

Ecohydrology modeling - example

**Naomi
(Christina)
Tague**

**University of California
at Santa Barbara**



Forest in Mediterranean Type Ecosystems (MTEs) (winter wet, summer dry) are changing



*Increased fire severity

*Increased forest mortality

*Changes in forest growth, health, carbon sequestration

*Changes in hydrology: floods and droughts

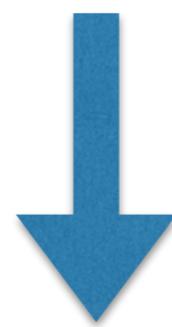
<https://www.climateassessment.ca.gov/>

Key Questions

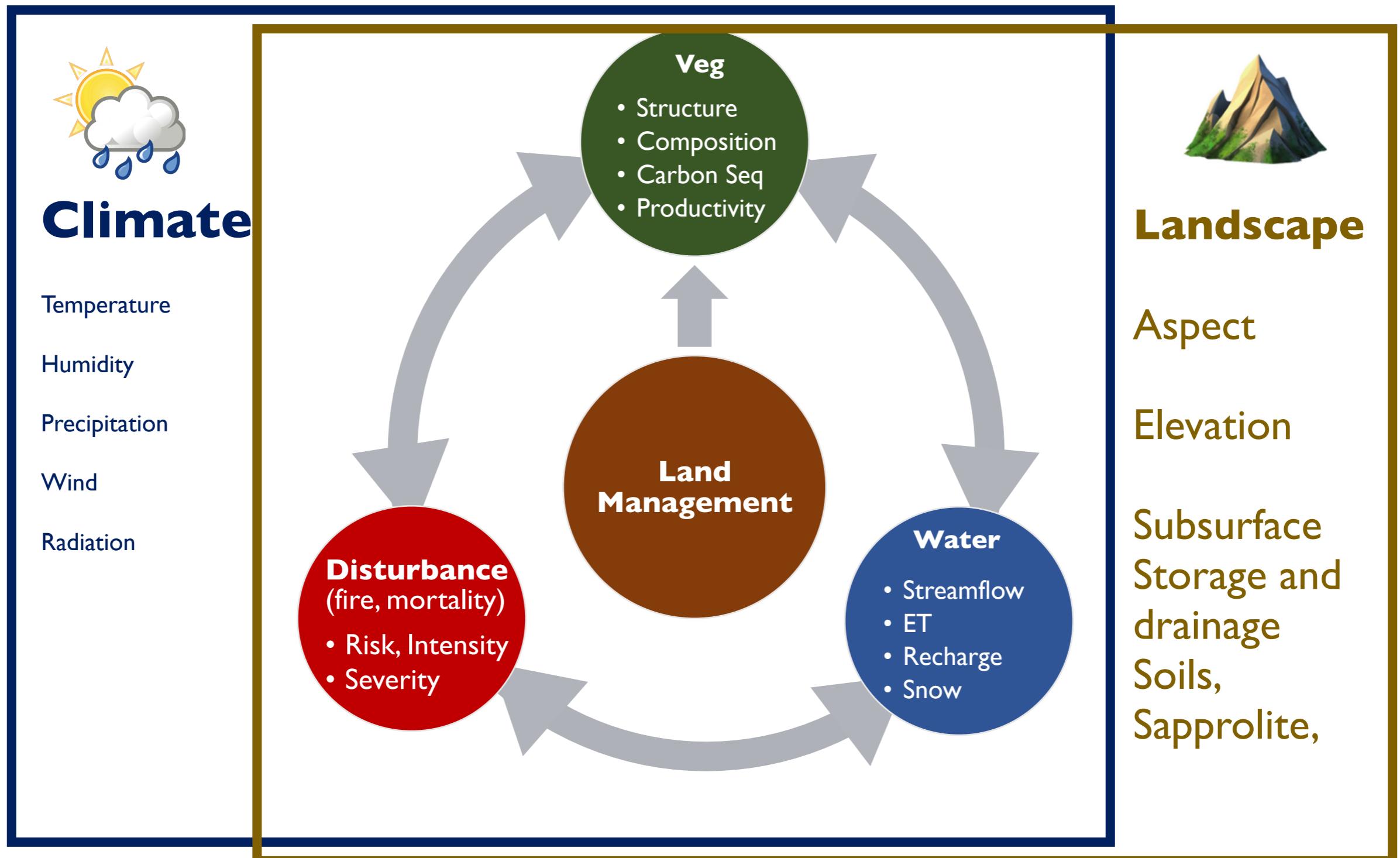


*What will future forests, water resources look like? Next decade..beyond

*What role can management, mitigation play?

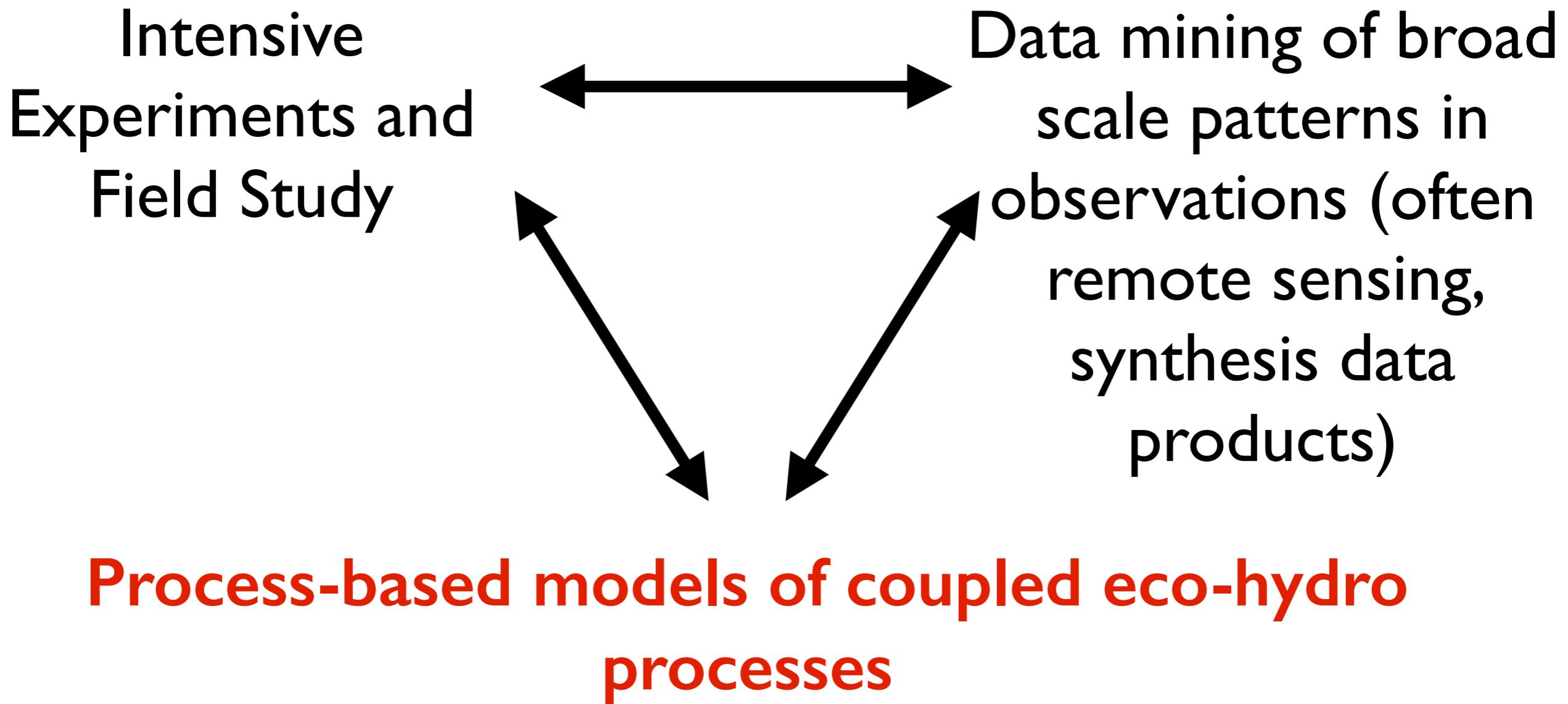


Answers are challenging because...

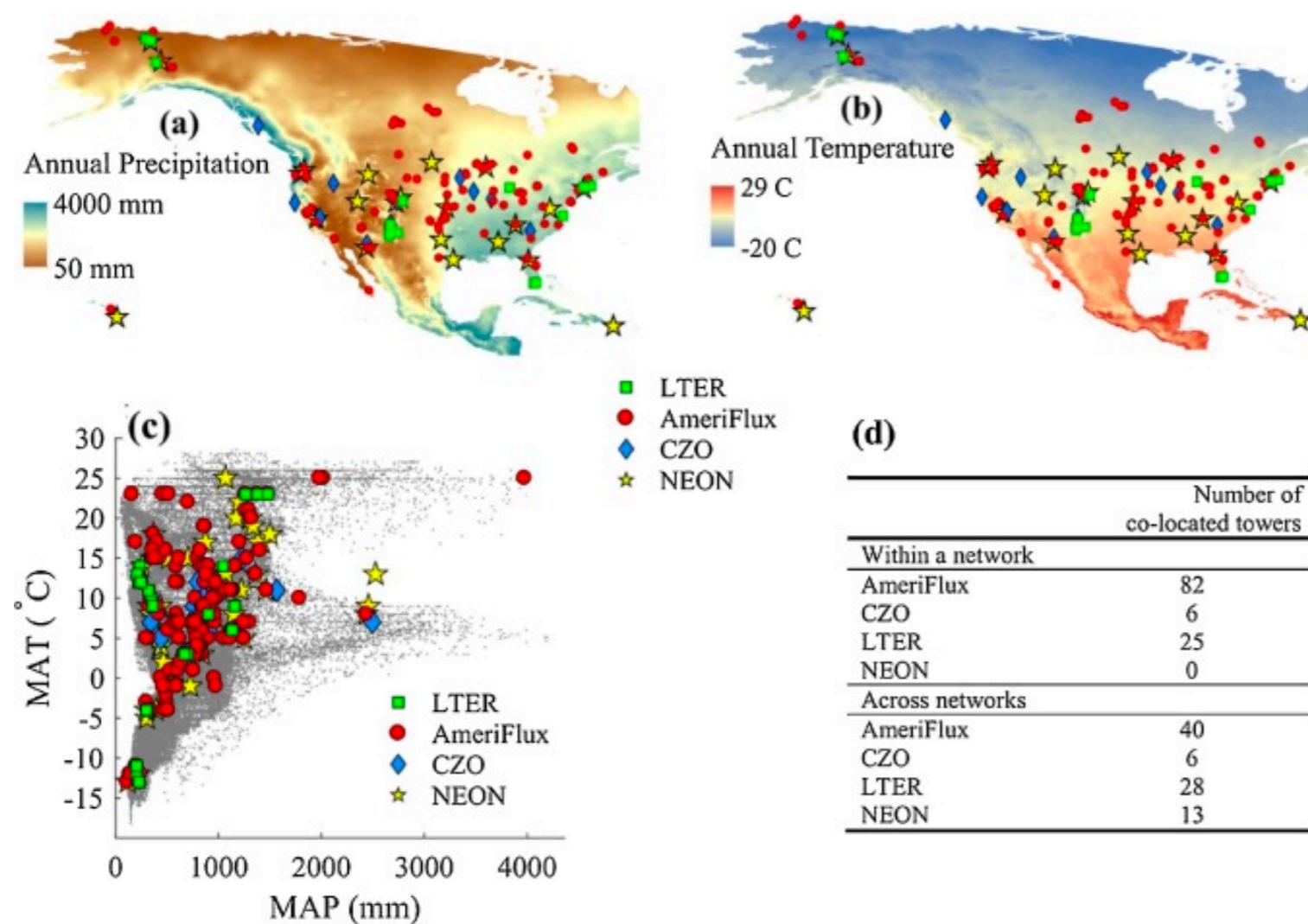


Processes are tightly coupled...highly heterogeneous
in space and vary in time (seasons...years)

