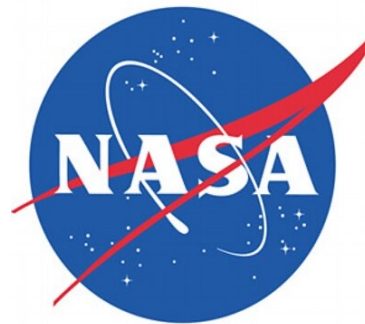


# Carbon Monitoring System in Mountains (CMS-Mountains): Development and Testing in the Western U.S. (Lin-CMS 2015 & Lin-CMS 2018)

John C. Lin, **Brett Raczka**, Henrique Duarte, David R. Bowling,  
Jeffrey L. Anderson, Timothy J. Hoar, Christian Frankenberg, Philipp  
Koehler, Karen Yuen



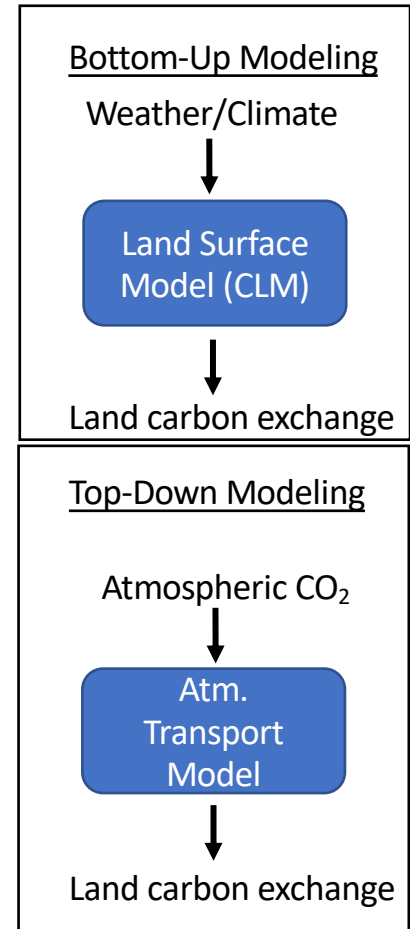
Caltech



# Goal: Monitor carbon flux across complex terrain of Western US

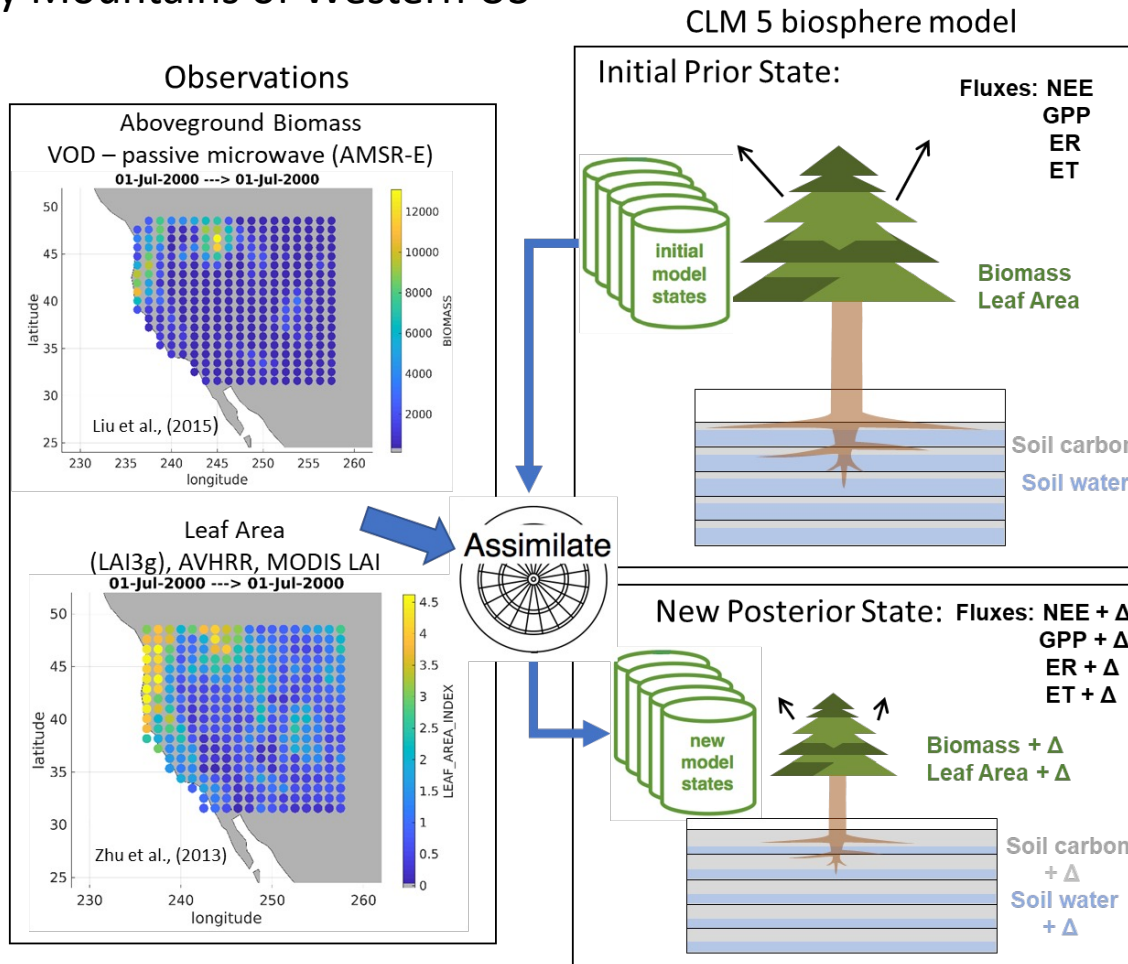


Vulnerable carbon stocks, drastic change to landscape and ecosystem functioning

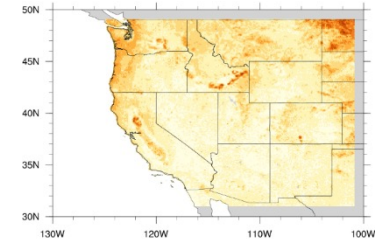


# Overarching Goal: Develop Land surface data assimilation system, CLM5-DART

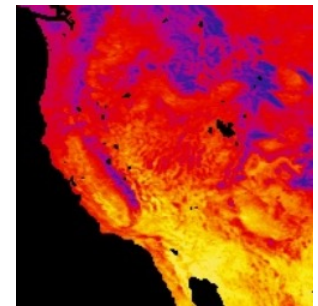
“Develop a carbon monitoring system across complex terrain of Rocky Mountains of Western US”



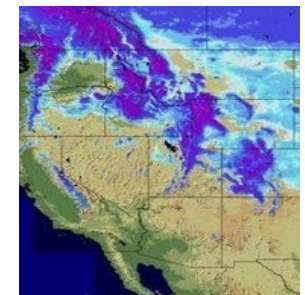
Solar Induced Fluorescence (SIF)



(TROPOMI, OCO-2/3, GOME-2)



Land Surface Temp  
(ECOSTRESS)



Snow Cover (MODIS)

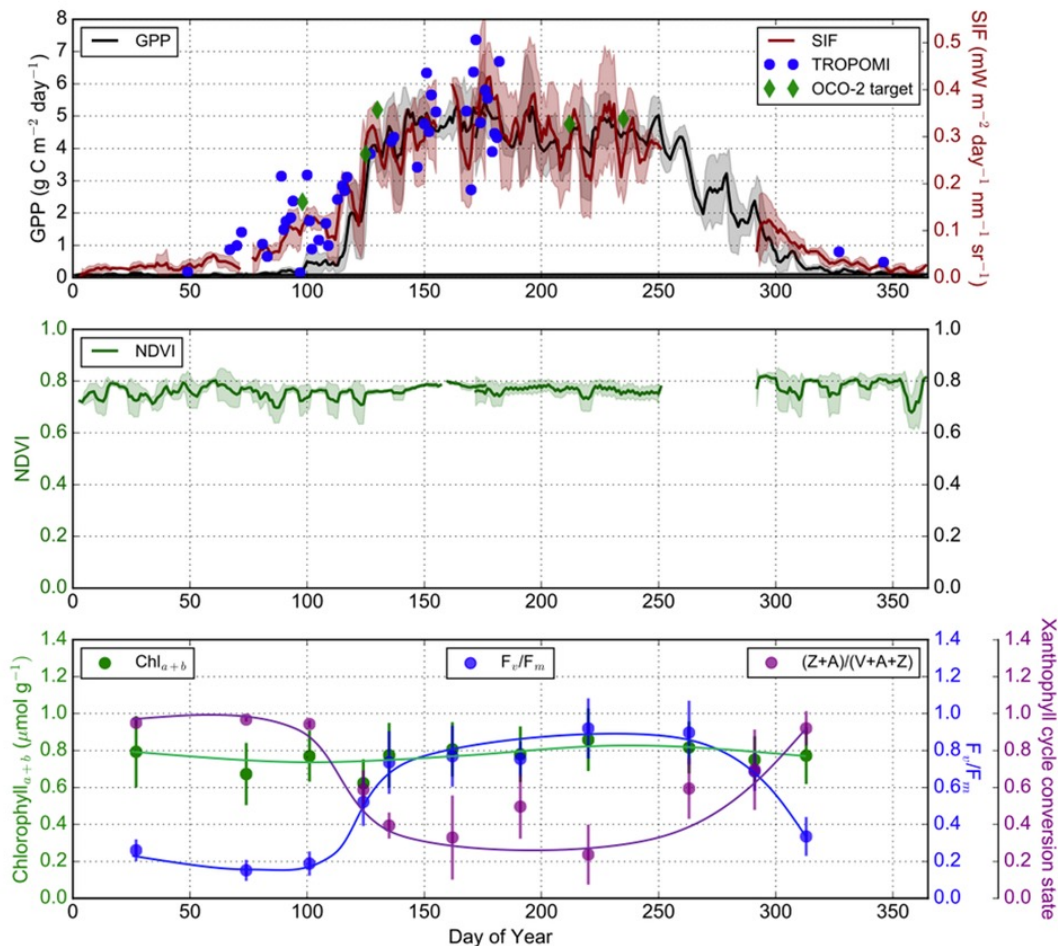
# CMS-Mountains-I Review

# Demonstration of strong SIF-GPP relationship for Western US evergreen species



## Mechanistic evidence for tracking the seasonality of photosynthesis with solar-induced fluorescence

Troy S. Magney<sup>a,b,1</sup>, David R. Bowling<sup>c</sup>, Barry A. Logan<sup>d</sup>, Katja Grossmann<sup>e,2</sup>, Jochen Stutz<sup>e</sup>, Peter D. Blanken<sup>f</sup>, Sean P. Burns<sup>g,3</sup>, Rui Cheng<sup>h</sup>, Maria A. Garcia<sup>c</sup>, Philipp Köhler<sup>c</sup>, Sophia Lopez<sup>d</sup>, Nicholas C. Parazoo<sup>b</sup>, Brett Raczka<sup>c</sup>, David Schimel<sup>b</sup>, and Christian Frankenberg<sup>a,b,1</sup>



