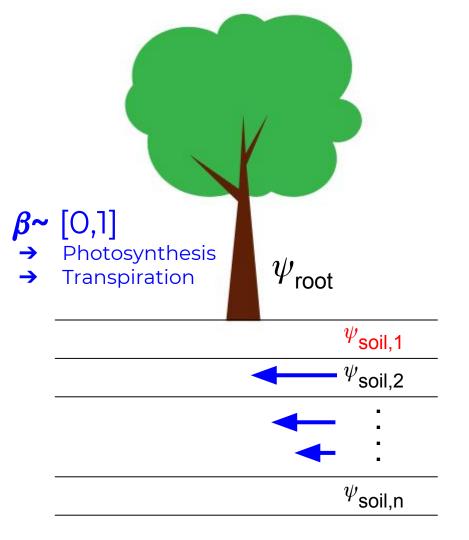


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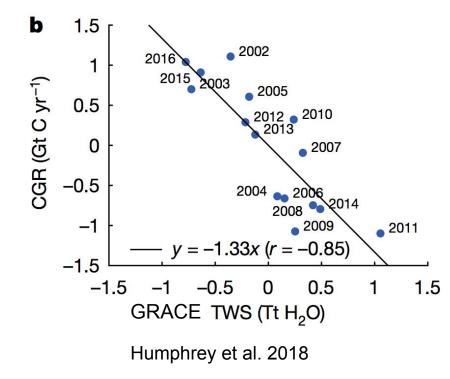
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# How will this influence the model?



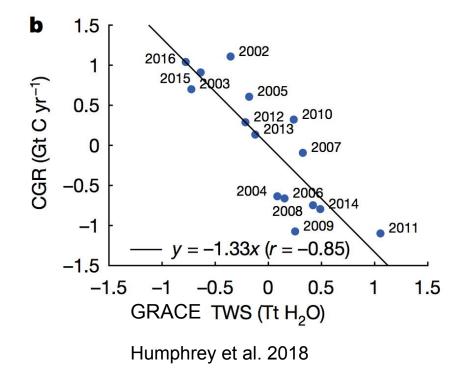
tree from vecteezy.com

# How will PHS root water uptake affect the global carbon cycle?



Interannual variability in the atmospheric carbon growth rate has been linked to total water storage

# How will PHS root water uptake affect the global carbon cycle?



Interannual variability in the atmospheric carbon growth rate has been linked to total water storage...

but this relationship is not well-represented in TRENDYv3

# How will PHS root water uptake affect the vertical distribution of soil water?

CMIP5 Simulations project stronger drying trends in the top 10cm of the soil column

# **AGU** PUBLICATIONS

### Geophysical Research Letters

#### **RESEARCH LETTER**

10.1002/2016GL071921

#### **Key Points:**

 We identify a robust vertical gradient of projected soil moisture changes under global warming with more negative changes near the surface

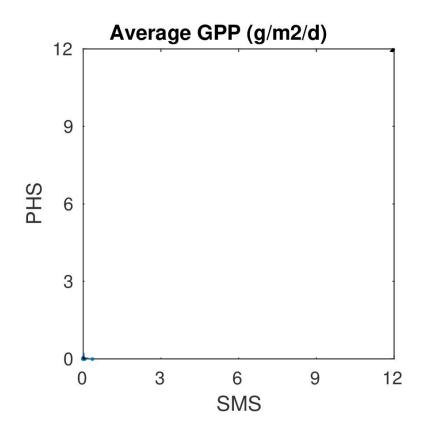
# Divergent surface and total soil moisture projections under global warming

Alexis Berg<sup>1</sup>, Justin Sheffield<sup>1,2</sup>, and P. C. D. Milly<sup>3</sup>

<sup>1</sup>Department of Civil and Environmental Engineering, Princeton University, Princeton, New Jersey, USA, <sup>2</sup>Geography and Environment, University of Southampton, Southampton, UK, <sup>3</sup>U.S. Geological Survey, and NOAA/Geophysical Fluid



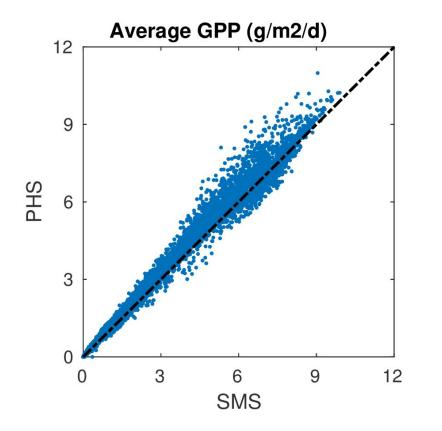
# What are the effects of plant hydraulics globally?



#### **Simulation details**

- 2 simulations (PHS,SMS)
- CLM5 (global)
- satellite phenology
- 1850-2014
- only difference is PHS/SMS

# What are the effects of plant hydraulics globally?



PHS features higher GPP on average

- plotting average GPP (1914-2013) for every gridcell
- Overall, GPP is ~7% higher with PHS
- Meaning that water stress is lower with PHS

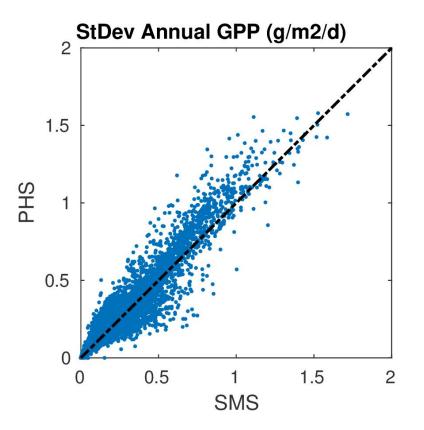
### **Simulation details**

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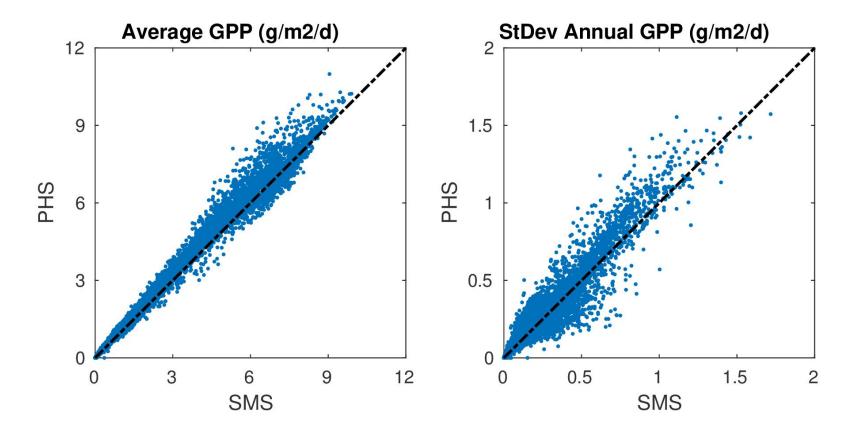
PHS also features higher GPP interannual variability