The “tools” for the job

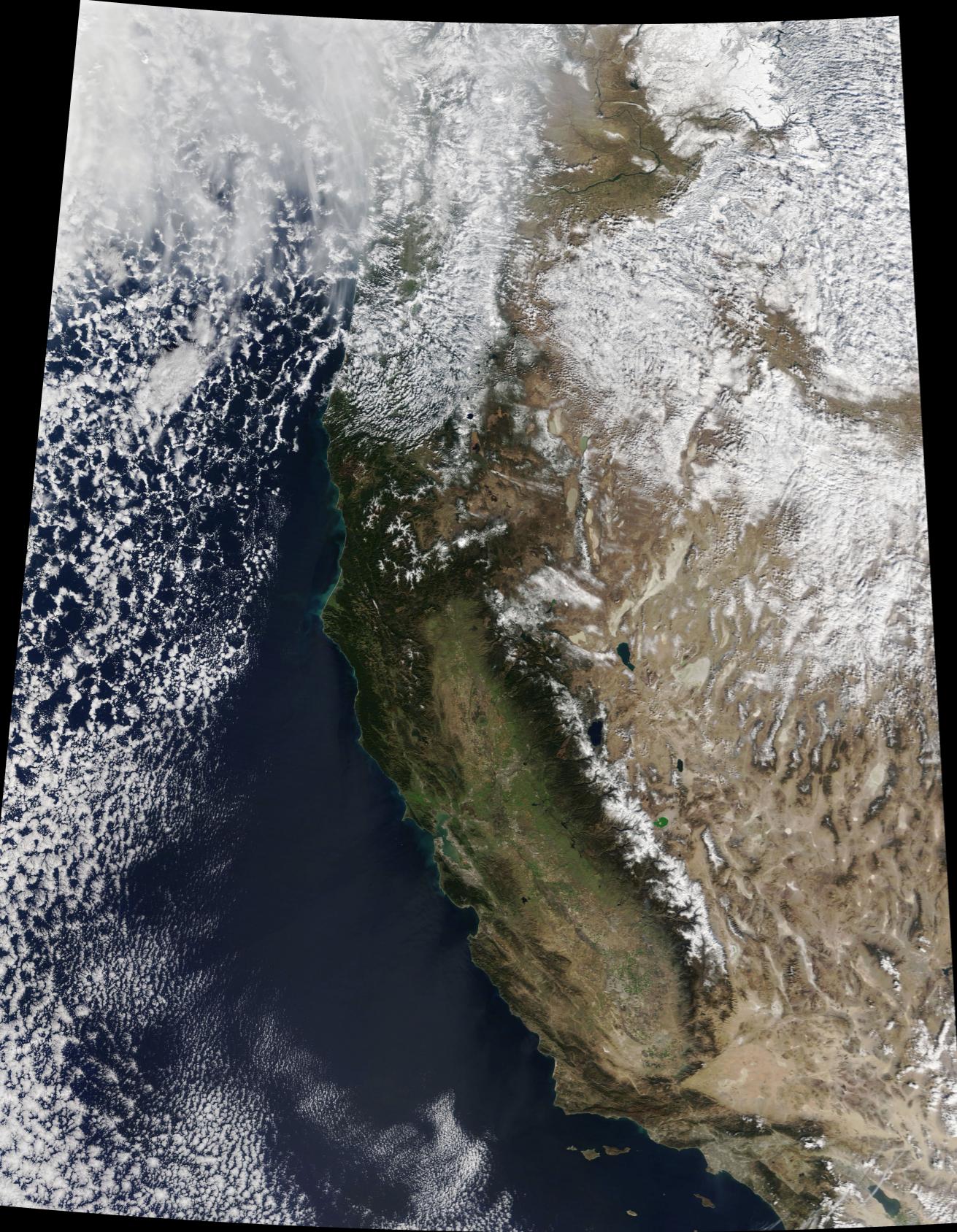


Networks of collaboration...site scale measurements (Example: Carbon flux towers)



NSF Observatory Networks

LTER - long term ecological research
CZO - critical zone observatory
NEON - national ecological observatory network
AmeriFlux



Remote sensing....snow,
veg....

NASA

Emerging sources...

USGS and other LiDar sources

(<https://prd-tnm.s3.amazonaws.com/LidarExplorer/index.html>)

Planet (<https://www.planet.com/>)

California Observatory

(<https://salo.ai/projects/california-forest-observatory>)

What is a “process-based” model

Encodes theories about ecohydrologic function
mechanisms of “how stuff works”

Example - process based model of leaf-scale transpiration combines

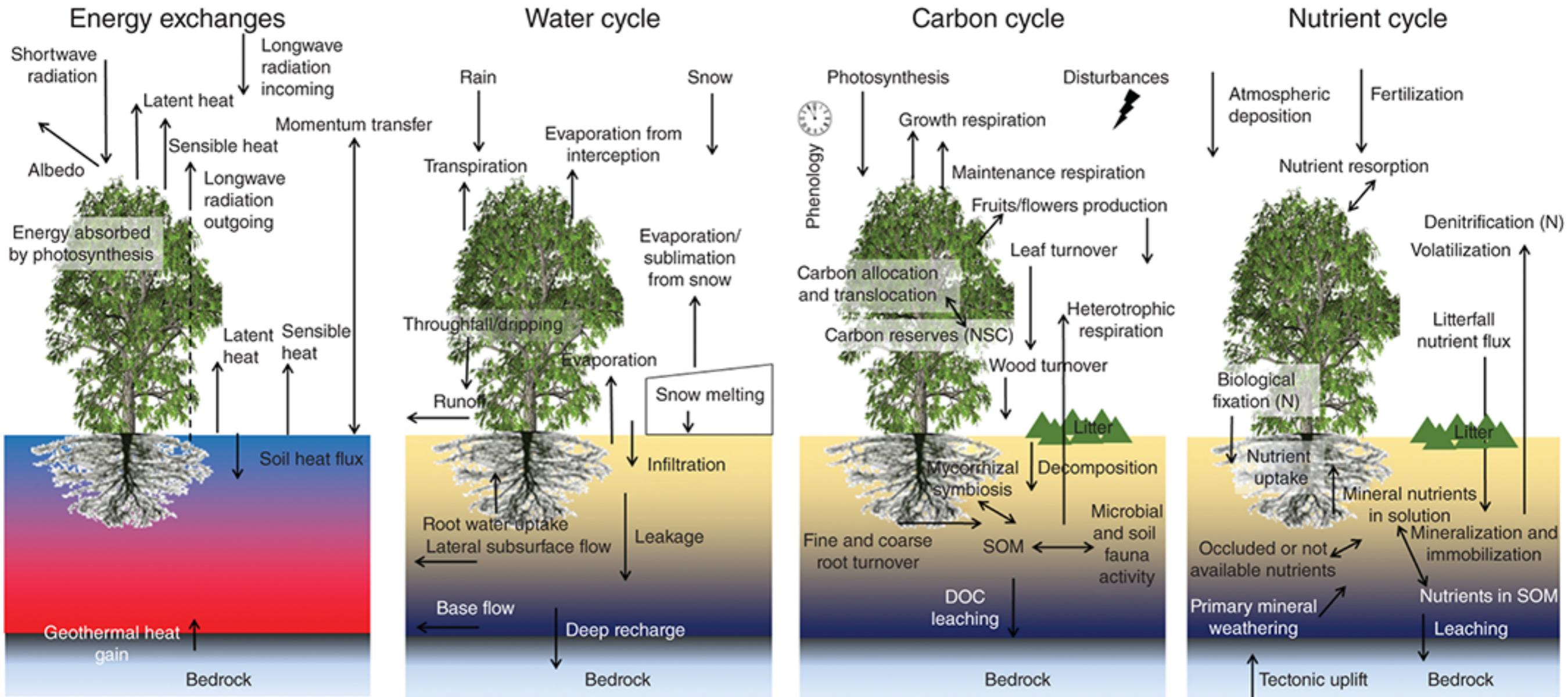
- Radiation absorption - leaf area
- Aerodynamic properties - vegetation height, wind, VPD
- Stomatal Function - active regulating water loss



Penman-Monteith

$$E = \frac{s \underline{R_N} + \rho_a c_p \underline{Cat} u [e_s(T_a) - e_a]}{[s + \gamma(1 + \underline{Cat/Ccan})] \lambda_v}$$