- SIF is a useful indicator of timing/magnitude of GPP (Niwot Ridge, CO)
- Traditional 'green-ness' indicators do not track seasonal GPP
- The GPP seasonality related to leaf pigment transition (xanthophyll cycle)

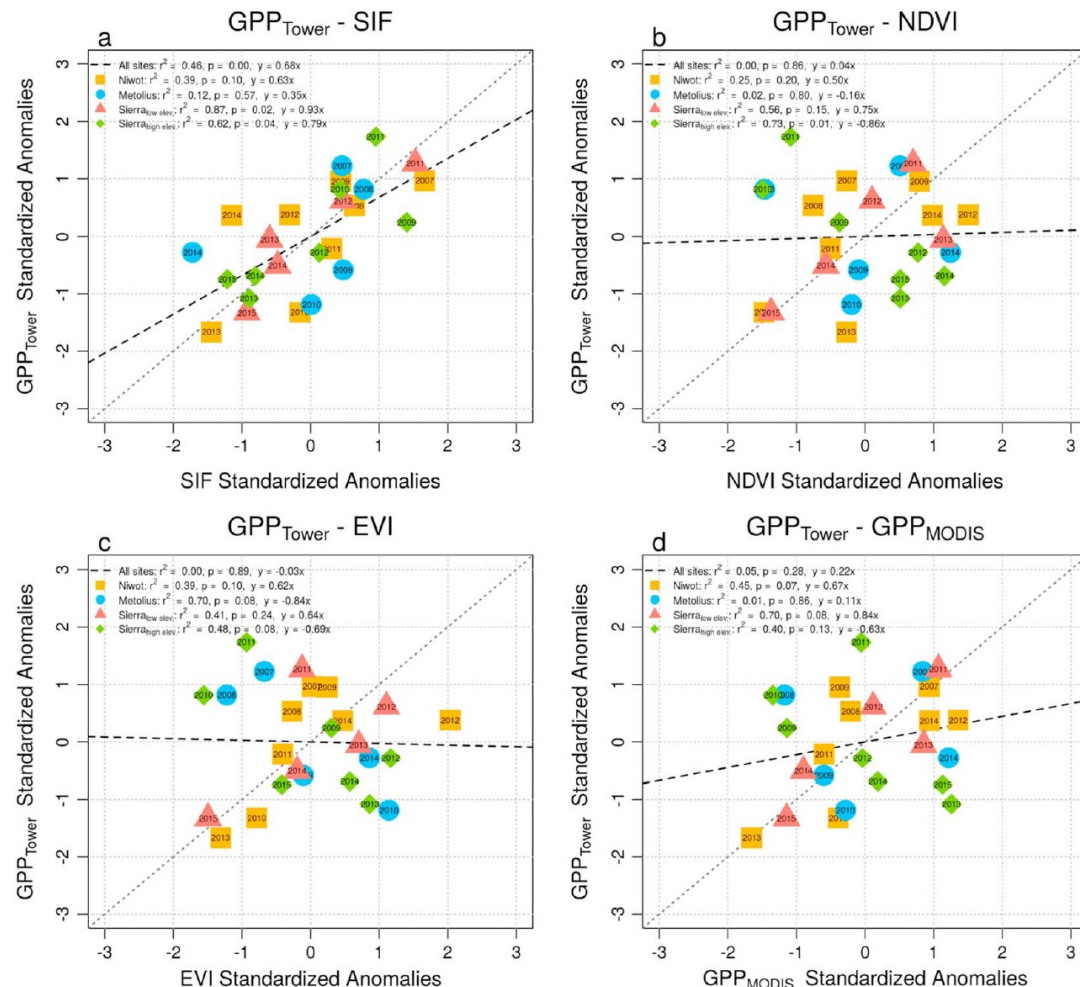


# Demonstration of strong SIF-GPP relationship for Western US evergreen species

## Solar-Induced Fluorescence Detects Interannual Variation in Gross Primary Production of Coniferous Forests in the Western United States

Geophysical Research Letters












Lauren M. Zurowski<sup>1</sup>, David R. Bowling<sup>1,2</sup>, Philipp Köhler<sup>3</sup>, Christian Frankenberg<sup>3</sup>,  
Michael L. Goulden<sup>4</sup>, Peter D. Blanken<sup>5</sup>, and John C. Lin<sup>1</sup>



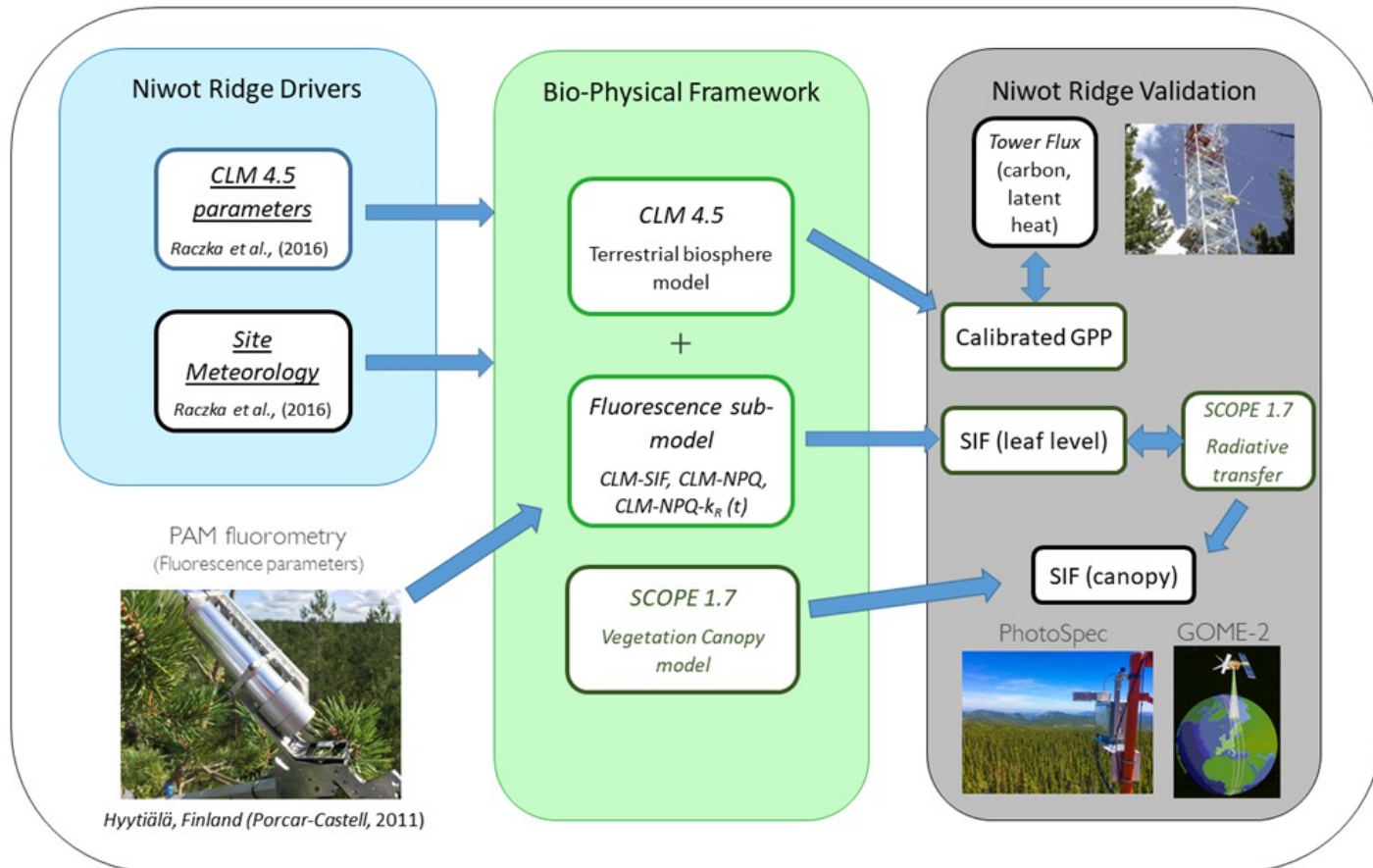
- Solar-induced fluorescence detected inter-annual variation in GPP and small disturbances with greater success than traditional satellite-based products.

# Add representation of SIF within a land surface model: Community Land Model (CLM)

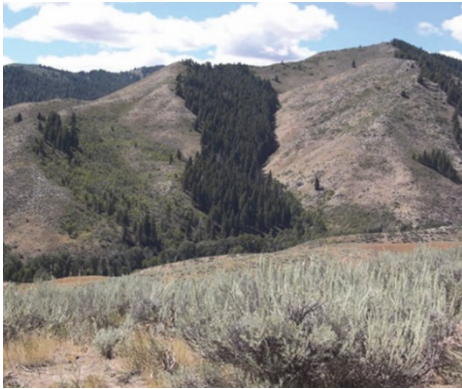
## Sustained Nonphotochemical Quenching Shapes the Seasonal Pattern of Solar-Induced Fluorescence at a High-Elevation Evergreen Forest

Brett Raczka<sup>1</sup> , A. Porcar-Castell<sup>2</sup> , T. Magney<sup>3,4</sup> , J. E. Lee<sup>5</sup> , P. Köhler<sup>4</sup>, C. Frankenberg<sup>3,4</sup> , K. Grossmann<sup>6,7,8</sup>, B. A. Logan<sup>9</sup> , J. Stutz<sup>6,7</sup> , P. D. Blanken<sup>10</sup>, S. P. Burns<sup>10,11</sup> , H. Duarte<sup>12</sup>, X. Yang<sup>13</sup> , J. C. Lin<sup>12</sup> , and D. R. Bowling<sup>1</sup> 

JGR Biogeosciences



# Do more spatially resolved land surface maps and meteorology improve biomass simulations?

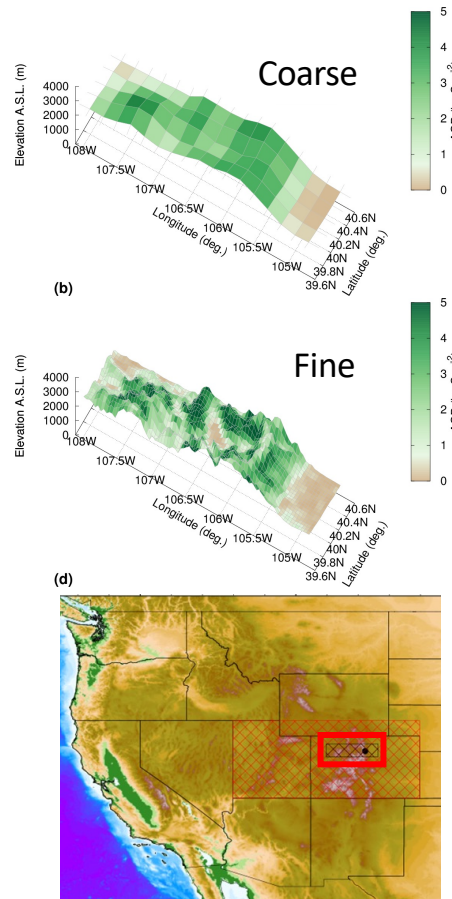


Fan et al., 2019:

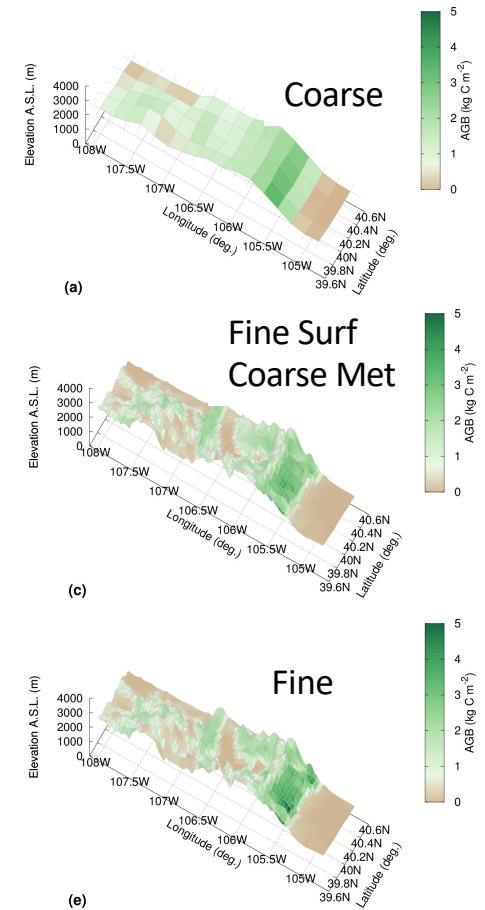
Fine ( $1/24^\circ$ ) and  
Coarse ( $1/2^\circ$ )  
CLM surface  
maps and  
GRIDMET  
meteorology

*"The next questions are as follows: where and when, across the diverse and dynamic environments of the globe, do we expect that these terrain influences will matter to ESM predictions of large-scale water, energy, and biogeochemical fluxes? ...will the hillslope-scale structures, however, deterministic and predictable, simply average out over an ESM grid cell and hence matter little to global predictions?"*

## Observations

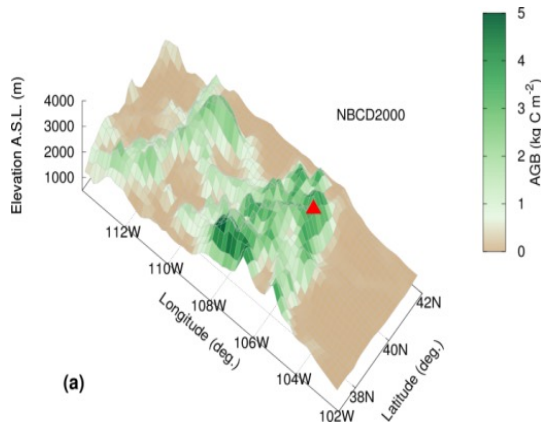
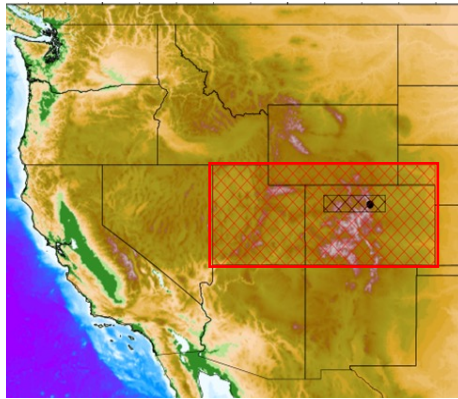


## Simulations



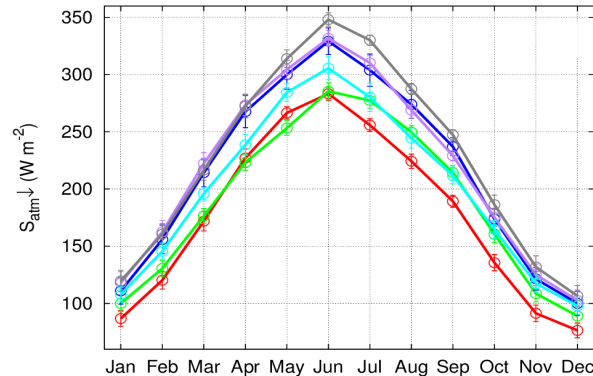


# Meteorology data products for complex terrain tend to be too warm/dry

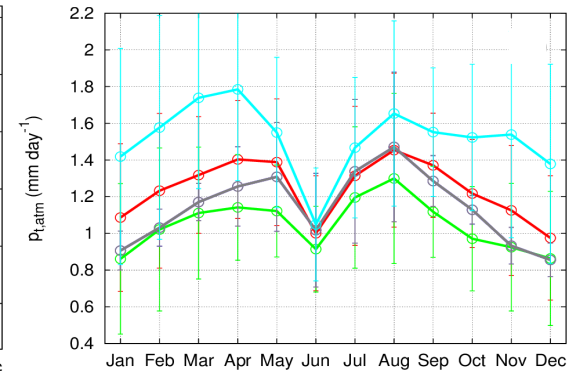


\*  $z > 2235$  m  
75% of total AGB  
(NBCD2000 product,  
Kelindorfer et al., 2013)

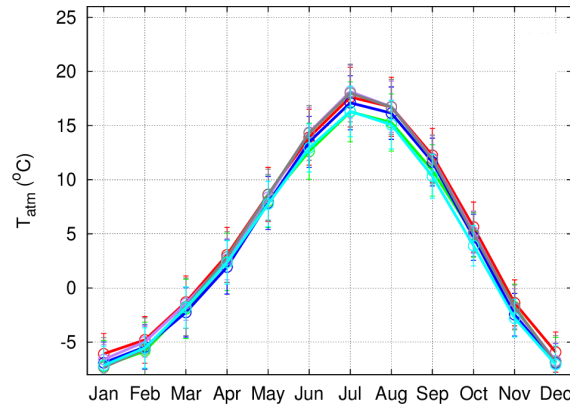
*Incident shortwave radiation*



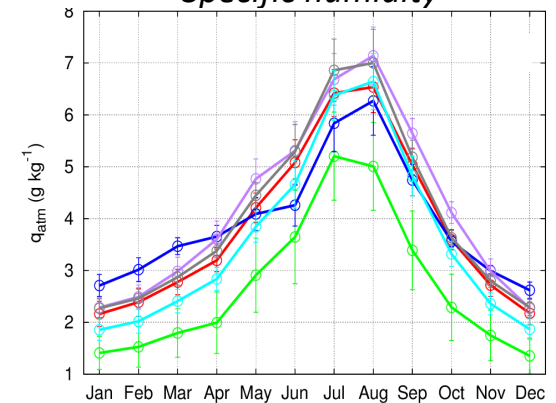
*Precipitation*



*Air temperature*



*Specific humidity*



NARR-MsTMIP (r)

CRU-NCEP

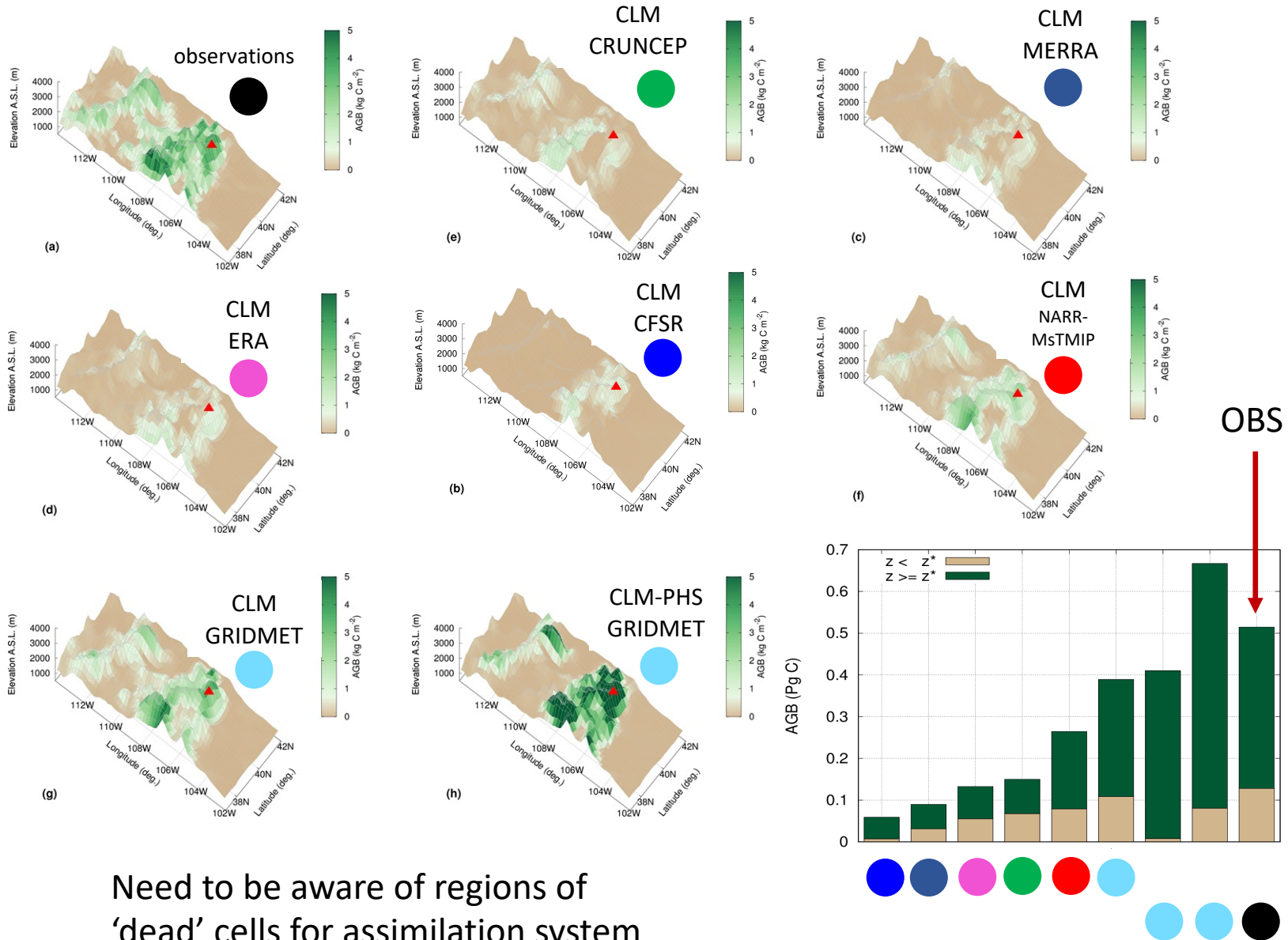
CFSR-Wang

ERA-Wang

MERRA-Wang

GRIDMET-Buotte (r)