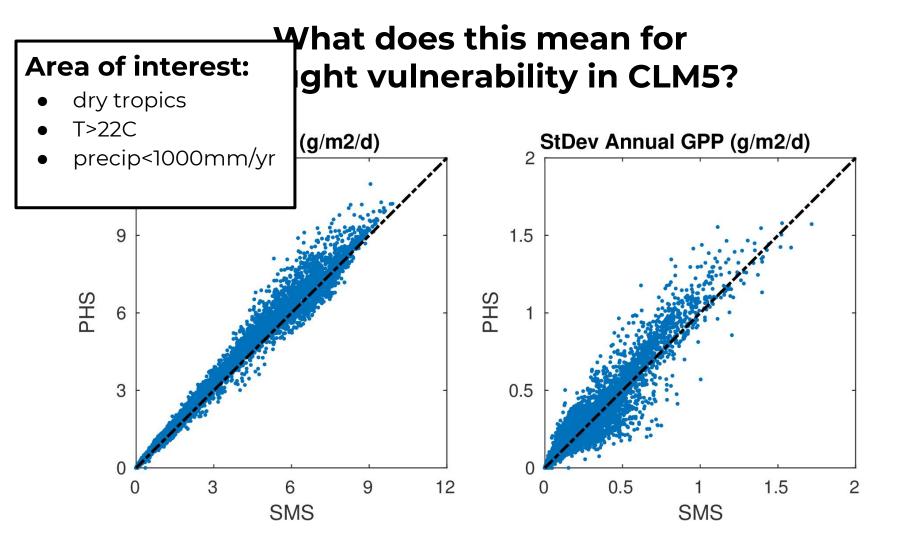
- plotting StDev of annual GPP (1914-2013) for every gridcell
- mixed bag, but overall GPP IAV increases with PHS (+7%)

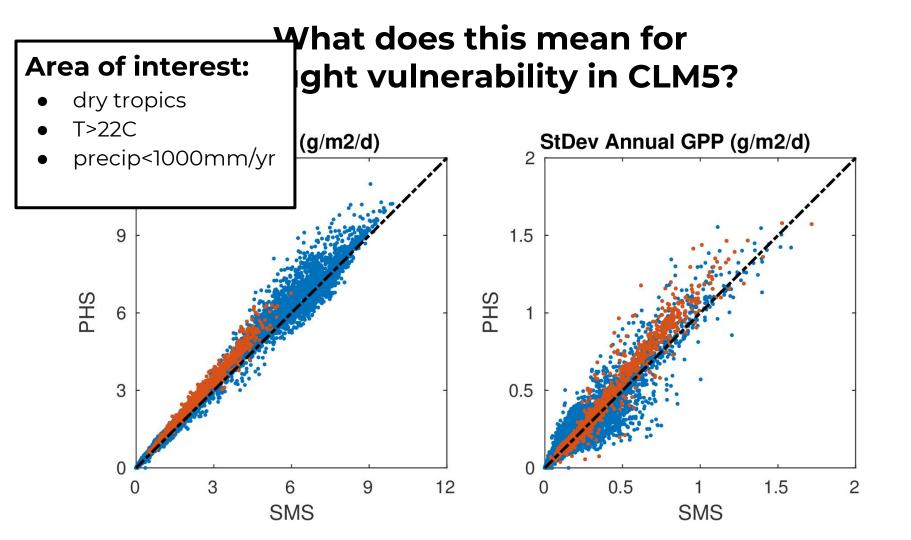
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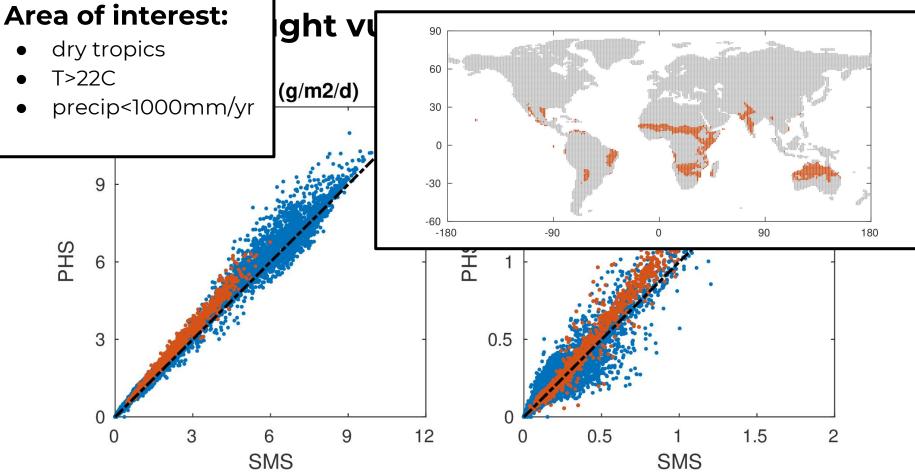
# What does this mean for drought vulnerability in CLM5?