Review of flow and grow models

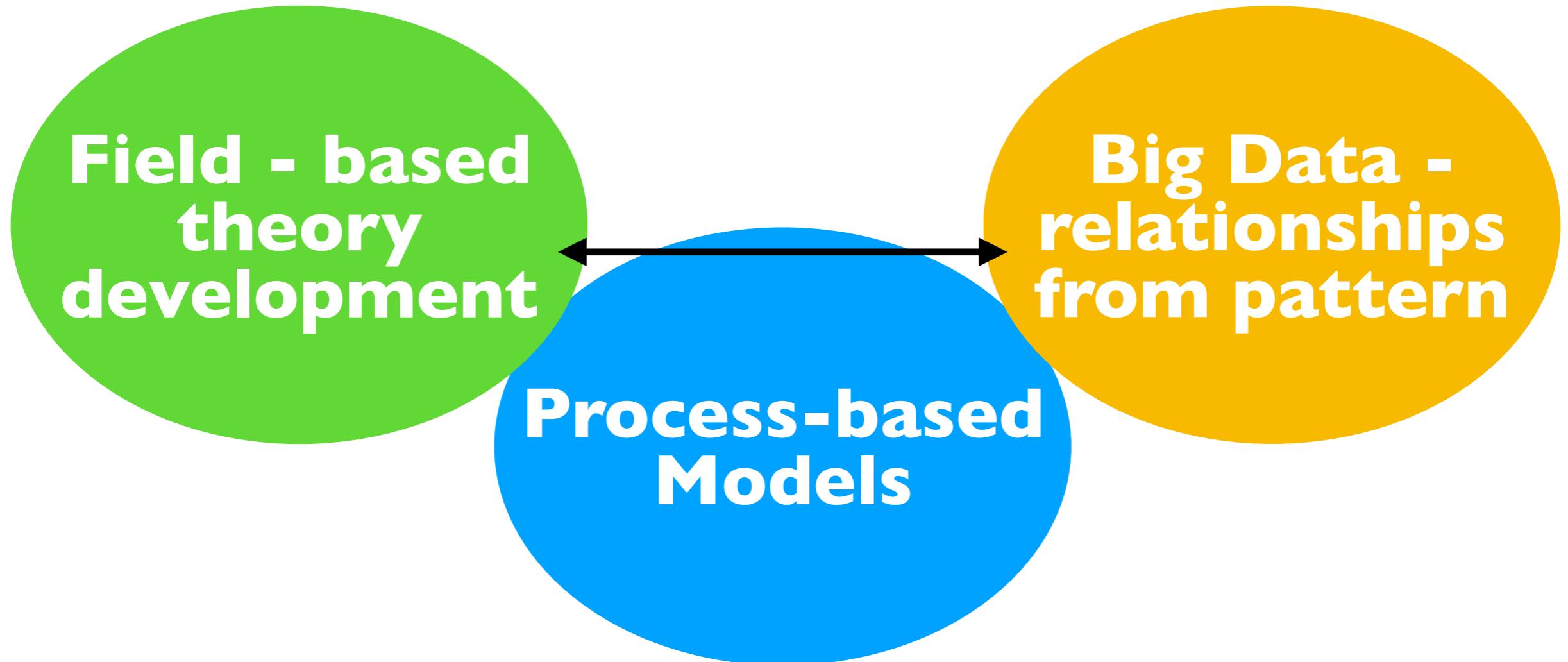


Wiley Interdisciplinary Reviews: Water

Volume 3, Issue 3, pages 327-368, 19 NOV 2015 DOI: 10.1002/wat2.1125

<http://onlinelibrary.wiley.com/doi/10.1002/wat2.1125/full#wat21125-fig-0006>

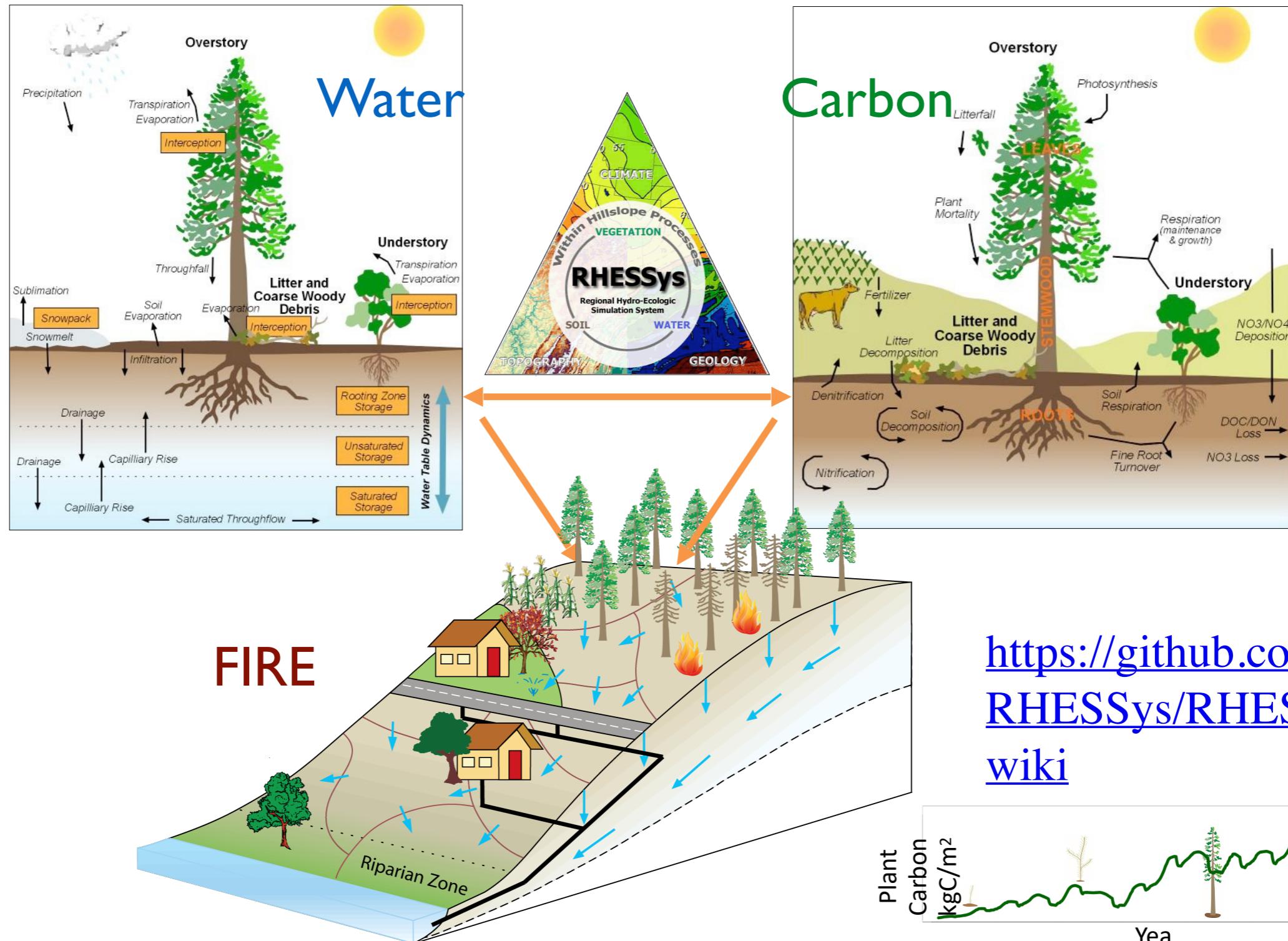
Advances in ecohydrology require integrating...



- what we are learning from field-investigations
- what we learn from patterns from big data

Process-based models can help us to do that

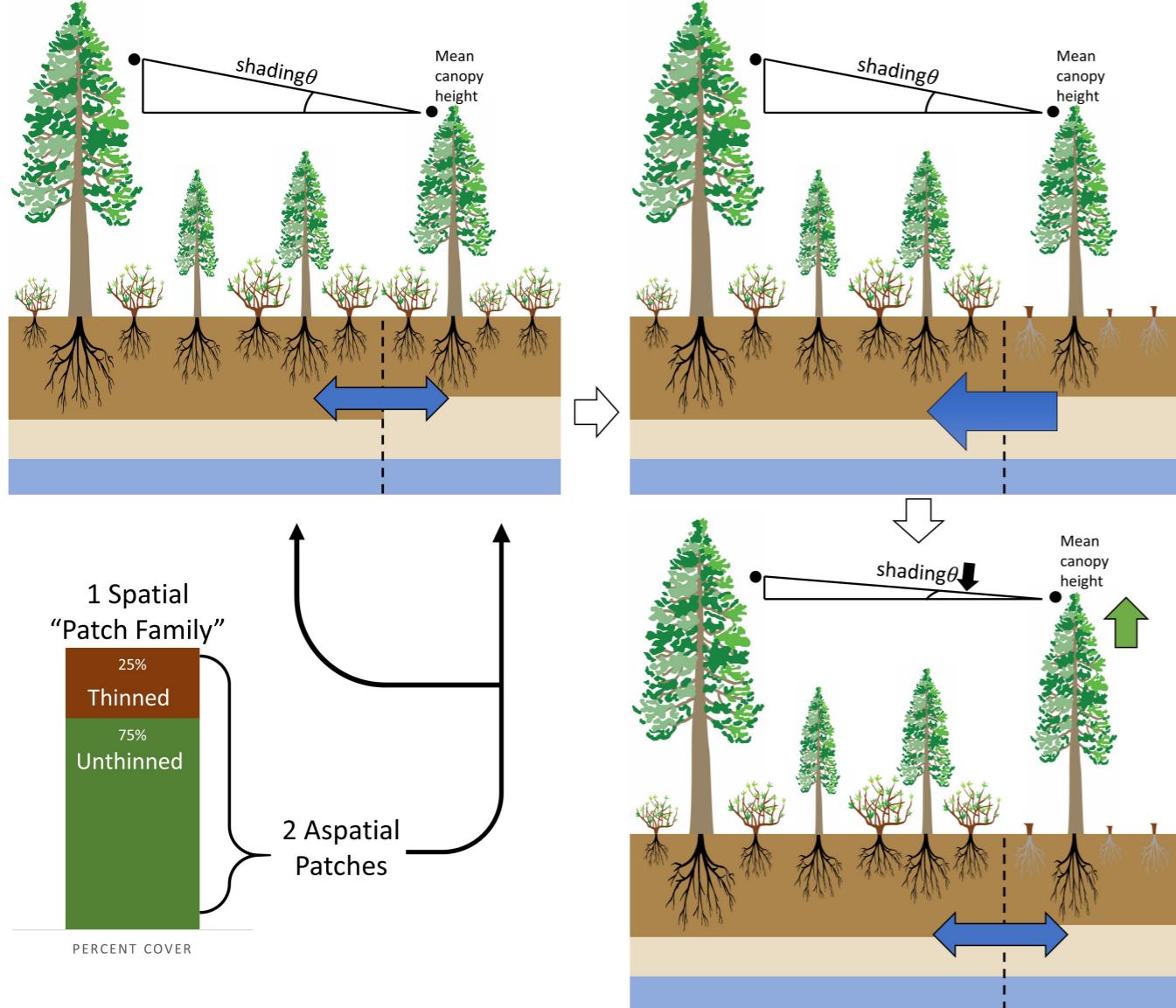
RHESSys process model does both space-time