# Simulation of biomass is highly sensitive to meteorological biases and representation of water limitation



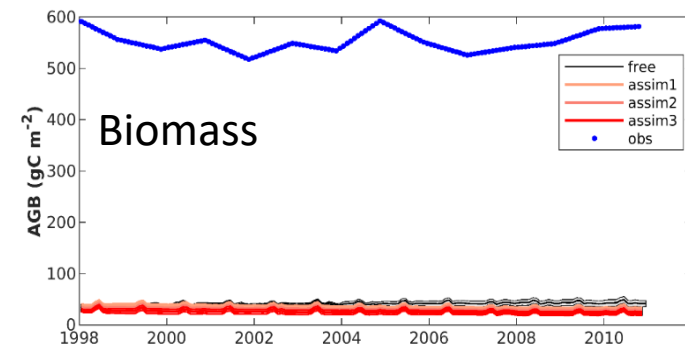
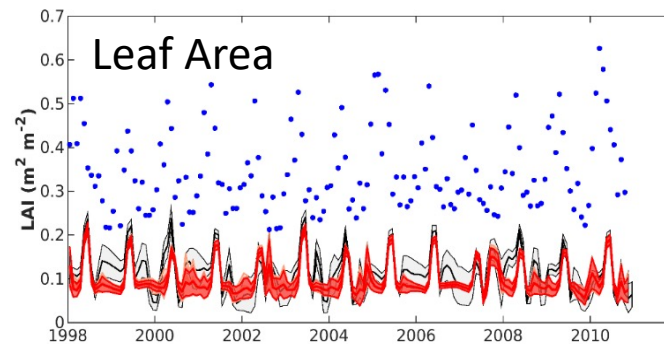
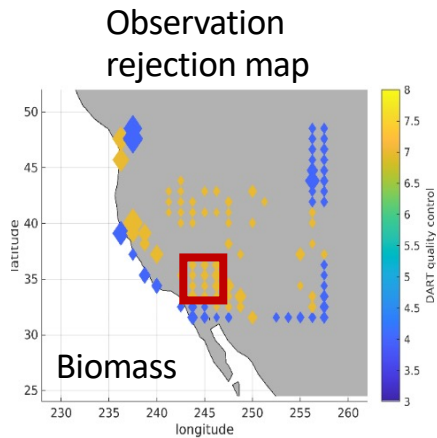
Need to be aware of regions of 'dead' cells for assimilation system

# Dead cell regions inhibit functioning of assimilation system

Observations - Blue

Assimilation (w/obs) - Red

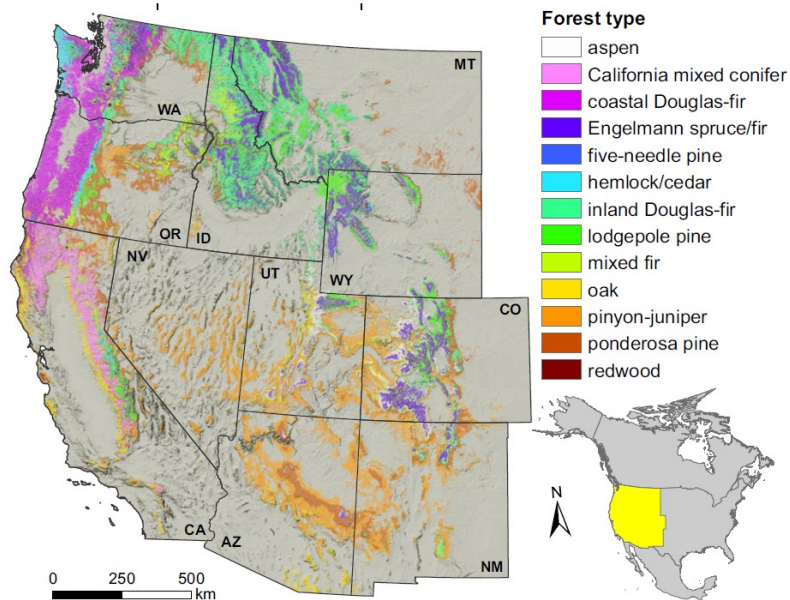
Free (no obs)- Black



Identifying the most favorable meteorological dataset and model configuration (GRIDMET –CLM5-PHS) helped avoid these dead cell regions that are highly resistant to assimilation updates

# Paths for continued land surface model improvement

## Custom PFT parameterization

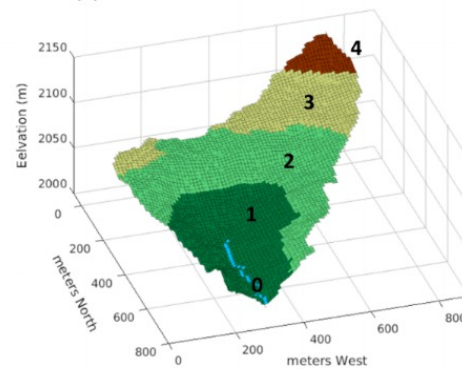


Buotte et al., 2018

## Hillslope, Subsurface Hydrology

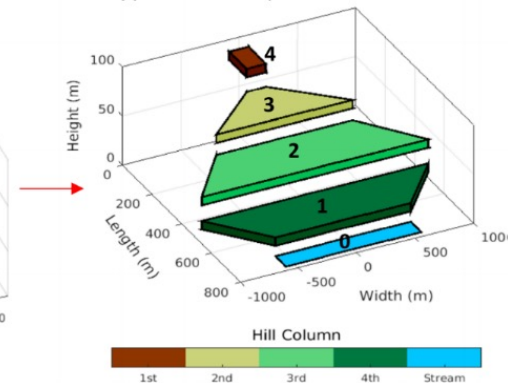
Spatially Explicit  
– computationally unrealistic

(e) HAND bins for a small catchment



Spatially Implicit –  
empirical, but less computation

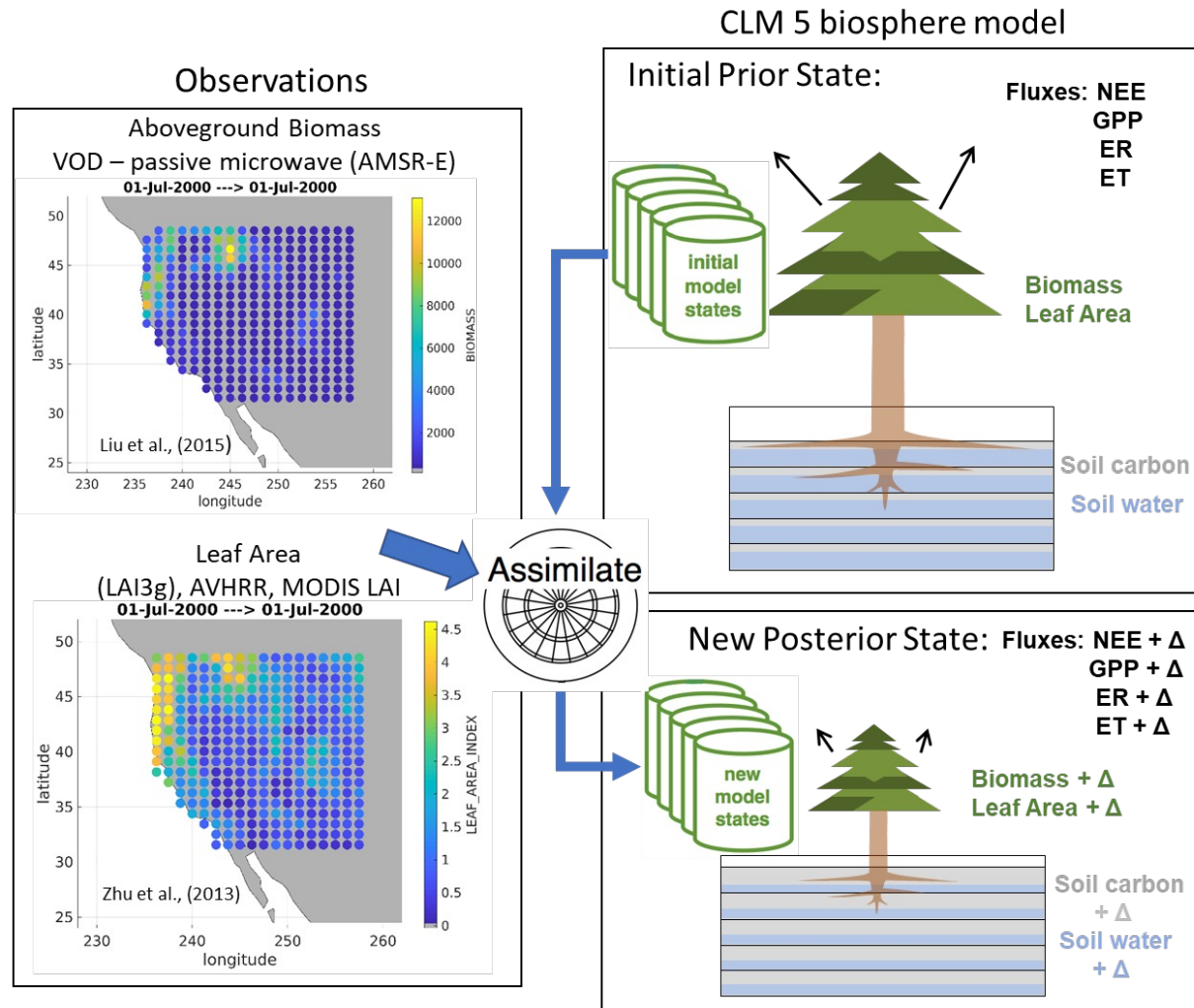
(f) HAND bins represented in model



Fan et al., 2019

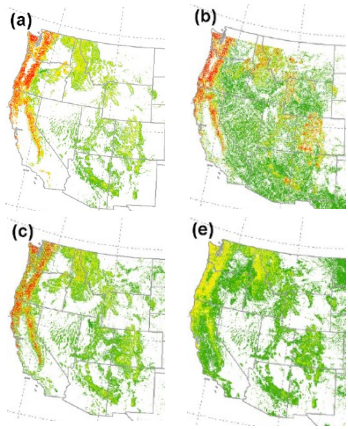


# Land surface data assimilation system: CLM5-DART 'Benchmark Case'

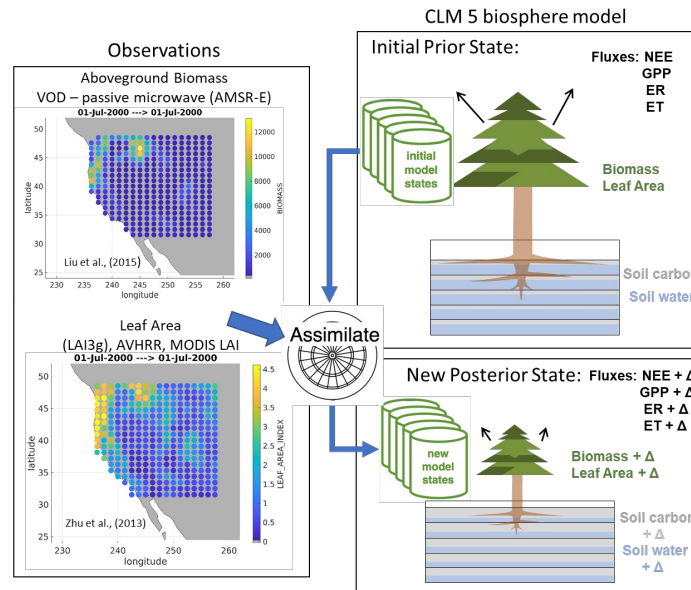
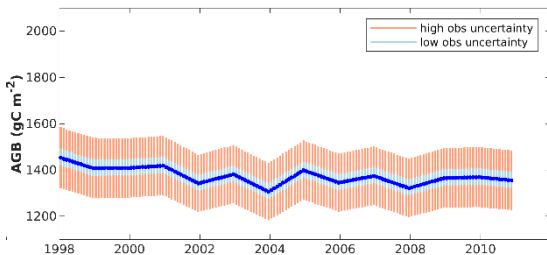