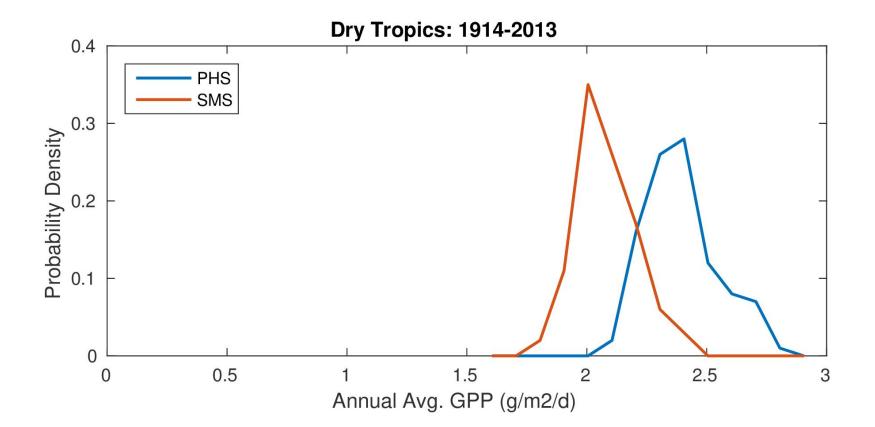


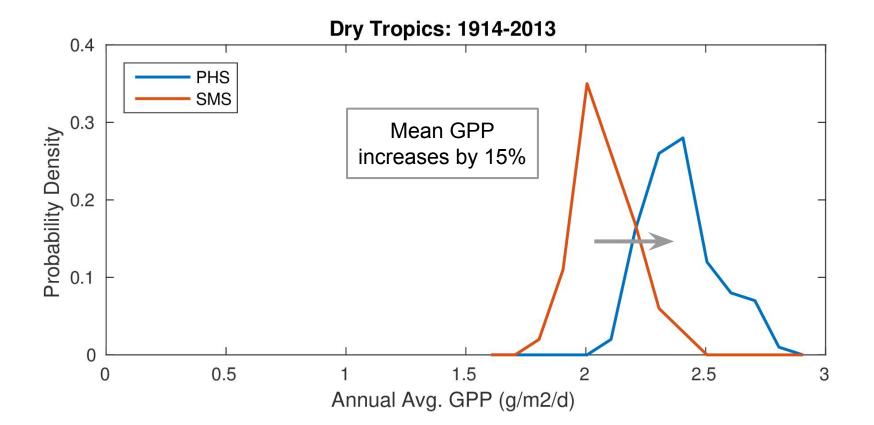


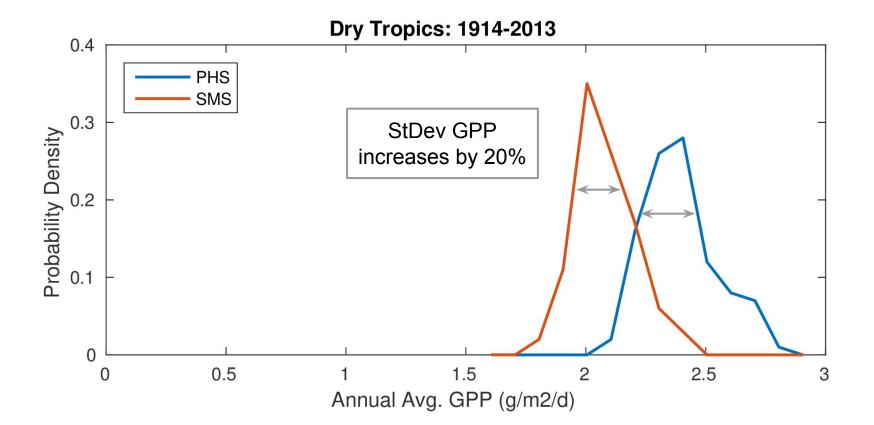


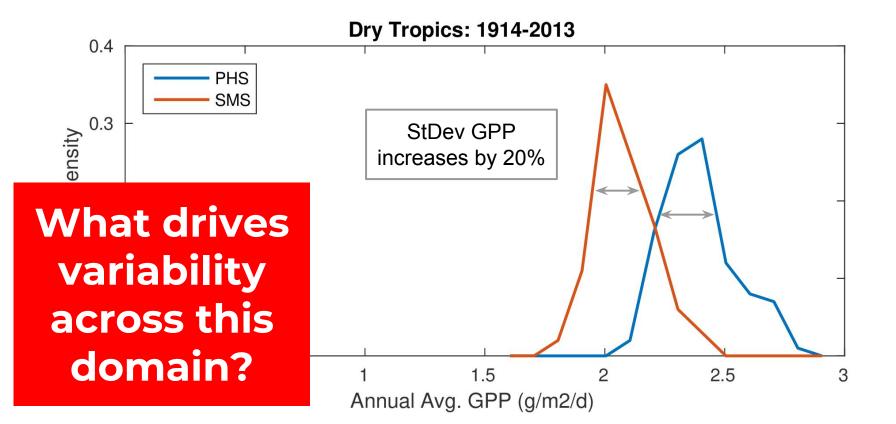
### **Y**hat does this mean for

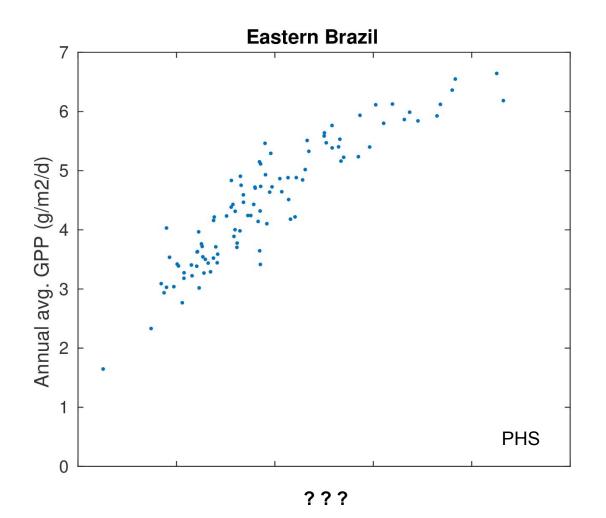








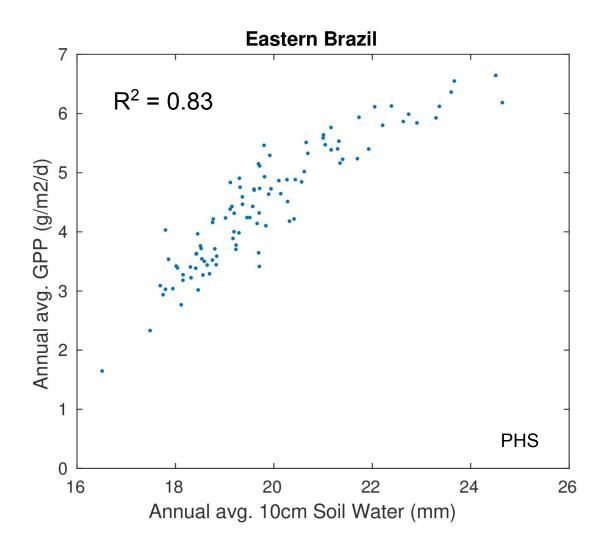




What drives model interannual variability in GPP?