[https://github.com/
RHESSys/RHESSys/
wiki](https://github.com/RHESSys/RHESSys/wiki)

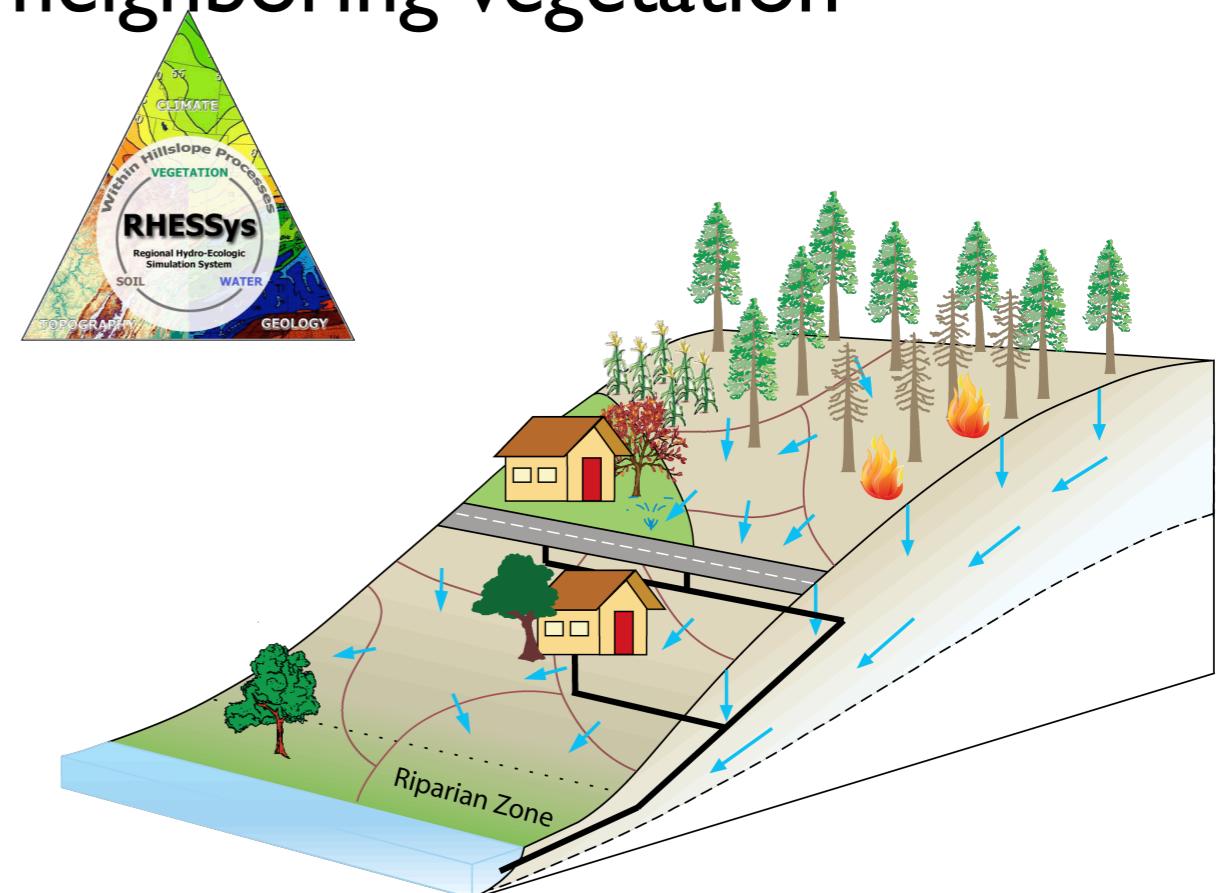
Models help to synthesize these data sets...and create 'best guess'
and 'plausible' scenarios of where things are going;

Within patch patch family - Storage is “shared”



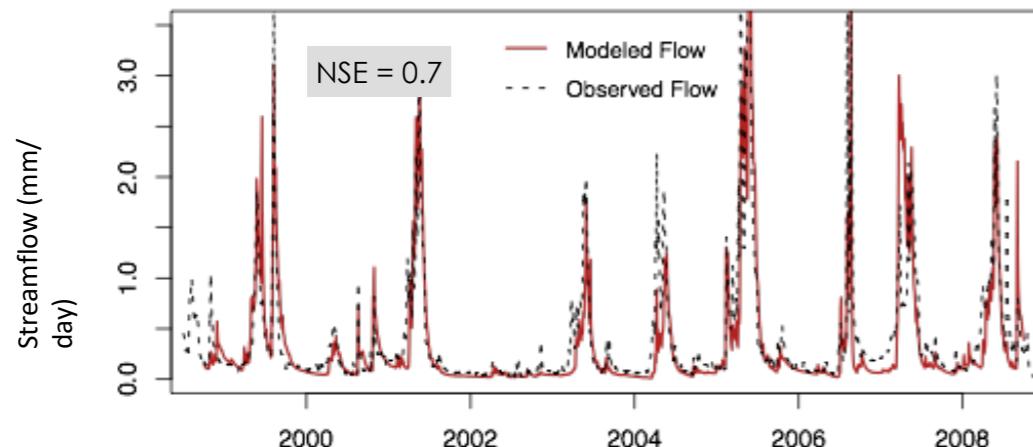
Within Plot/Stand

Multi-Scale RHESSys -
allows us to resolve the impact
of vegetation removal on
neighboring vegetation

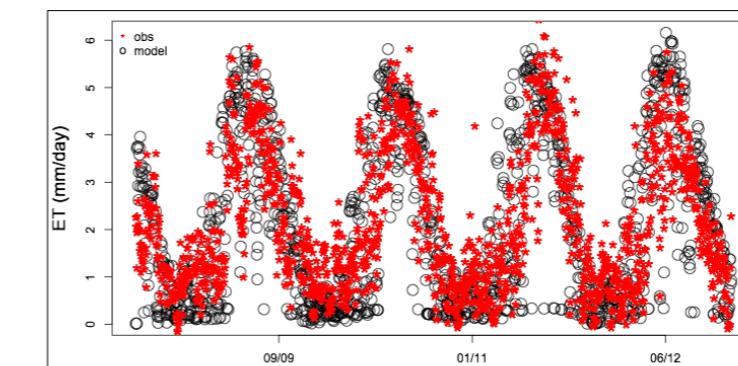


Between Plot/Stands

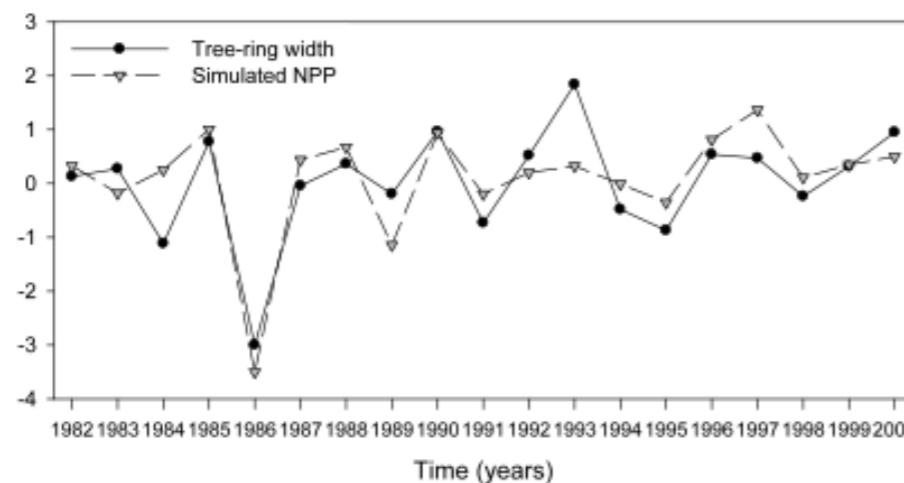
Multiple dataset used for model validation over times - General Model Evaluation - Why?



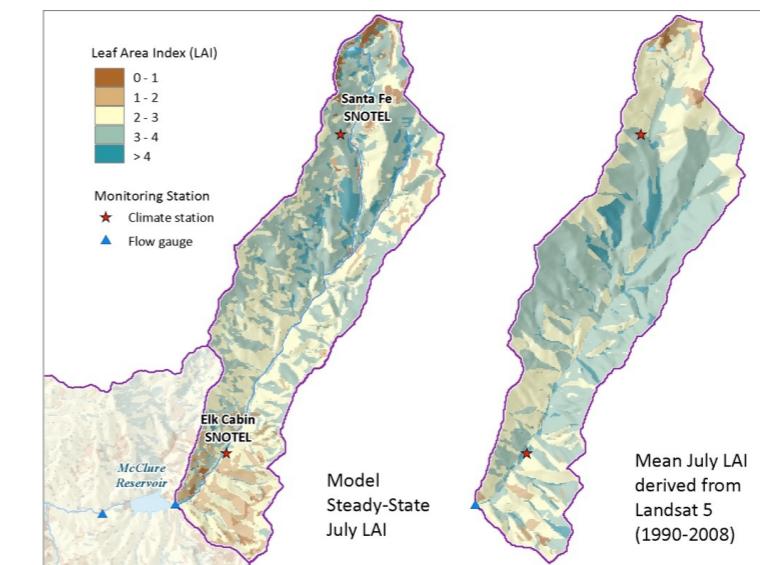
Daily Streamflow Record Tgue and Peng (2013) Journal of Geophysical Research: Biogeosciences 118(2): 875-887. doi: 10.1002/jgrg.20073



Carbon - flux Bart et al., (2017) PLoS ONE 11(8): e0161805. doi:10.1371/journal.pone.0161805