# Land surface data assimilation system: CLM5-DART 'Benchmark Case'

## NASA CMS Biomass WG (biomass products)



## NASA CMS Uncertainty WG

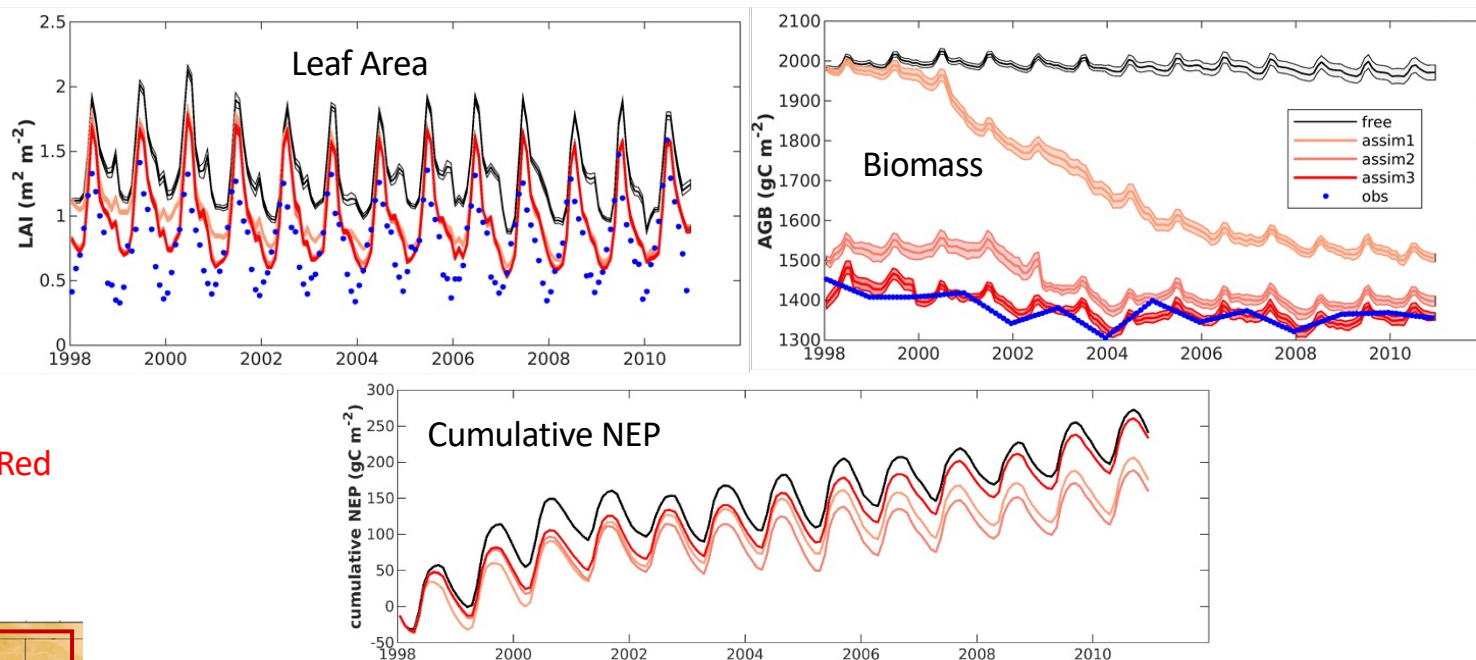


## Model State localization

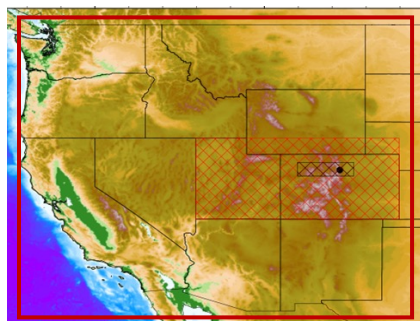
Leaf carbon  
Live stem carbon  
Dead stem carbon  
Leaf area index  
Fine root carbon  
Live coarse root carbon  
Dead coarse root carbon  
Live stem nitrogen  
Dead stem nitrogen  
Litter carbon, slow  
Litter carbon, medium  
Litter carbon, fast  
Litter nitrogen, slow  
Litter nitrogen, medium  
Litter nitrogen, fast

# Assimilation of leaf area and biomass reduce simulated biomass, GPP, ER. Net carbon exchange holds steady

Western US  
Assimilation  
overview



Observations - Blue  
Assimilation (w/obs) - Red  
Free (no obs) - Black



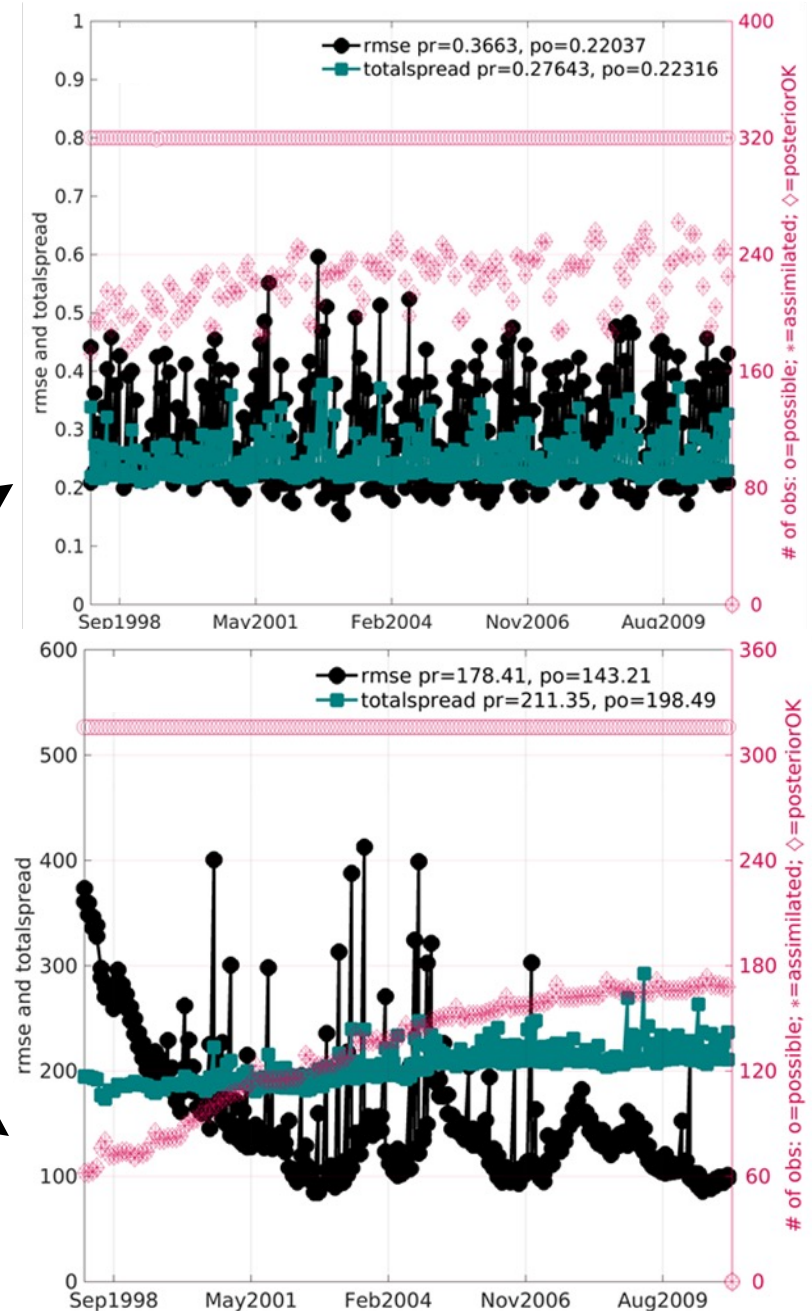
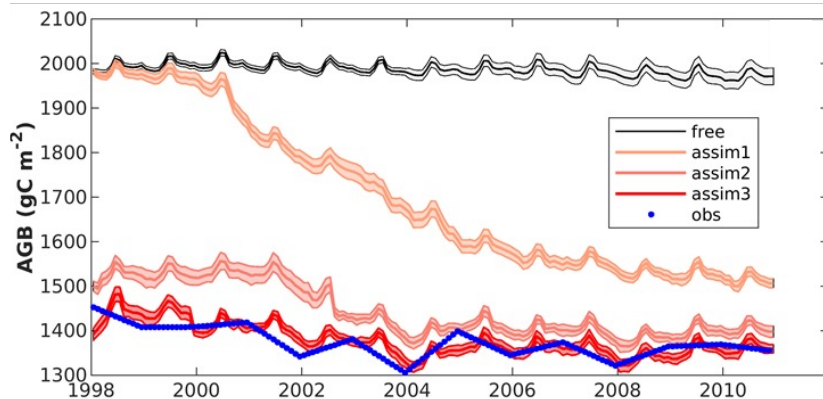
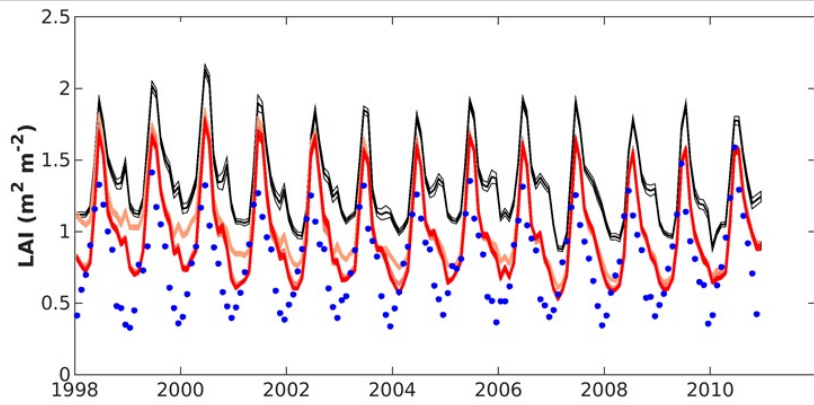
Simulation Name	AGB ( $\text{kgC m}^{-2}$ )	LAI ( $\text{m m}^{-2}$ )	GPP ( $\text{gC m}^{-2} \text{month}^{-1}$ )	ER ( $\text{gC m}^{-2} \text{month}^{-1}$ )	NEP ( $\text{gC m}^{-2} \text{month}^{-1}$ )
<i>Free</i>	1.98	1.31	48.18	47.18	1.00
<i>State-15</i>	1.33	0.93	37.08	39.52	-2.43
<i>State-9</i>	1.36	0.96	38.49	37.21	1.28
<i>State-4</i>	1.44	0.92	37.01	37.15	-0.05