- example gridcell
- each point represents one year
   101( 2017)
  - 1914-2013

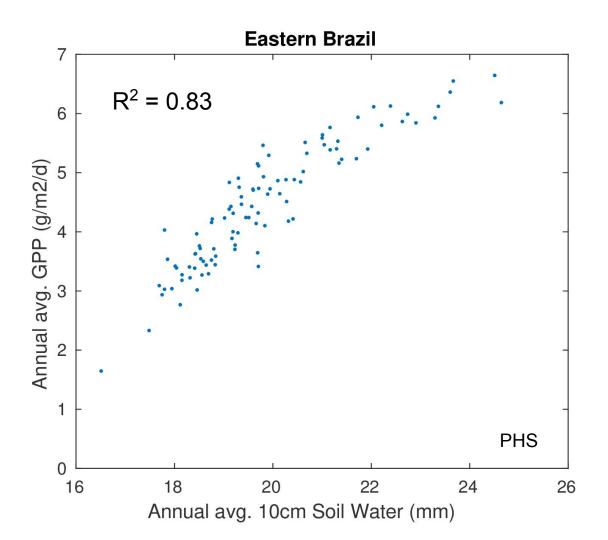
• 
$$\sigma = 1.03 \text{ g/m}^2/\text{d}$$



What drives model interannual variability in GPP?

~soil moisture~

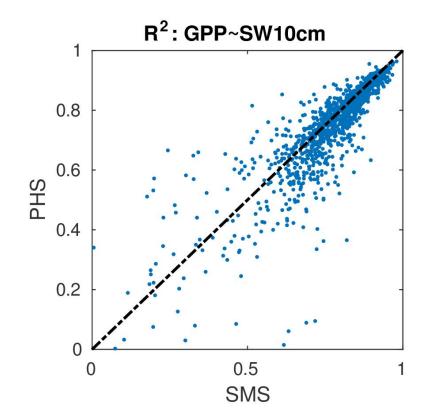
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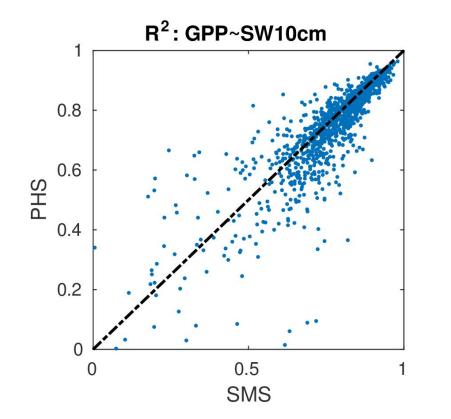
~soil moisture~

- example gridcell
- each point represents one year
  - o **1914-2013**
- σ = 1.03 g/m2/d
- is this true across the full domain?
- with both models?



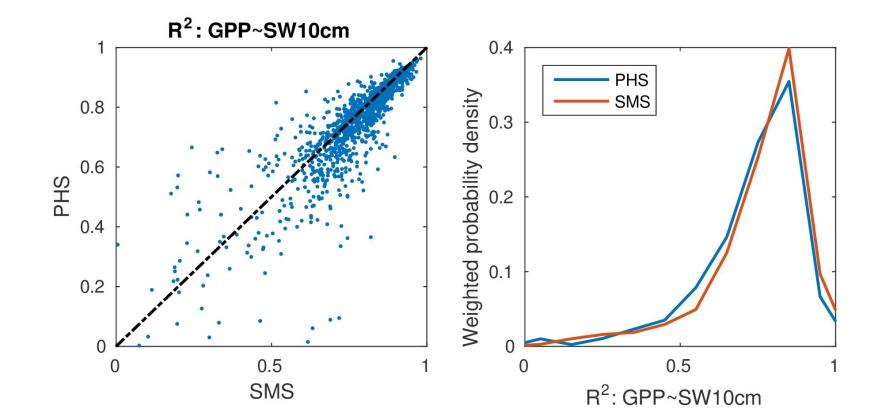
### Plot details:

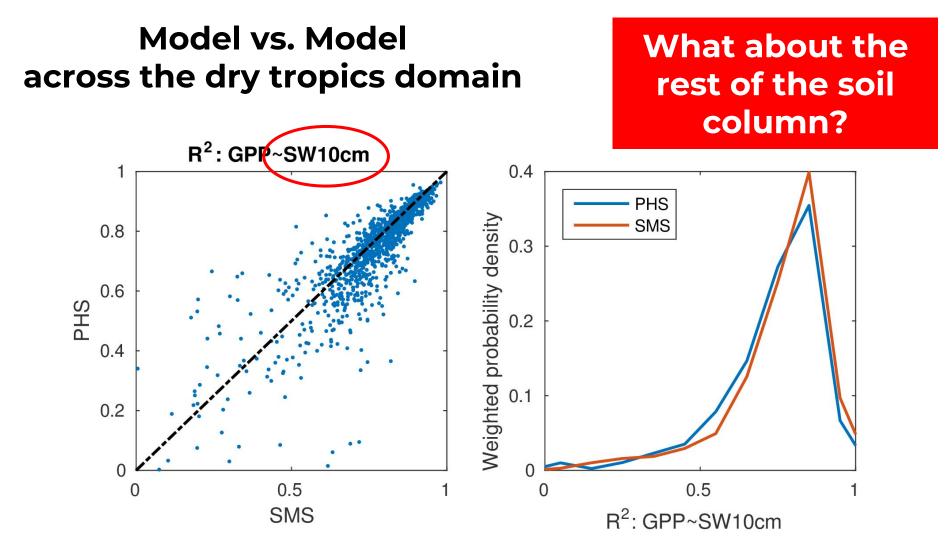
- Each point is a different grid cell
- Looking at the correlation between GPP and 10cm soil water
- Computed from time-series of annual average values (1914-2013)