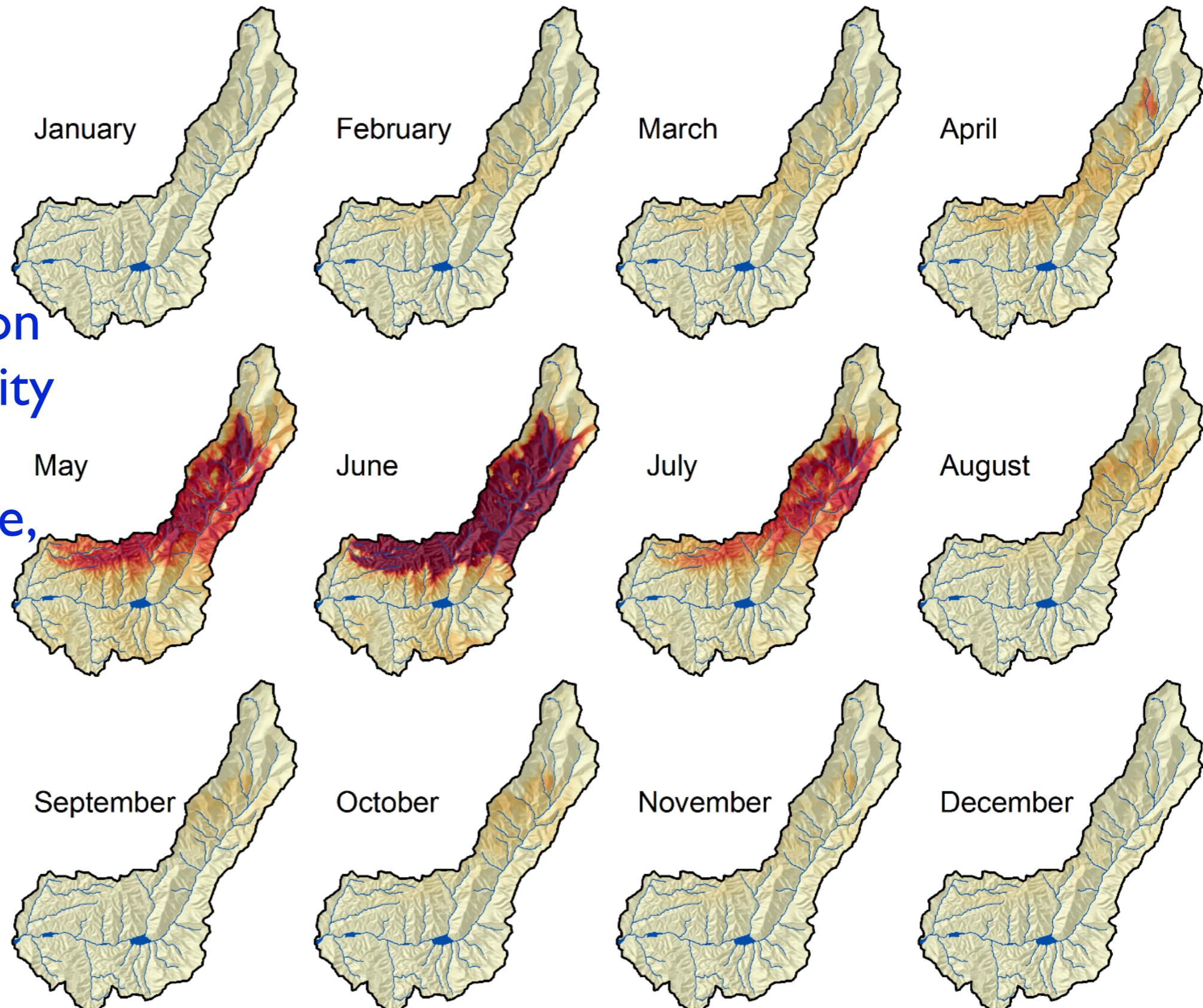
Tree Rings Vicente-Serrano et al., (2015), Agricultural and Forest Meteorology 206: 45-54. doi:10.1016/j.agrformet.2015.02.017



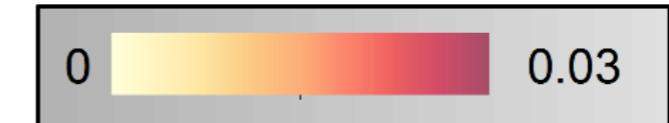
Remote Sensing Vegetation Indices

Turbance...RHESSys-Fire

Spatial estimation
of burn probability
under historic
climate: Santa Fe,
New Mexico

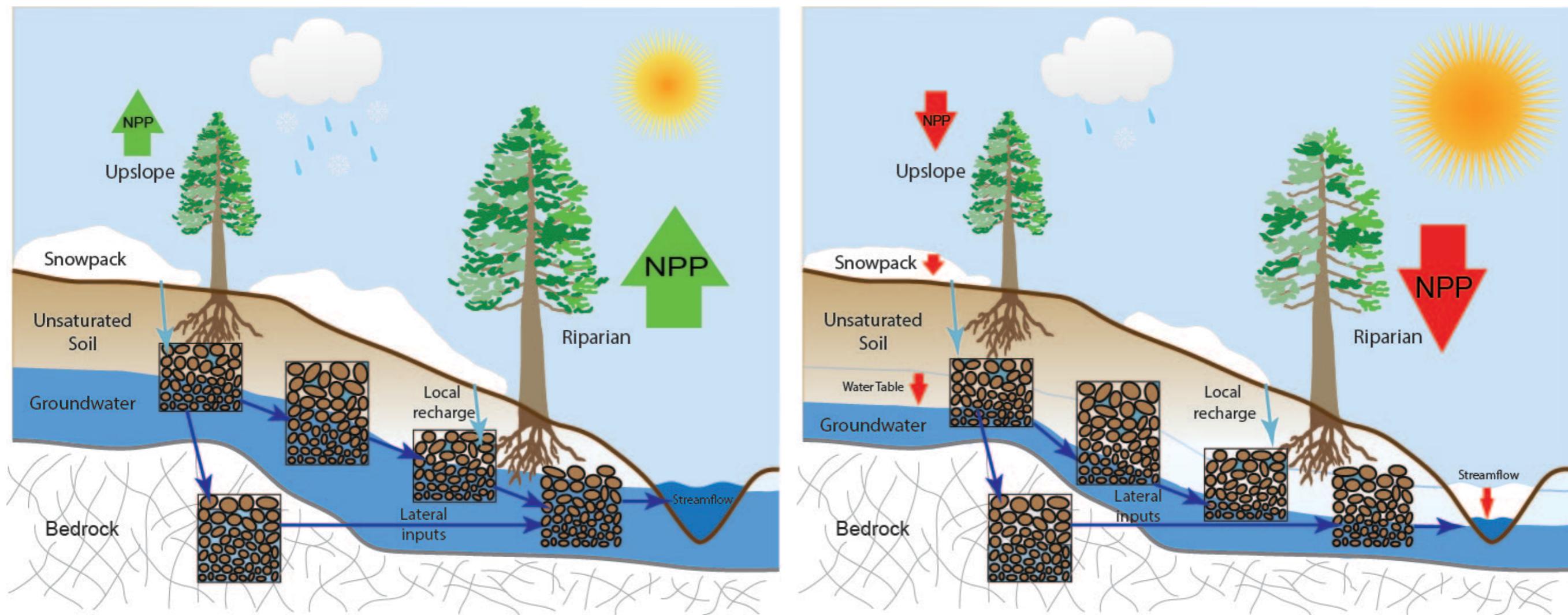


Kennedy et al. (2017) IJWF,
Bart et al., (2019) Eco Model



Conceptual Model

(Subsurface Lateral Flows Buffer Riparian Water Stress
Against Snow Drought, JGR, Graup et al., in 2022)