# Was the assimilation successful? Diagnosing observation acceptance rate and RMSE

Observations - Blue

Assimilation (w/obs) - Red

Free (no obs) - Black

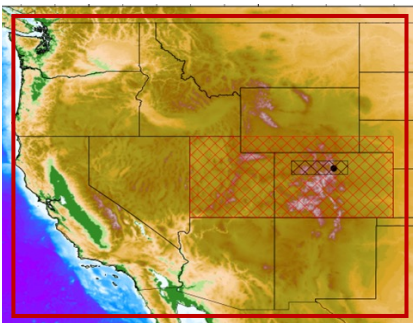




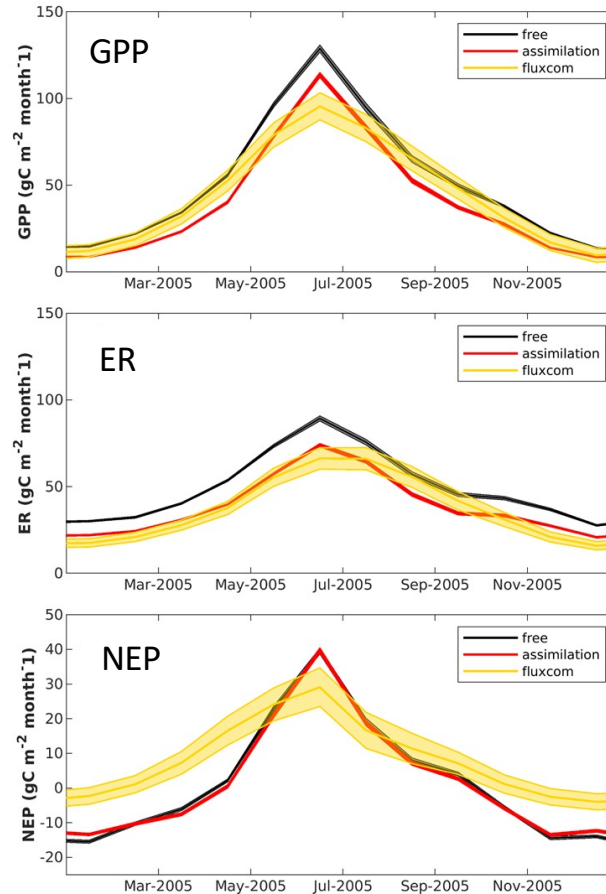
Our assimilation estimate of carbon uptake much weaker than another observation-constrained product (FLUXCOM)

CLM5-DART vs. FLUXCOM  
(observation constrained,  
machine learning, model ensemble)

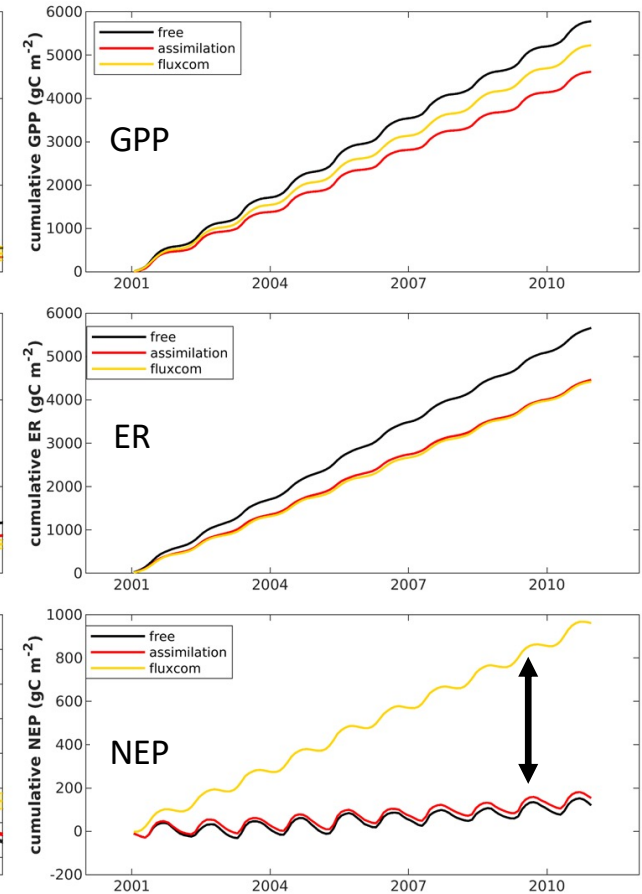
- GPP mismatch
- ER very similar
- NEP: strong winter mismatch



### Seasonal fluxes



### Cumulative fluxes



# Our assimilation estimate of carbon uptake much weaker than another observation-constrained product (FLUXCOM)

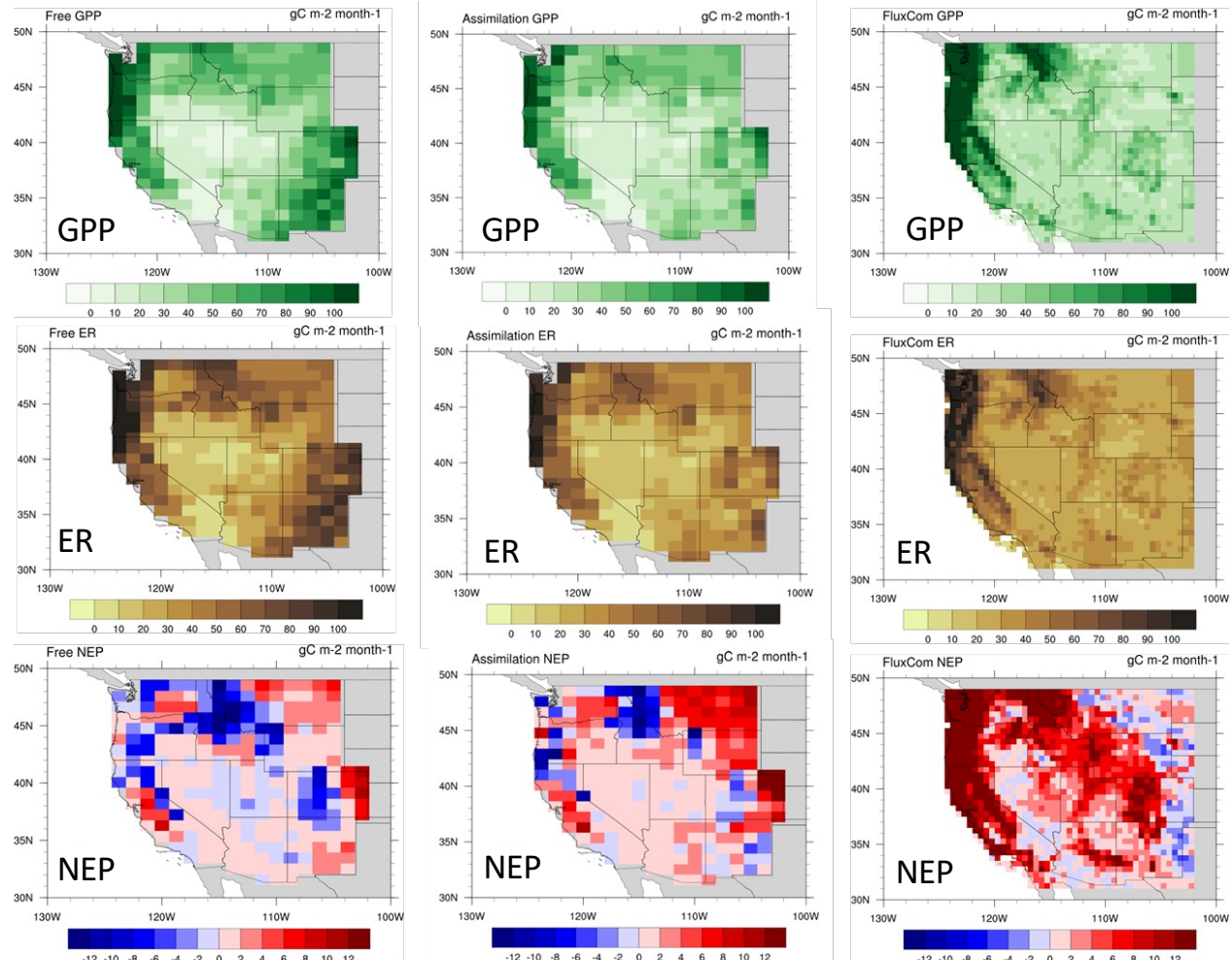
Free Run

Assimilation

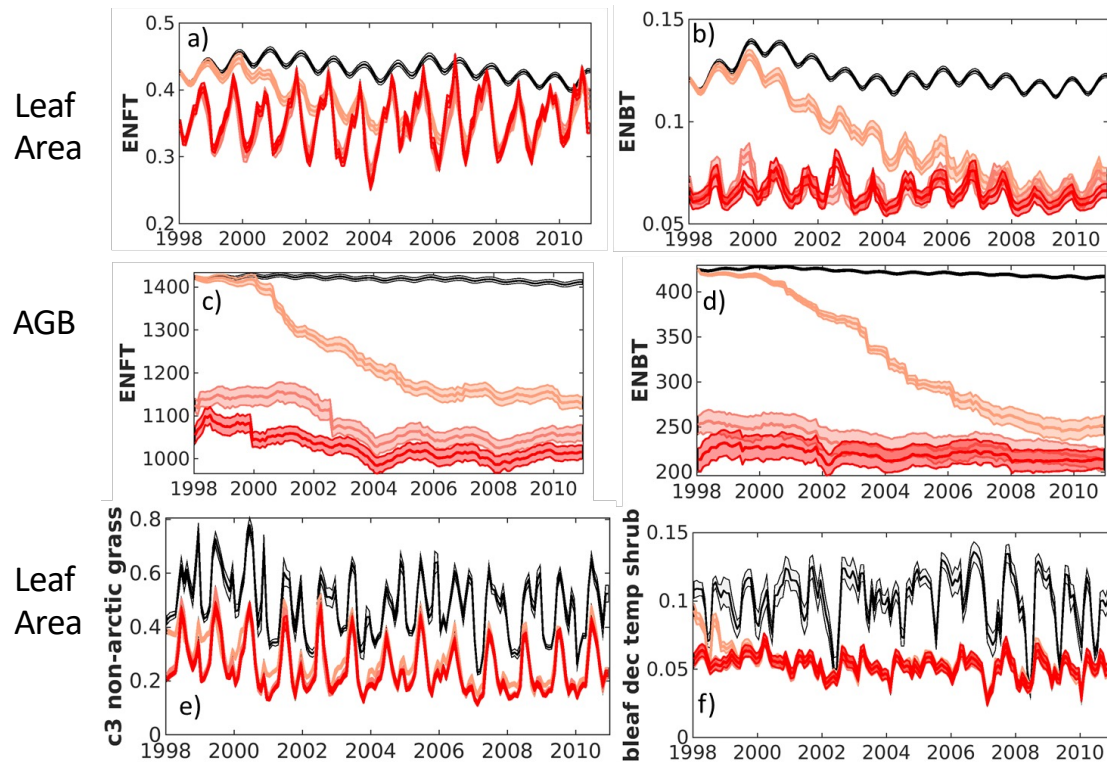
FLUXCOM

CLM5-DART not getting high elevation uptake, low elevation neutral

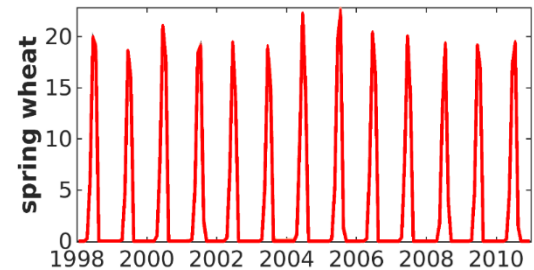
- AGB observations in interior West relatively low
- Water variables in CLM not receiving direct adjustments, (downstream variables)