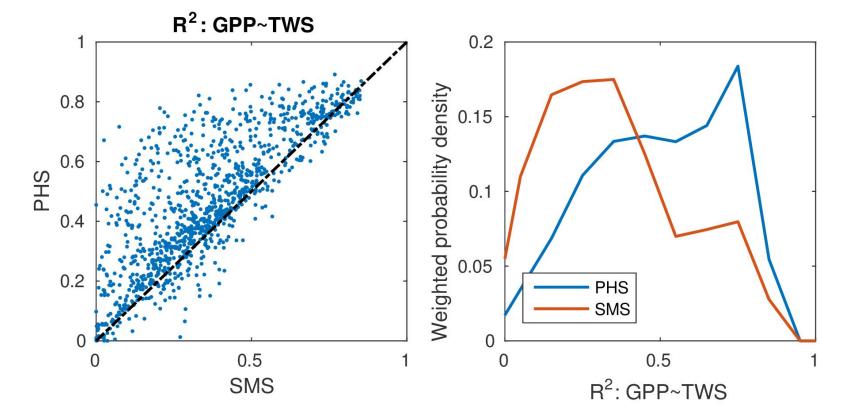
### Plot details:

- Each point is a different grid cell
- Looking at the correlation between GPP and 10cm soil water
- Computed from time-series of annual average values (1914-2013)
- Overall: soil water explains a very large portion of the interannual variability in GPP
- True for both models

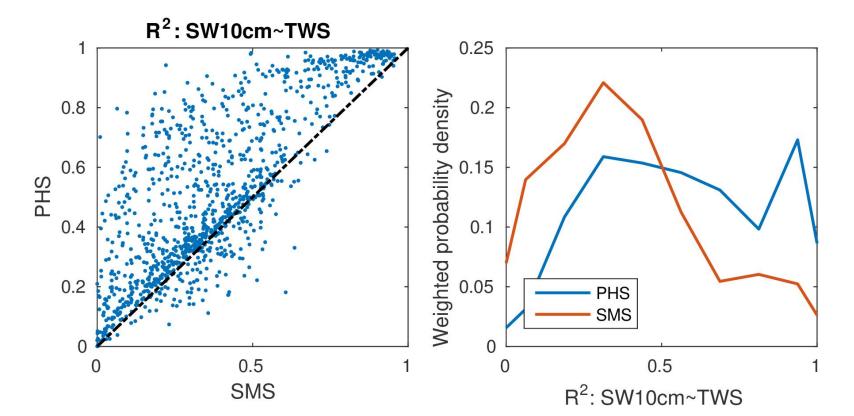


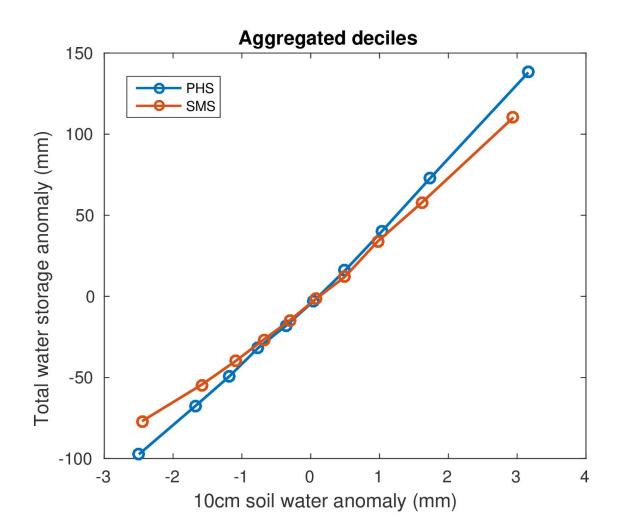


#### PHS: GPP more closely coupled with Total Water Storage (TWS)



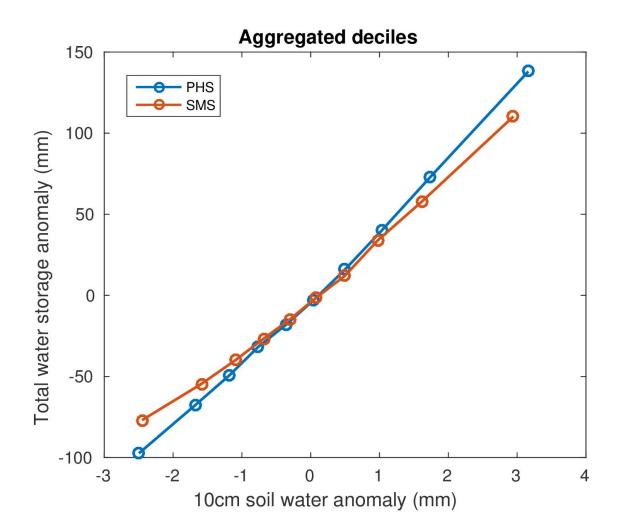
PHS: Likewise Total Water Storage (TWS) more closely coupled with 10cm soil water





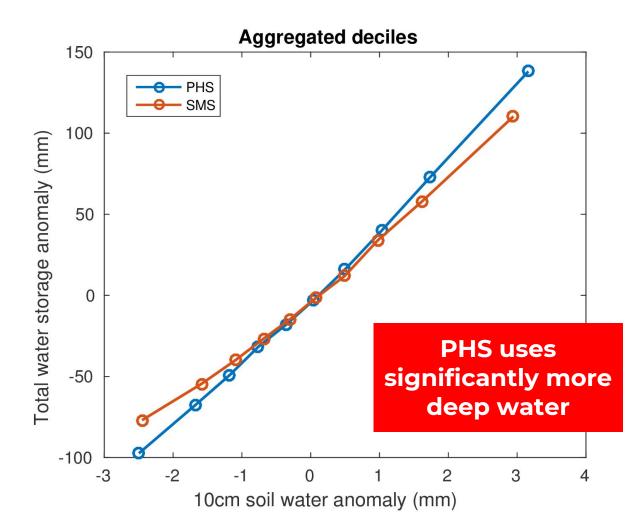
### Plot details

- bin each gridcell into deciles based on 10cm soil water
  - 10 yrs per decile
- then average across the full study area



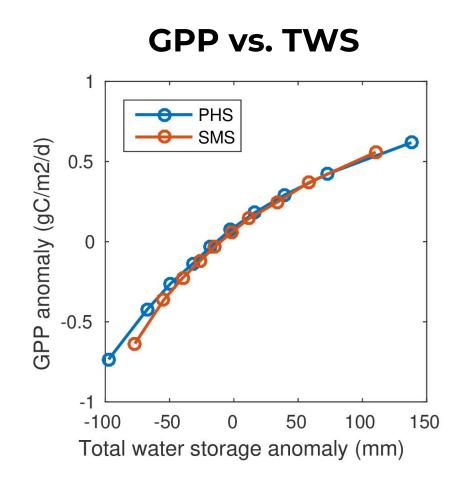
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- bin each gridcell into deciles based on 10cm soil water
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- PHS yields ~41.5 mm TWS for every 1mm from the top 10cm
- SMS = 35.1