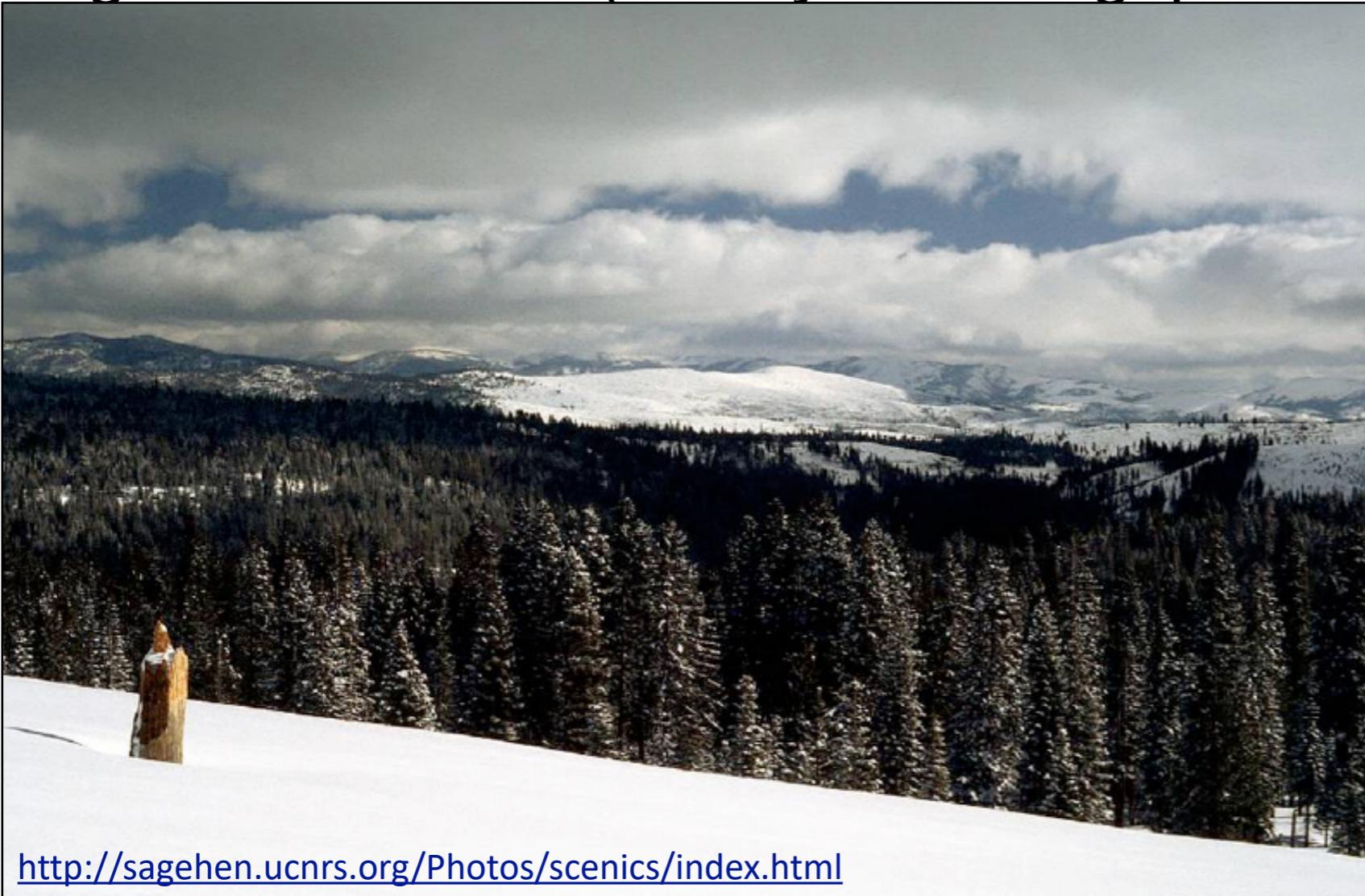


Focus Site

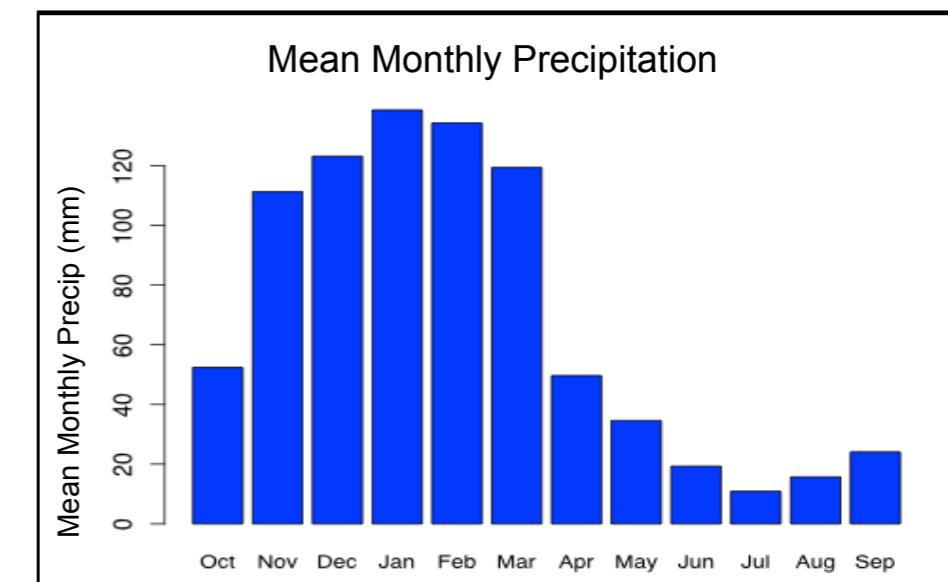
Sagehen Experimental Watershed (UC Berkley Field Station)

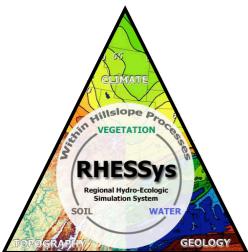
Sierra Nevada Mountain watershed
(183ha) Elevation range 1800-2700m

Vegetation: conifer (Jeffrey and Lodgepole



<http://sagehen.ucnrs.org/Photos/scenics/index.html>





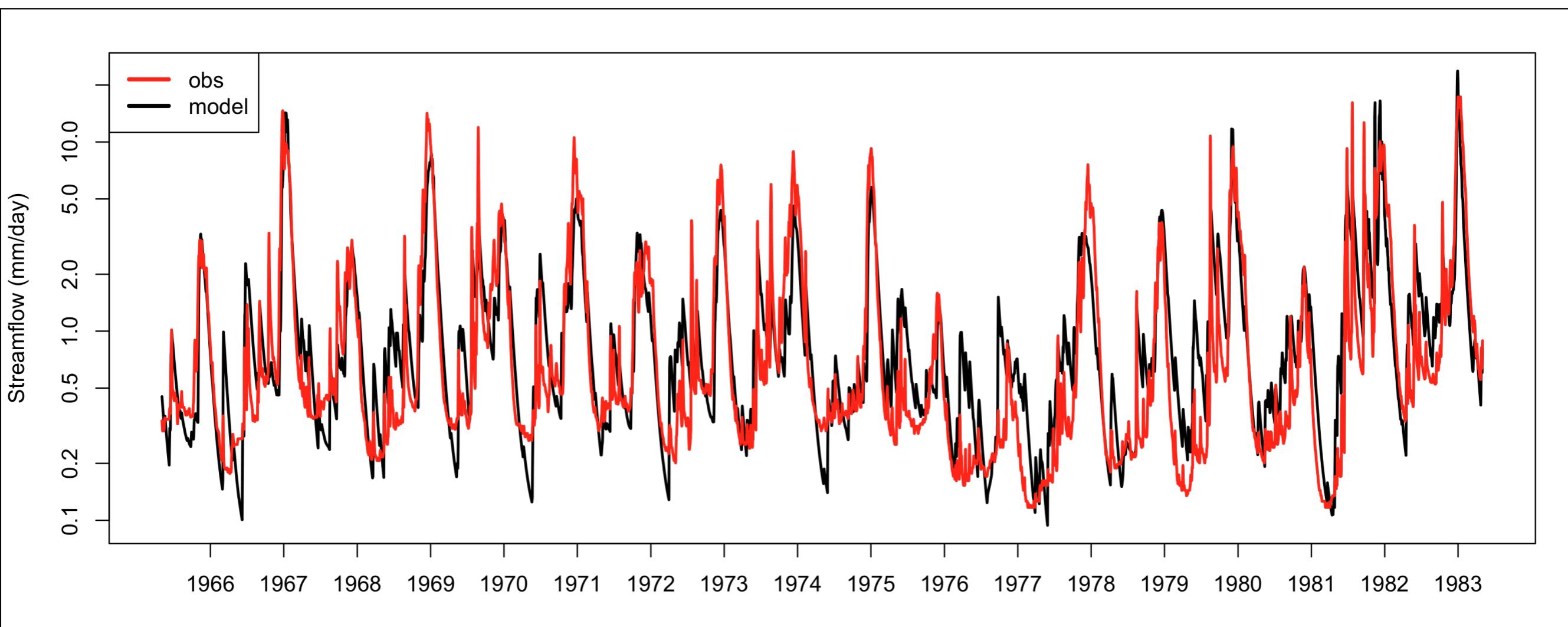
RHESSys hydrologic model performance – post calibration Streamflow (1960-2000)

NSE (daily) 0.6

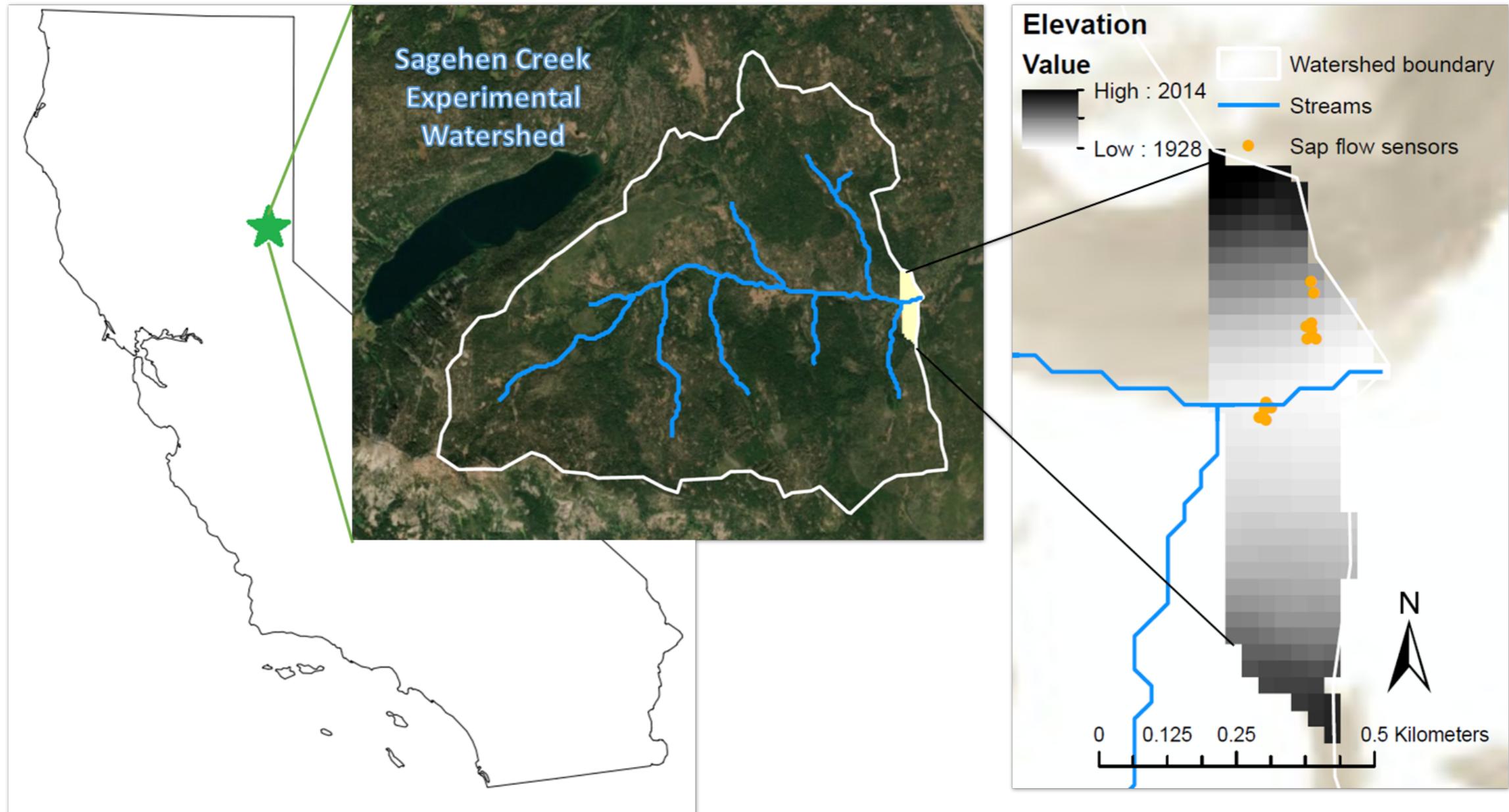
NSE (log transformed daily) 0.6

Bias < 10%

Monthly R2 > 0.9



RHESSys application - Sagehen Creek - Hillslope scale redistribution and drought



RHESSys application - Sagehen Creek - Diversity of observations for model evaluation

Growth Metric	Observed Range	Source	Modeled Mean (St. Dev.)
Tree Height (m)	3 - 32	Xu et al. (2018)	17.7 (4.7)
Δ Height (m)	0 - 5	Xu et al. (2018)	1.3 (0.4)
LAI (m^2/m^2)	1 - 8	Garcia et al. (2016)	2.5 (0.5)
Leaf Carbon ($g C / m^2$)	60 - 363	Law et al. (2001)	244 (45)
Stem Carbon ($g C / m^2$)	1,656 - 13,542	Law et al. (2001)	1,692 (351)
Coarse Root Carbon ($g C / m^2$)	1,500	Law et al. (2001)	836 (209)
	806 ± 142	Chatterjee et al. (2009)	
Fine Root Carbon ($g C / m^2$)	423 ± 95	Law et al. (2001)	188 (51)
	151 ± 7	Chatterjee et al. (2009)	
Plant Carbon ($g C / m^2$)	$3,640 \pm 770$	Johnson et al. (2008)	2,959 (610)
	$2,520 \pm 199$	Chatterjee et al. (2009)	