# Opportunities for improved assimilation: PFT specific observations



The assimilation adjustments to natural vegetation looks fine, crops are resistant to assimilation adjustments

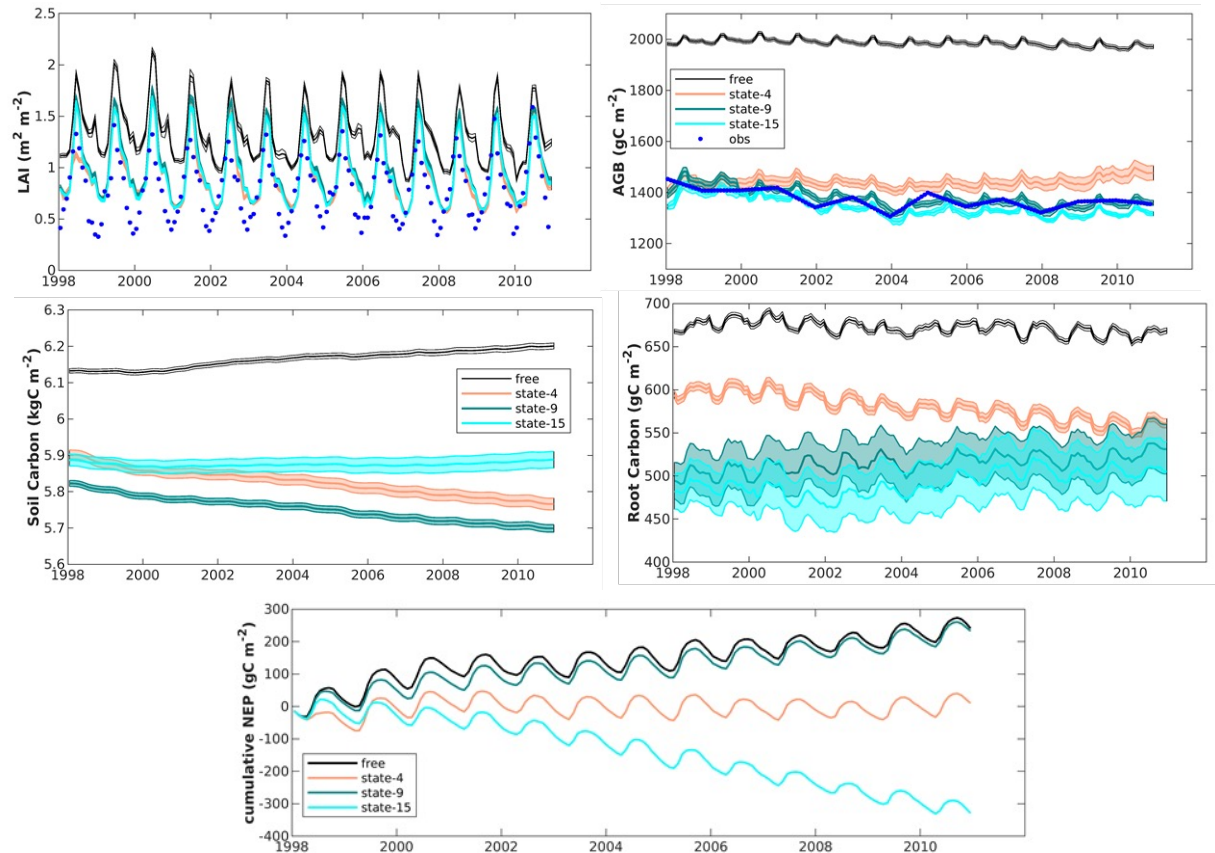


- Robert Kennedy, OSU (LandTrender biomass)

# Opportunities for improved assimilation: expanding CLM adjusted state variables

Leaf carbon  
Live stem carbon  
Dead stem carbon  
Leaf area index  
Fine root carbon  
Live coarse root carbon  
Dead coarse root carbon  
Live stem nitrogen  
Dead stem nitrogen  
Litter carbon, slow  
Litter carbon, medium  
Litter carbon, fast  
Litter nitrogen, slow  
Litter nitrogen, medium  
Litter nitrogen, fast

- Should expand to include soil carbon, water state variables.

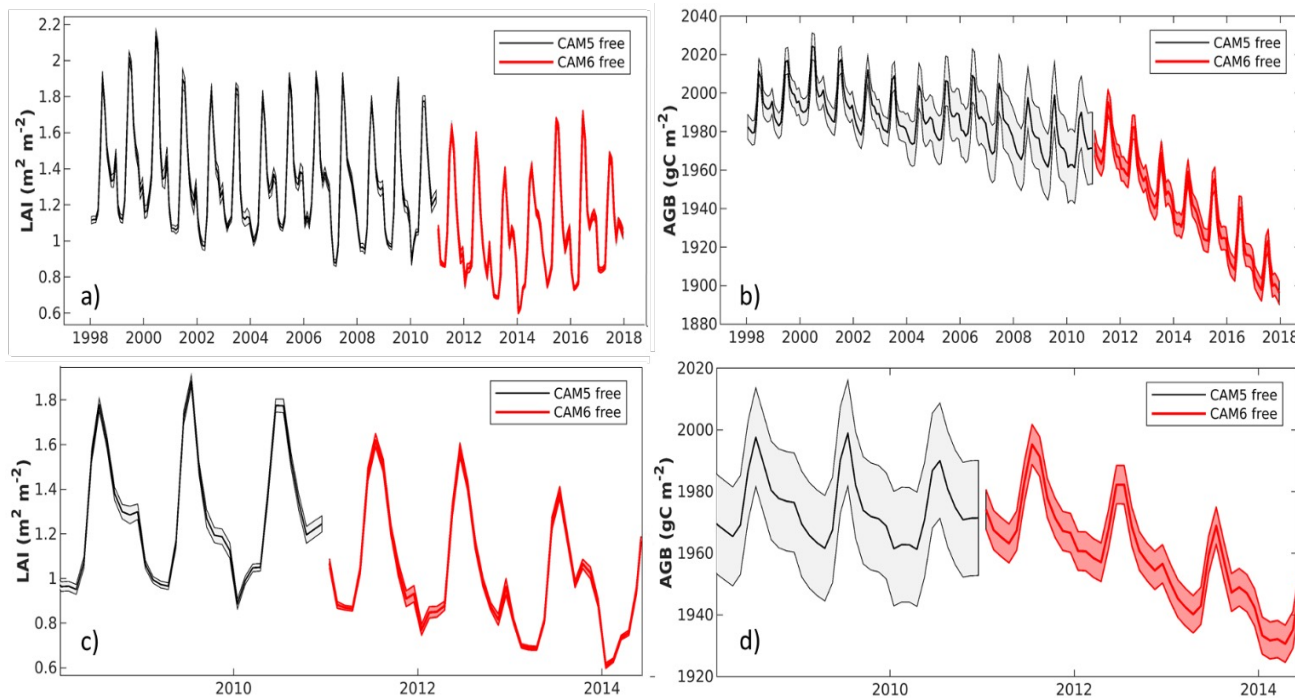




# CMS-Mountains-II

# CMS-II (Lin 2018) advances

- Successfully extended the CLM5 ensemble simulation through 2018 (2019)
- We are poised to add new data streams for 1998-2019 assimilation

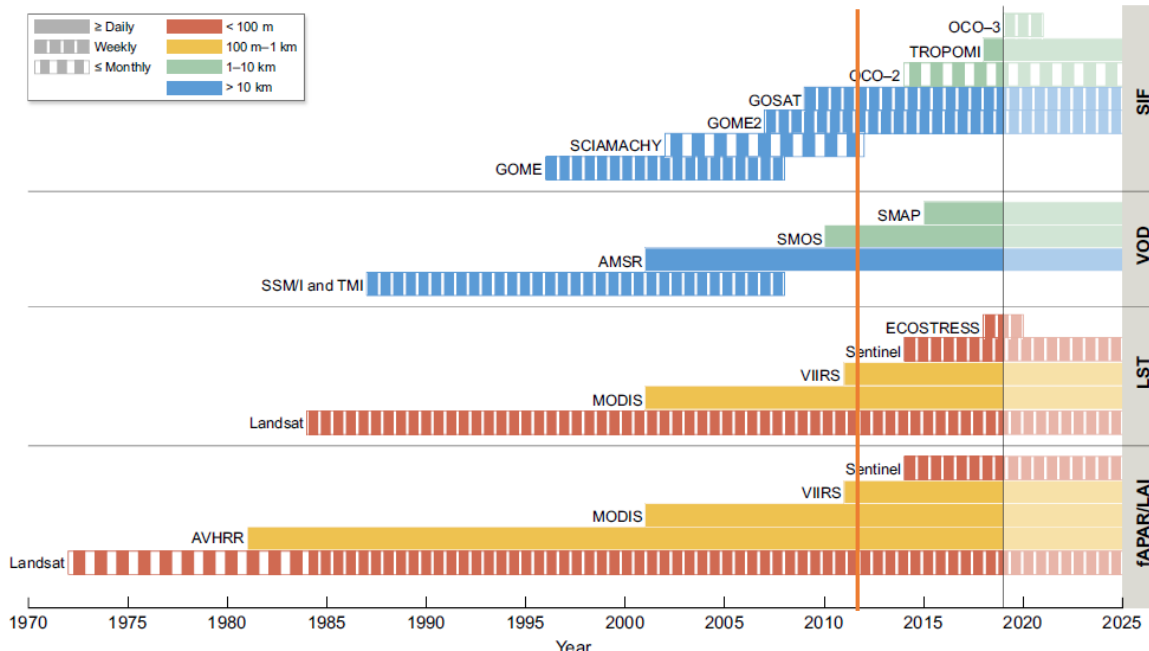


# CMS-II (Lin 2018) goals: add data streams

- Successfully extended the CLM5 ensemble simulation through 2018 (2019)
- We are poised to add new data streams for 1998-2019 assimilation



William K. Smith<sup>1</sup> , Andrew M. Fox<sup>1</sup> , Natasha MacBean<sup>2</sup> , David J. P. Moore<sup>1</sup> and Nicholas C. Parazoo<sup>3</sup>