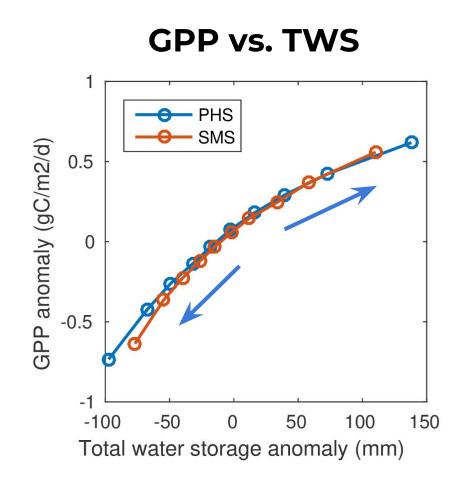


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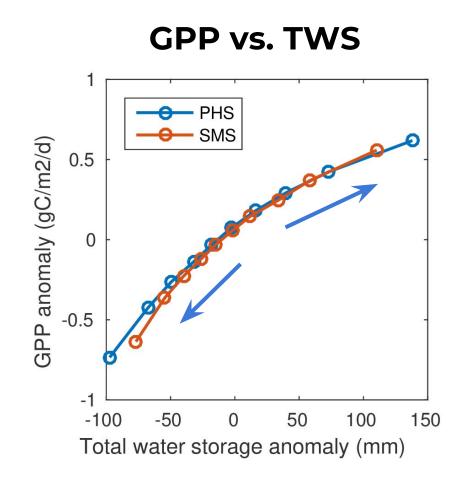
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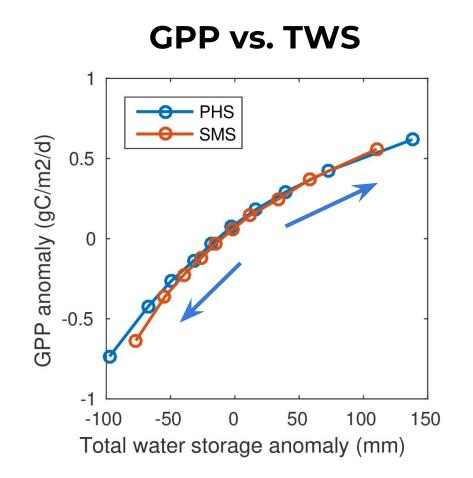
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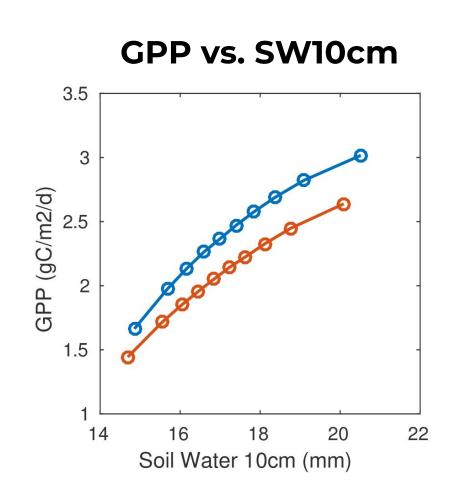


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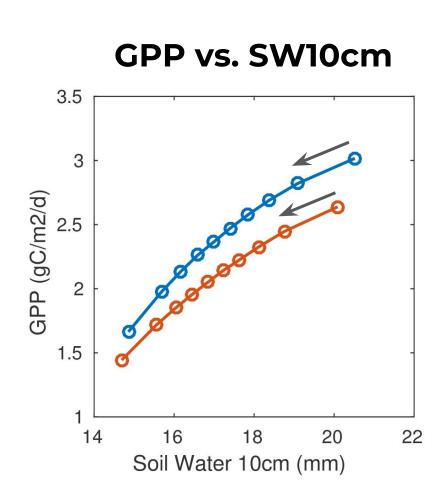


- SMS productivity looks slightly more sensitive to TWS anomalies (steeper slope)
- But PHS samples a wider range of TWS
- Associated with the observed increase in GPP variability
- How does this affect
  - 10cm soil water?
  - Mean GPP?

- PHS GPP is about 15% higher
   requires higher ET
- Despite this, surface soils are slightly wetter with PHS
  - partially explains higher
     GPP



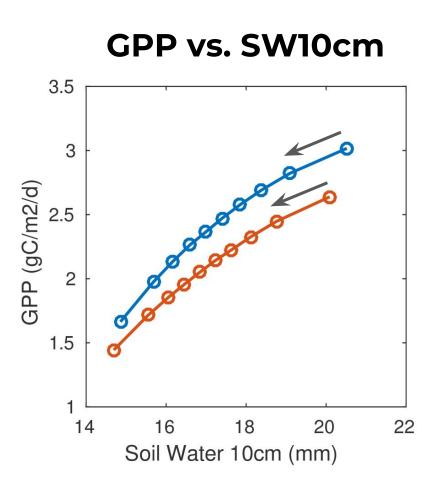
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### **Conclusions:**

- PHS uses more 'deep' soil water (beyond 10cm)
- Leads to higher GPP
- But also associated with higher IAV, higher correlation with TWS



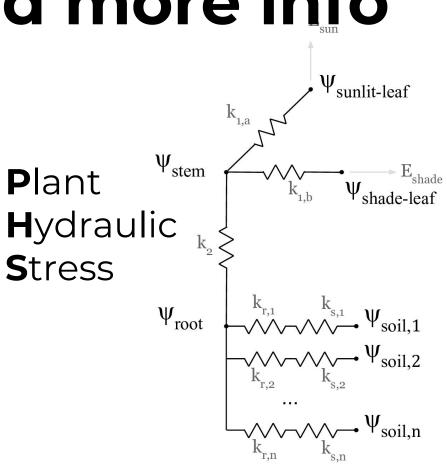
# where to find more info

### **CLM Technical note**

- Section 2.11
- cesm.ucar.edu/models/cesm2/land/

#### Model development paper

- Implementing plant hydraulics in the Community Land Model, version 5
- Kennedy et al. 2019, JAMES