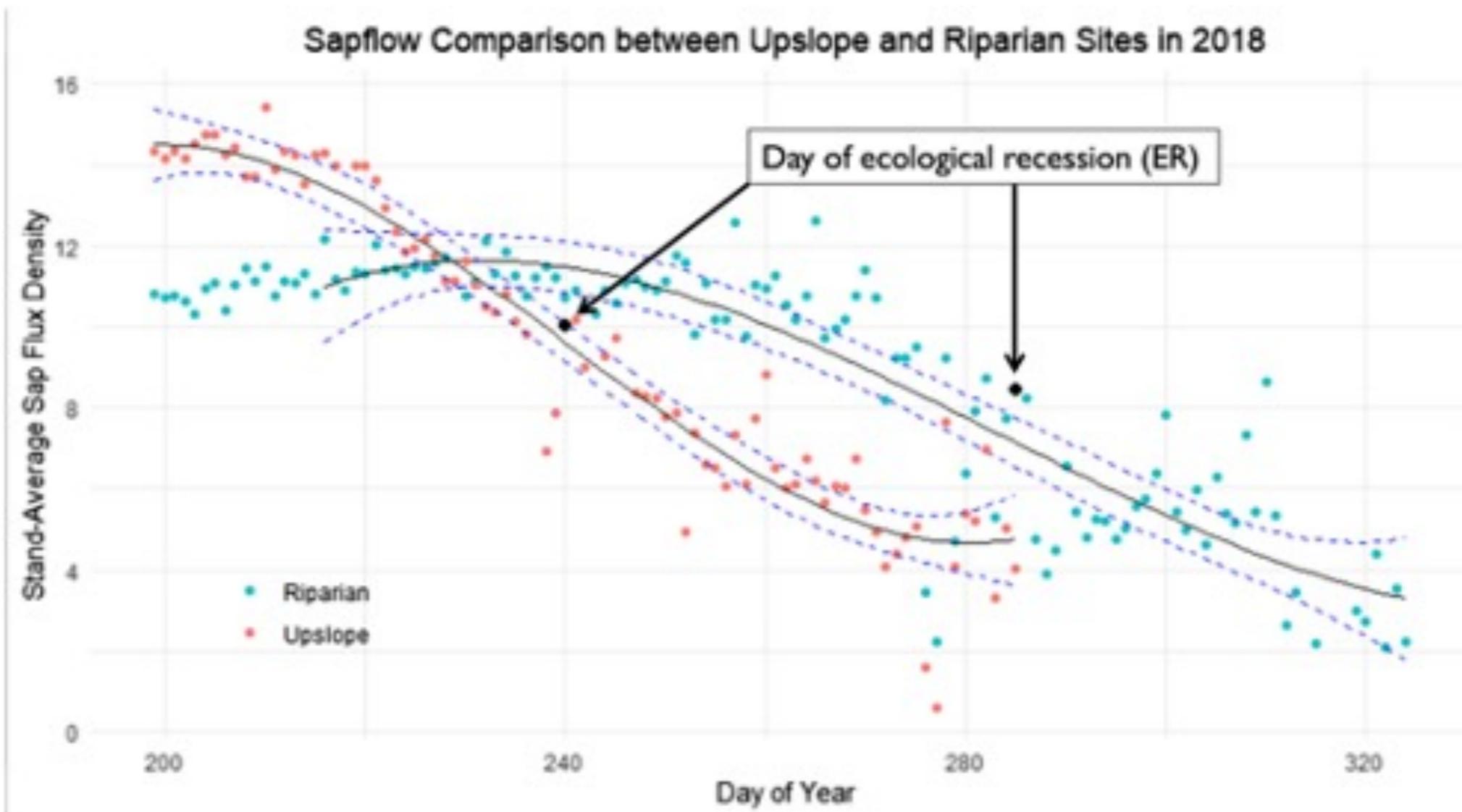
Table S3. Growth metrics derived from literature and Lidar data

First step - only parameters that RHESSys estimates of vegetation biomass (from carbon cycling model) have “reasonable” ranges

- * fit with literature values for these species in Sierra region
- * Fit with local estimates from Lidar

Figure S2: Tree height comparison of RHESSys model outputs against Lidar data in Sagehen

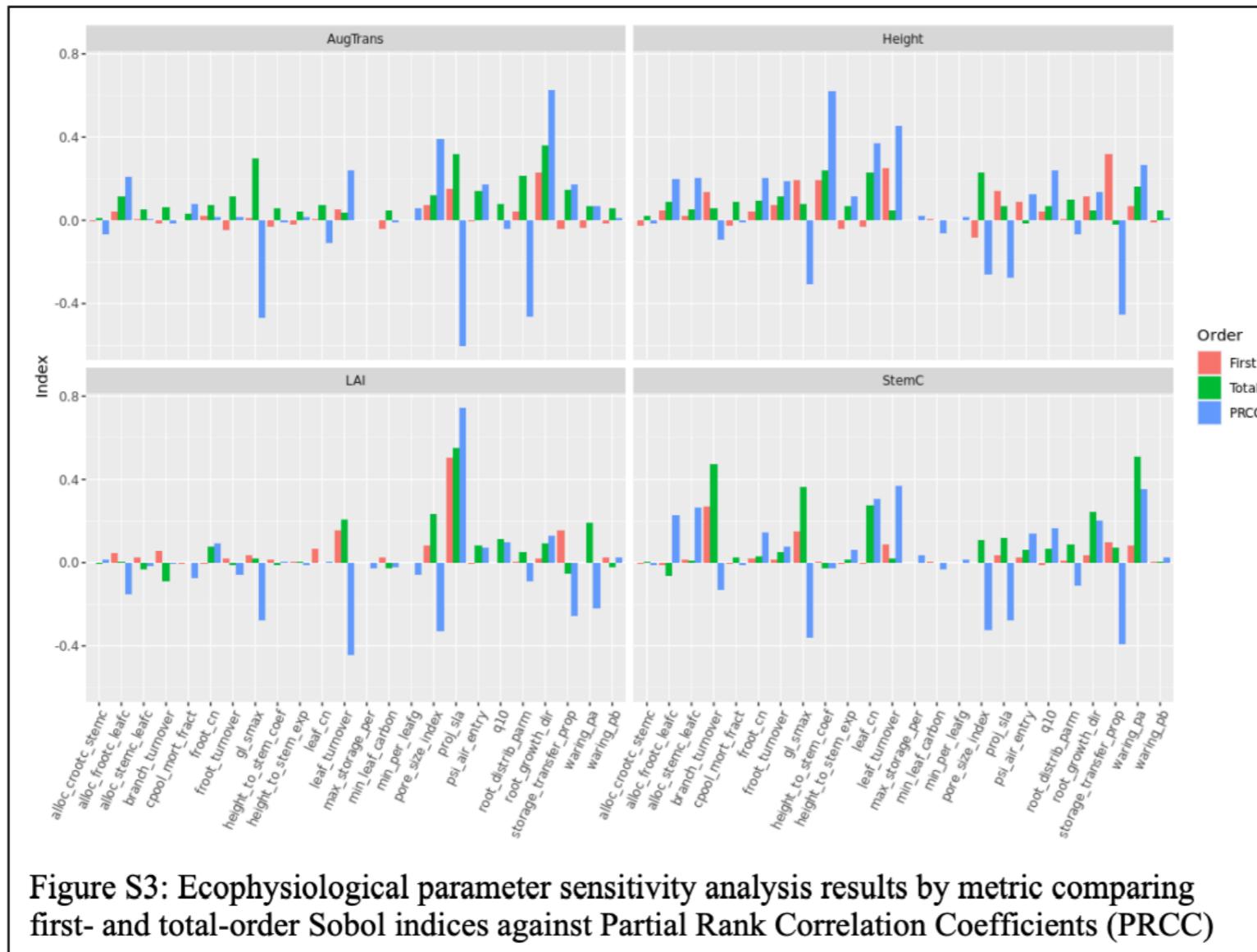
RHESSys application - Sagehen Creek - Calibration with Observed Sapflow



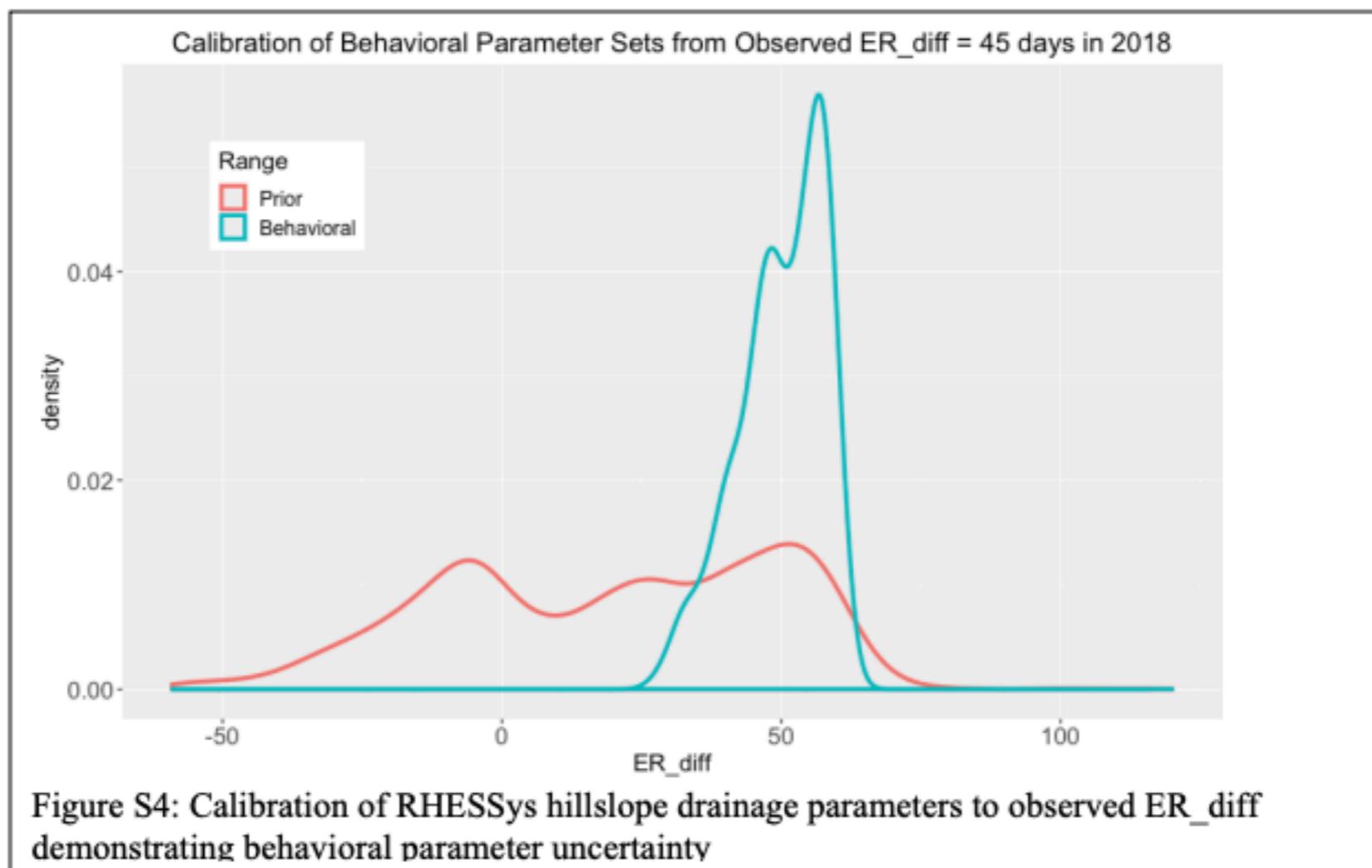
Second step (drainage parameters)