## Add Observation Streams

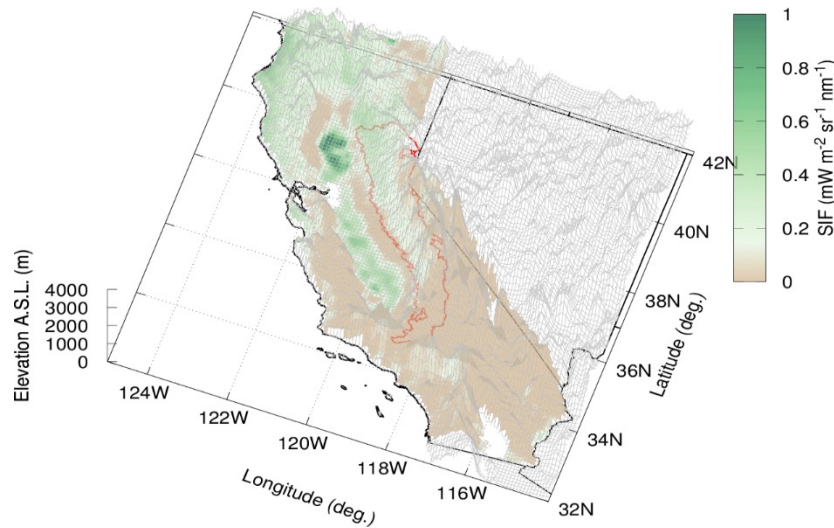
- GLASS LAI
- LandTrendr biomass (PFT)
- ECOSTRESS, LST
- SIF-TROPOMI
- SNODAS

Assimilation framework



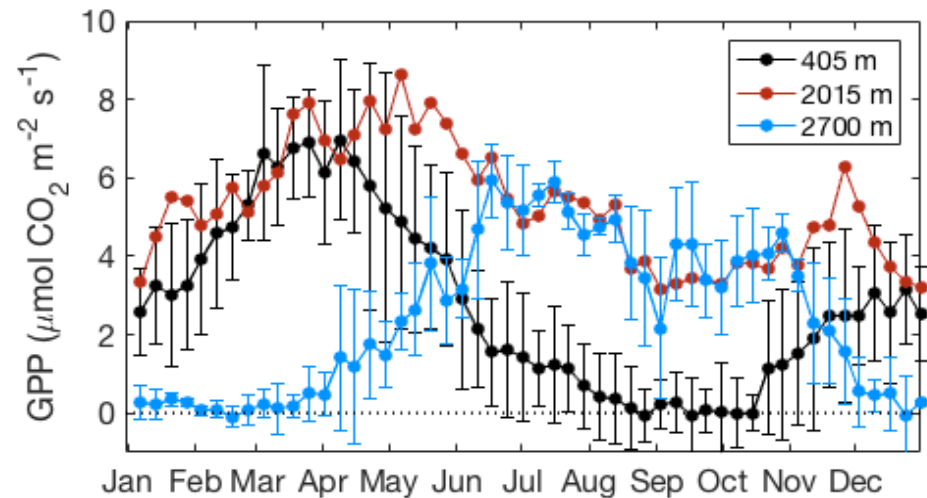
# CMS-II (Lin 2018) goals: High resolution TROPOMI-SIF to diagnose phenology

High Resolution TROPOMI-SIF (5x5 km, 4-8 days)



- Characterize seasonal phenology based on elevation, slope and aspect

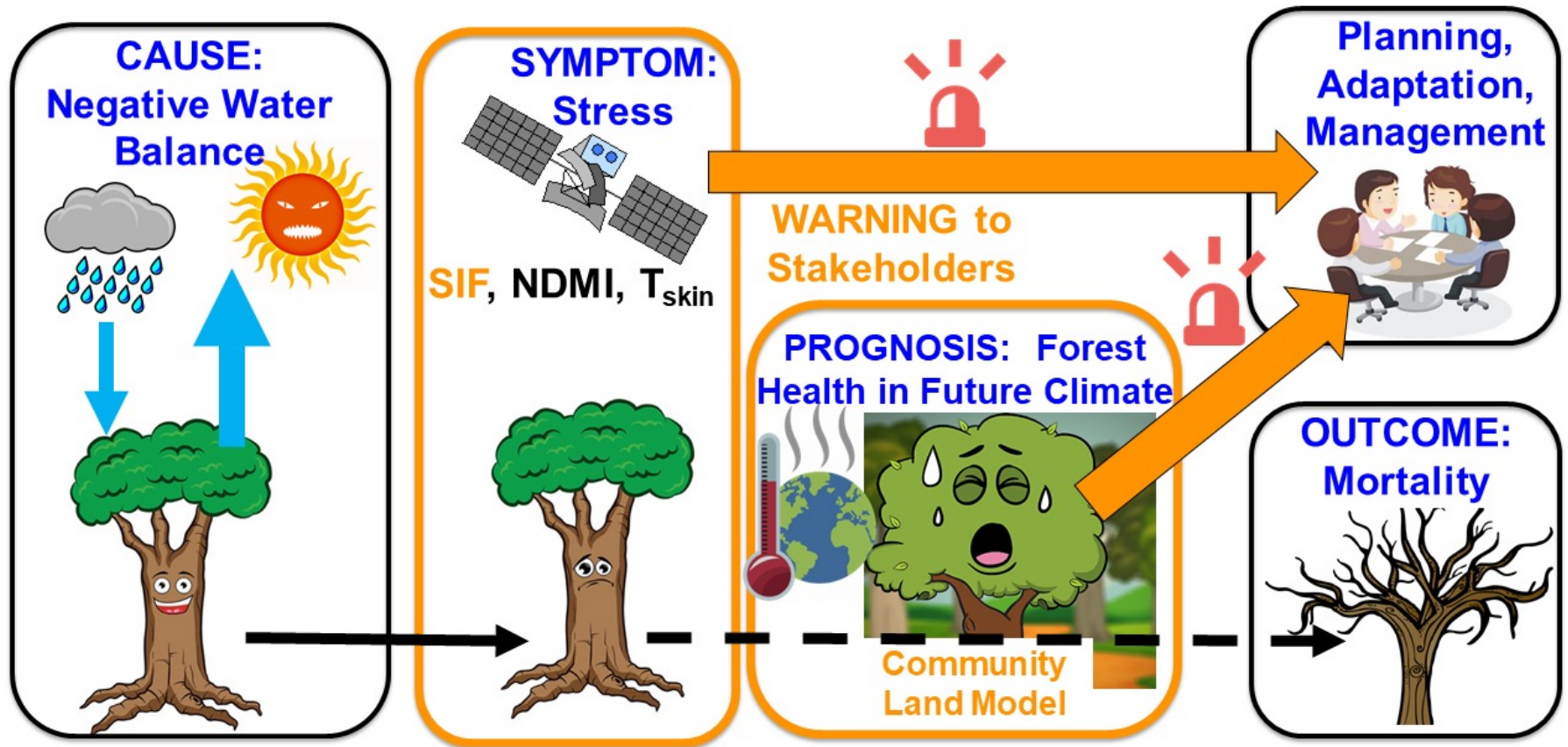
Southern Sierra Critical Zone  
Observatory Flux Tower Sites



- Provide insight into phenological transition across elevation gradient
- Implement this understanding into improved phenological model in CLM 5



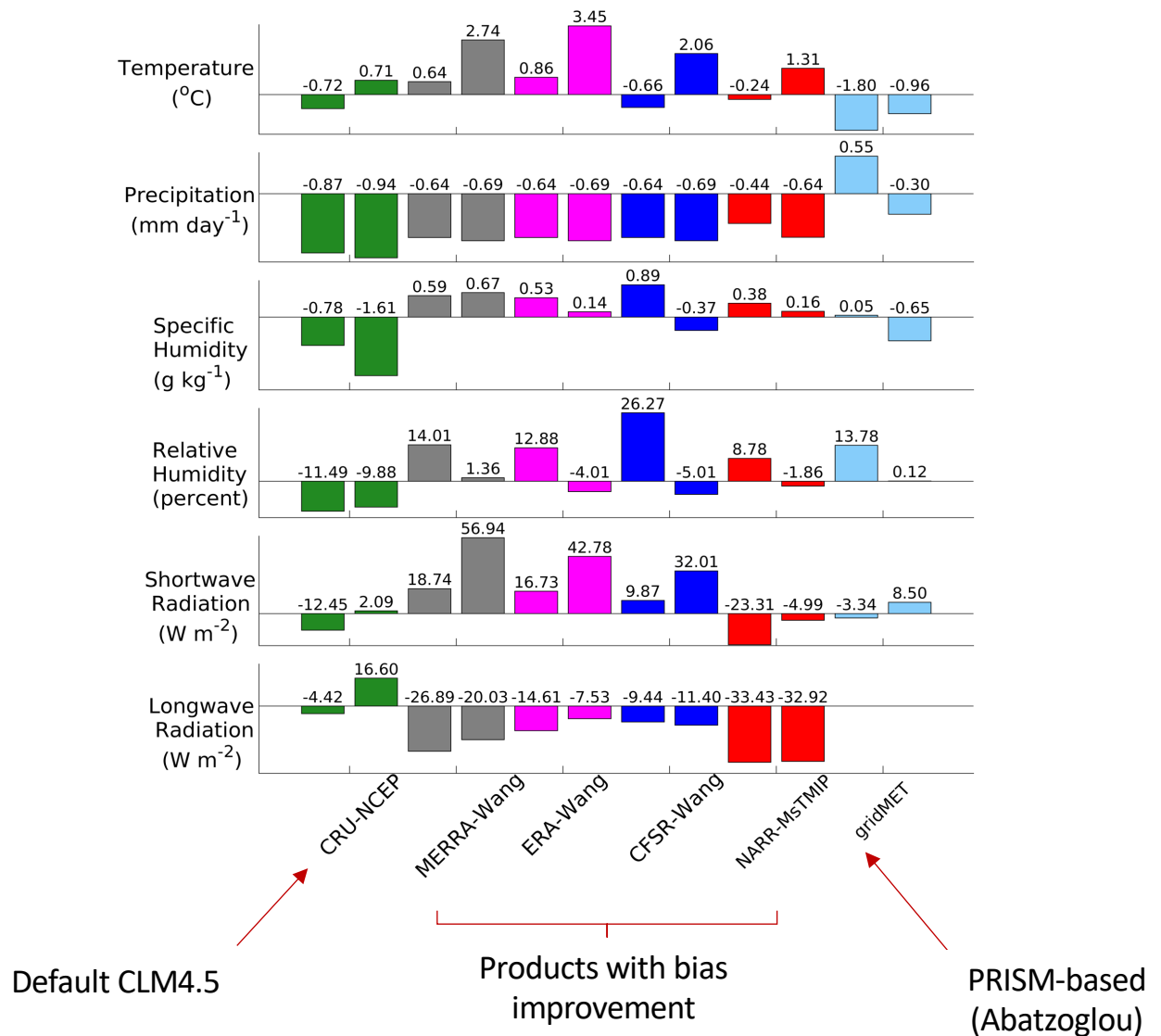
# CMS-II (Lin 2018) goals: Forest Health Early Warning System



# Questions?

# Meteorological datasets tend to be too warm/dry across Western US

- Meteorological biases at Niwot Ridge, Colorado
- High temp
- High SW radiation
- Low precip
- Asking for trouble within a water limited region



# Simulation of biomass is highly sensitive to meteorological biases and representation of water limitation