#### djk2120@columbia.edu

# where to find more info

### **CLM Technical note**

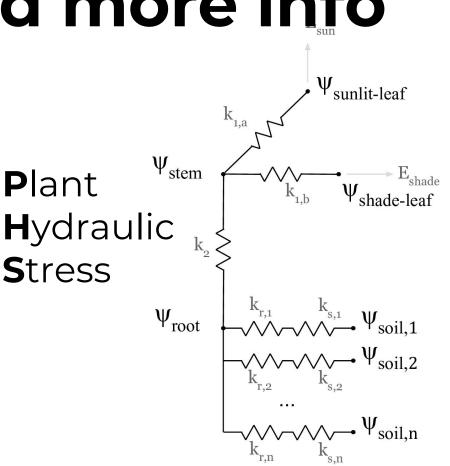
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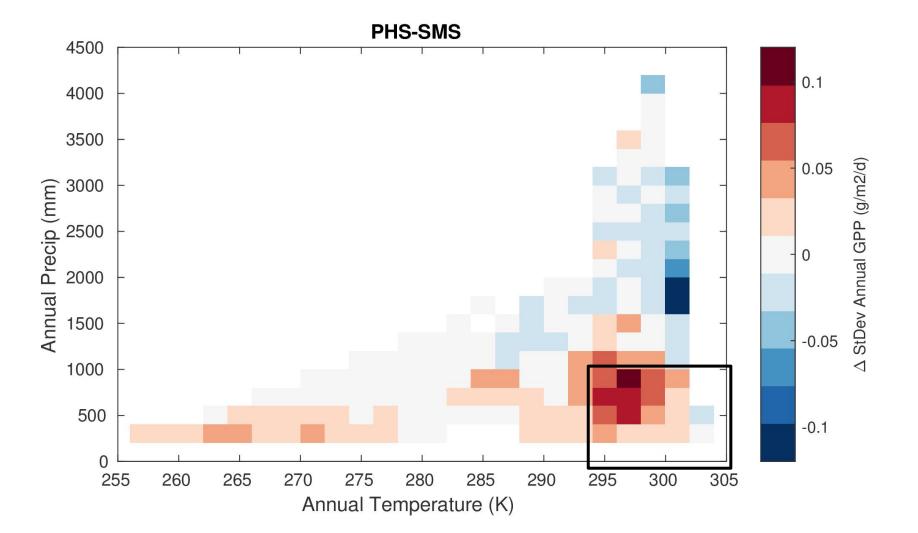
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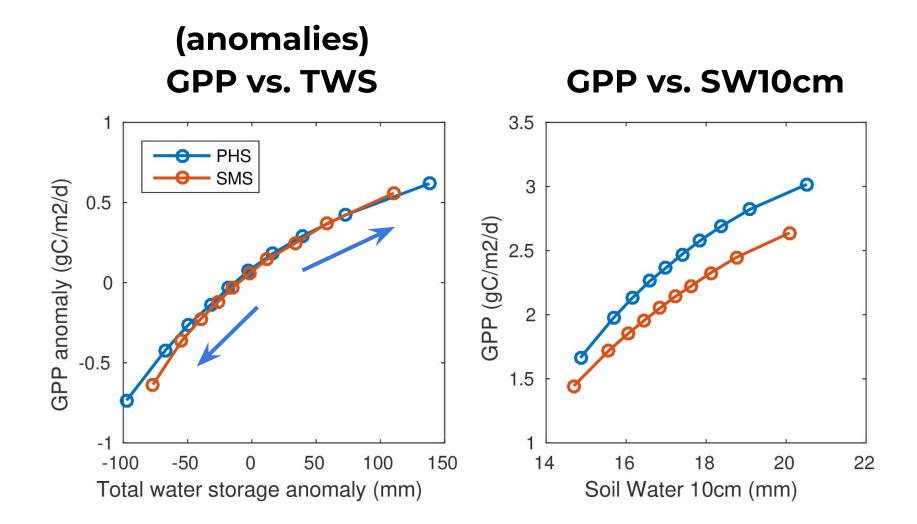
## thanks!

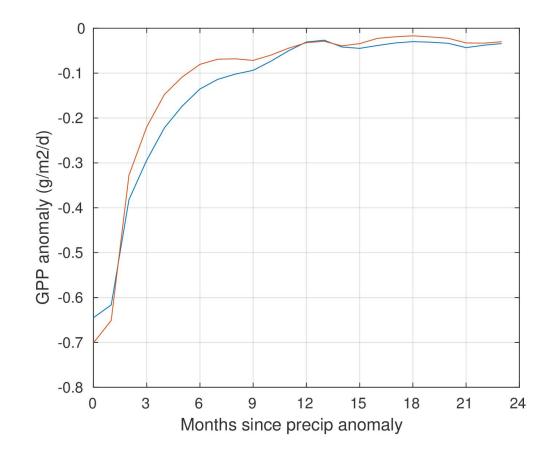
djk2120@columbia.edu



extra slides

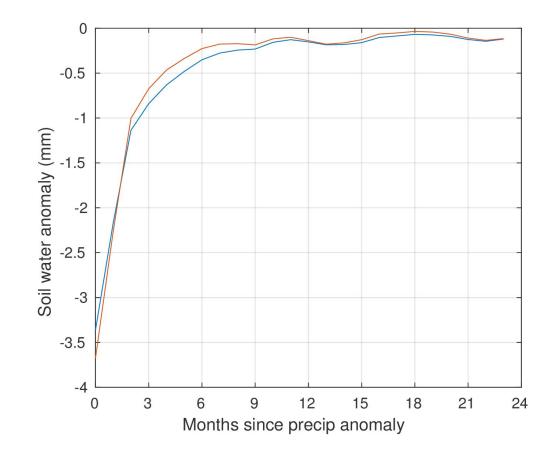






Precip anomalies have longer coupling to GPP (with PHS)

- lost GPP (year 1)
  - PHS = 1.23 PgC
  - SMS = 1.08 PgC
- lost GPP (year 2)
  - PHS = 0.18 PgC
  - SMS = 0.14 PgC
- monthly mean GPP
  - ~0.9 PgC
  - (for reference)



Similar pattern with soil water (top 10cm)

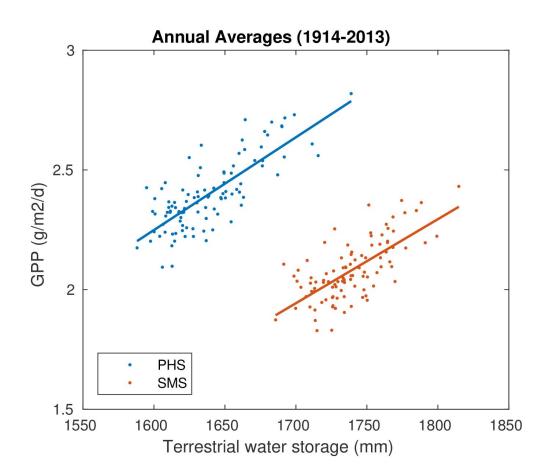
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monthly mean GPP

• ~0.9 PgC

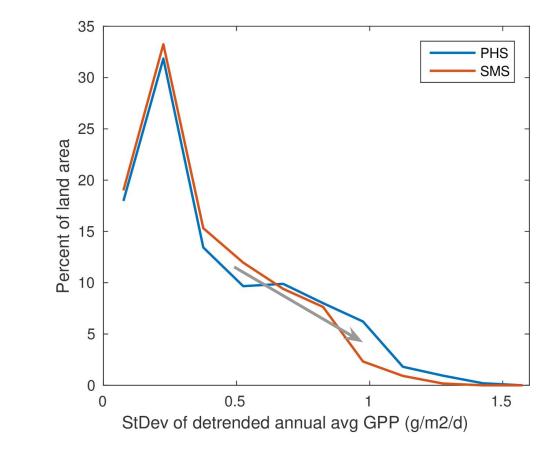
• (for reference)



## PHS features higher variability in GPP

### Interannual Variability also increases

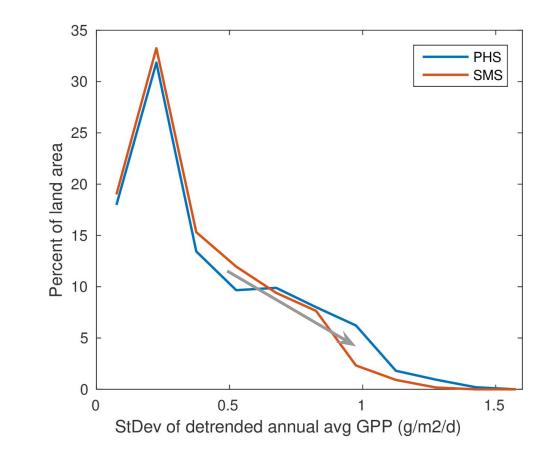
 distribution shifts to higher variability



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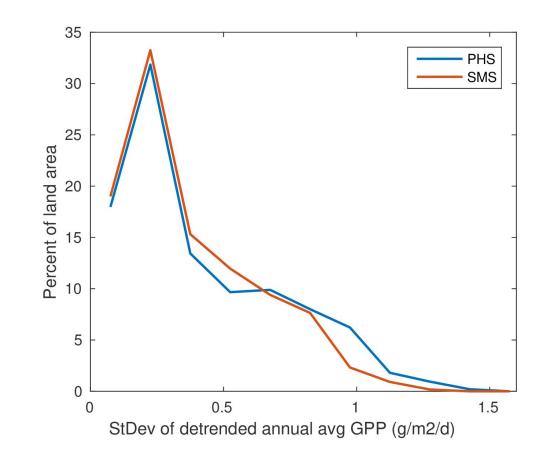
- distribution shifts to higher variability
- average stdev:
  - PHS = 0.41 g/m<sup>2</sup>/d
  - SMS =  $0.37 \text{ g/m}^2/\text{d}$



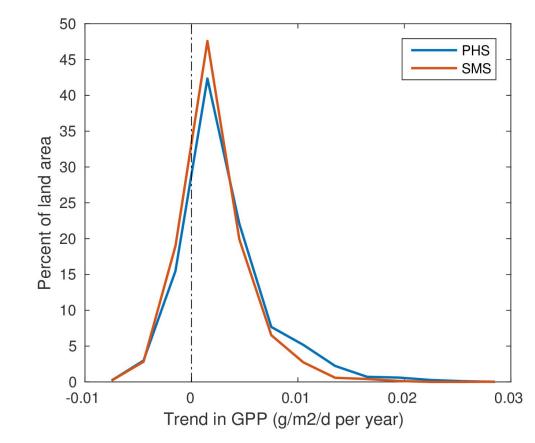
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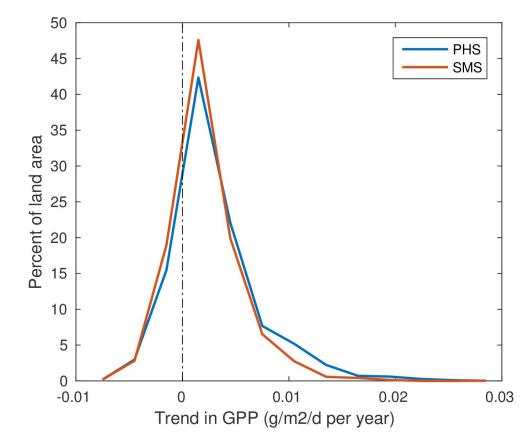
- distribution shifts to higher variability
- average stdev:
  - PHS = 0.41 g/m<sup>2</sup>/d
  - SMS =  $0.37 \text{ g/m}^2/\text{d}$
- approx 12% increase



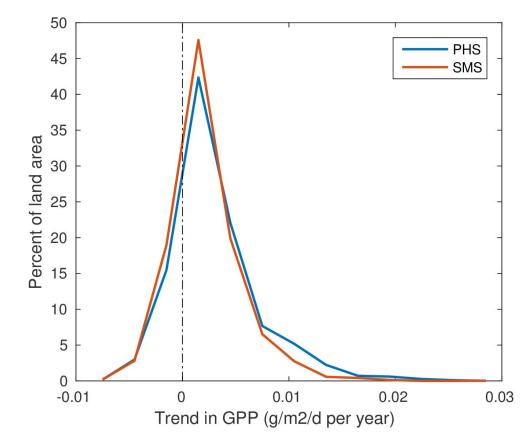
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- distribution shifts to higher trends
- integrated over the study domain, GPP trend adds
   1.53 Pg/yr with PHS



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- distribution shifts to higher trends
- integrated over the study domain, GPP trend adds
   1.53 Pg/yr with PHS
- compared to
   1.06 Pg/yr with SMS
- (1914) GPP is approx. 11 Pg/yr