* fit with a metric derived from comparing upland and riparian sap flow

RHESSys application - Sagehen Creek - Find most sensitive parameters



RHESSys application - Sagehen Creek - Maintaining parameter uncertainty “Behavioral Parameters”



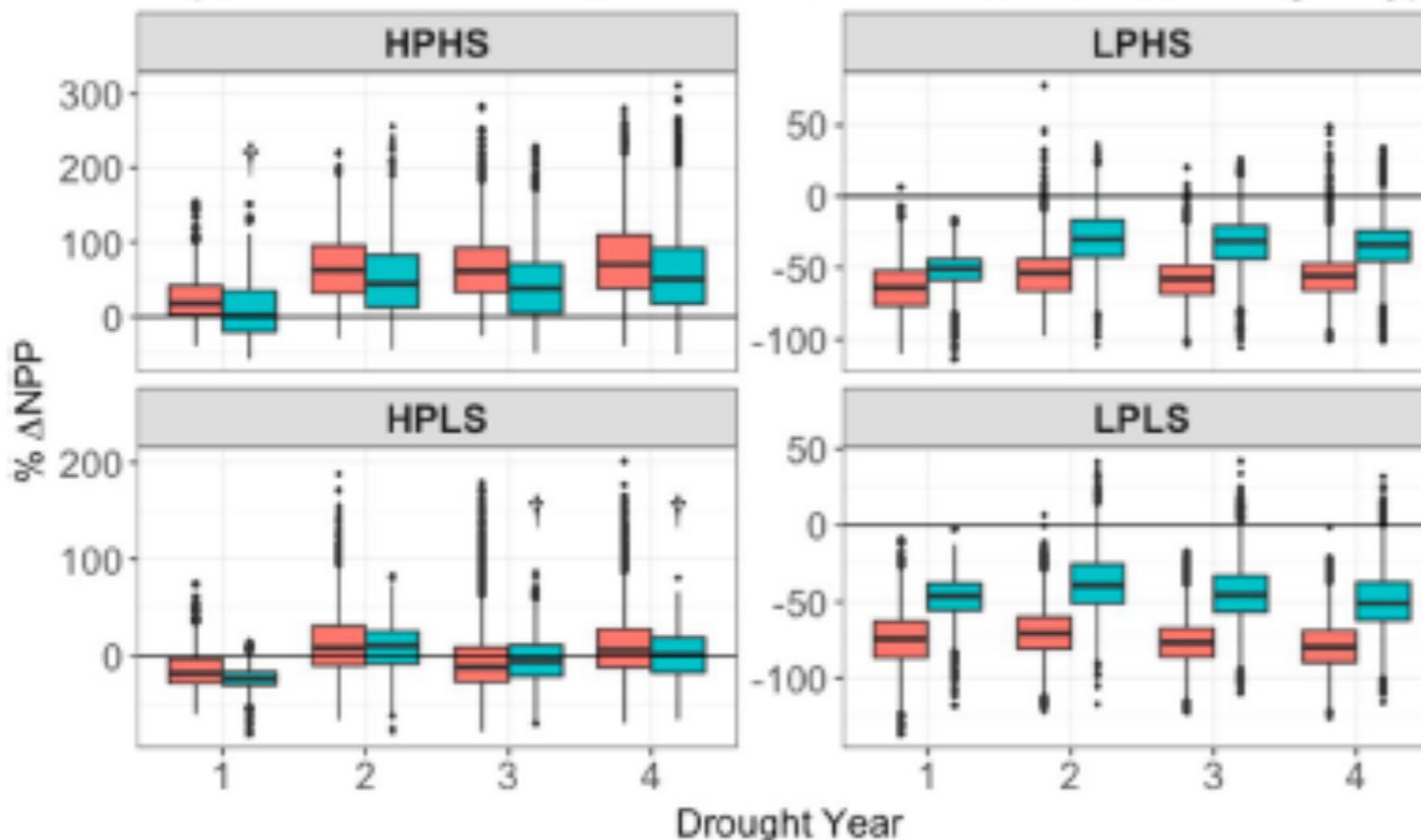
Multi-year drought how does NPP (water use by plant) change

Table 1
Drought Type Classification

P SWE	> Mean (677 mm)	< Mean
> Mean (128 mm)	HPHS	LPHS
< Mean	HPLS	LPLS

Note. P and SWE are accumulated annual precipitation and snow water equivalent, respectively, on April 1st. H is high, L is low, S is SWE—HPHS is high P, high SWE; LPLS is low P, low SWE (i.e., dry snow drought); HPLS is high P, low SWE (i.e., warm snow drought), etc.

Comparison of NPP Deviation from Control Year across Drought Types



- Riparian trees are buffered 28% of annual NPP loss during drought versus 45% for upslope trees
- But there were exceptions - particular parameters (fast draining soil, especially deep rooted, low conductance)

Site



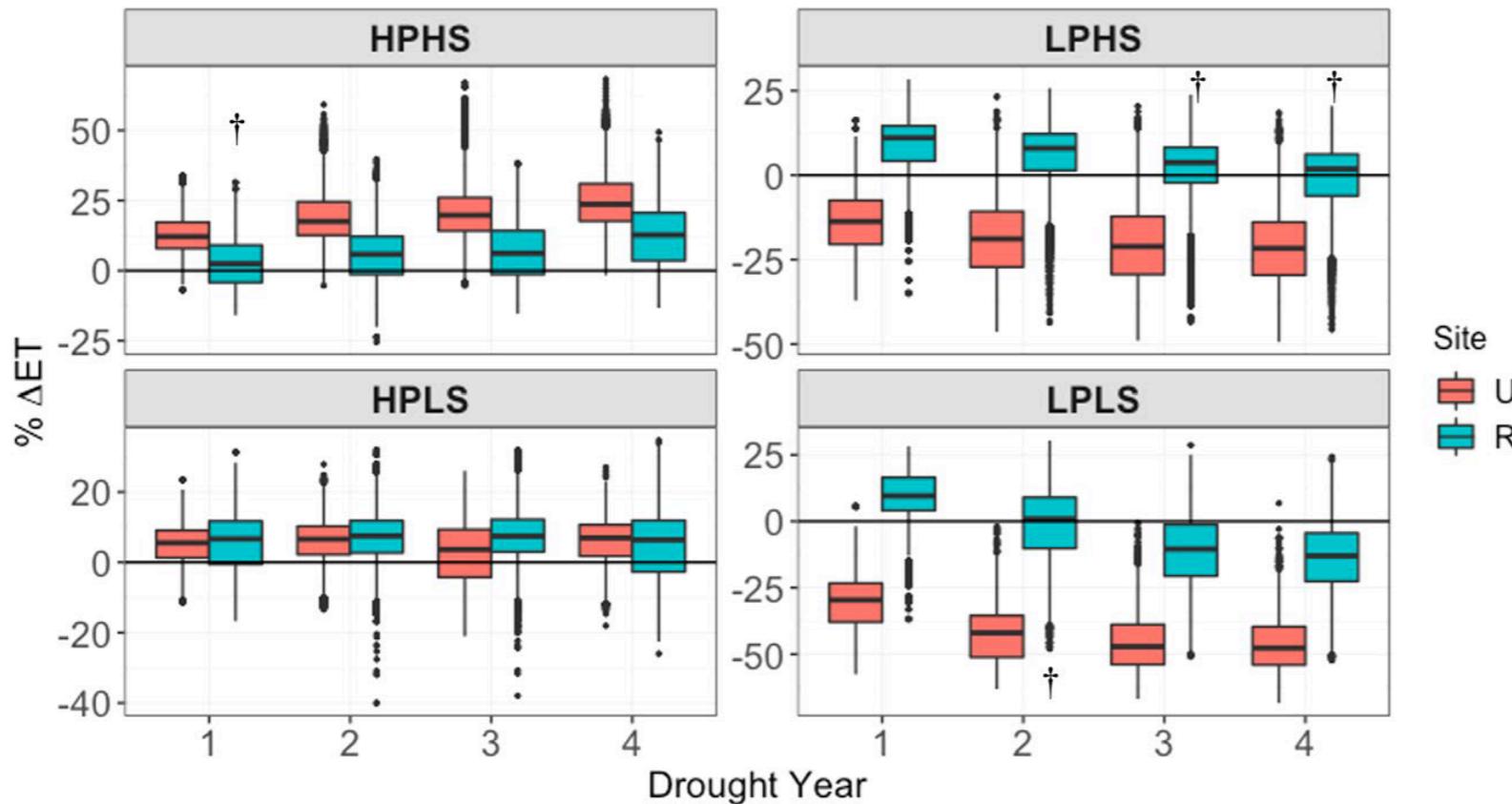
Multi-year drought how does ET(water use by plant) change

Table 1
Drought Type Classification

P SWE	> Mean (677 mm)	< Mean
> Mean (128 mm)	HPHS	LPHS
< Mean	HPLS	LPLS

Note. P and SWE are accumulated annual precipitation and snow water equivalent, respectively, on April 1st. H is high, L is low, S is SWE—HPHS is high P, high SWE; LPLS is low P, low SWE (i.e., dry snow drought); HPLS is high P, low SWE (i.e., warm snow drought), etc.

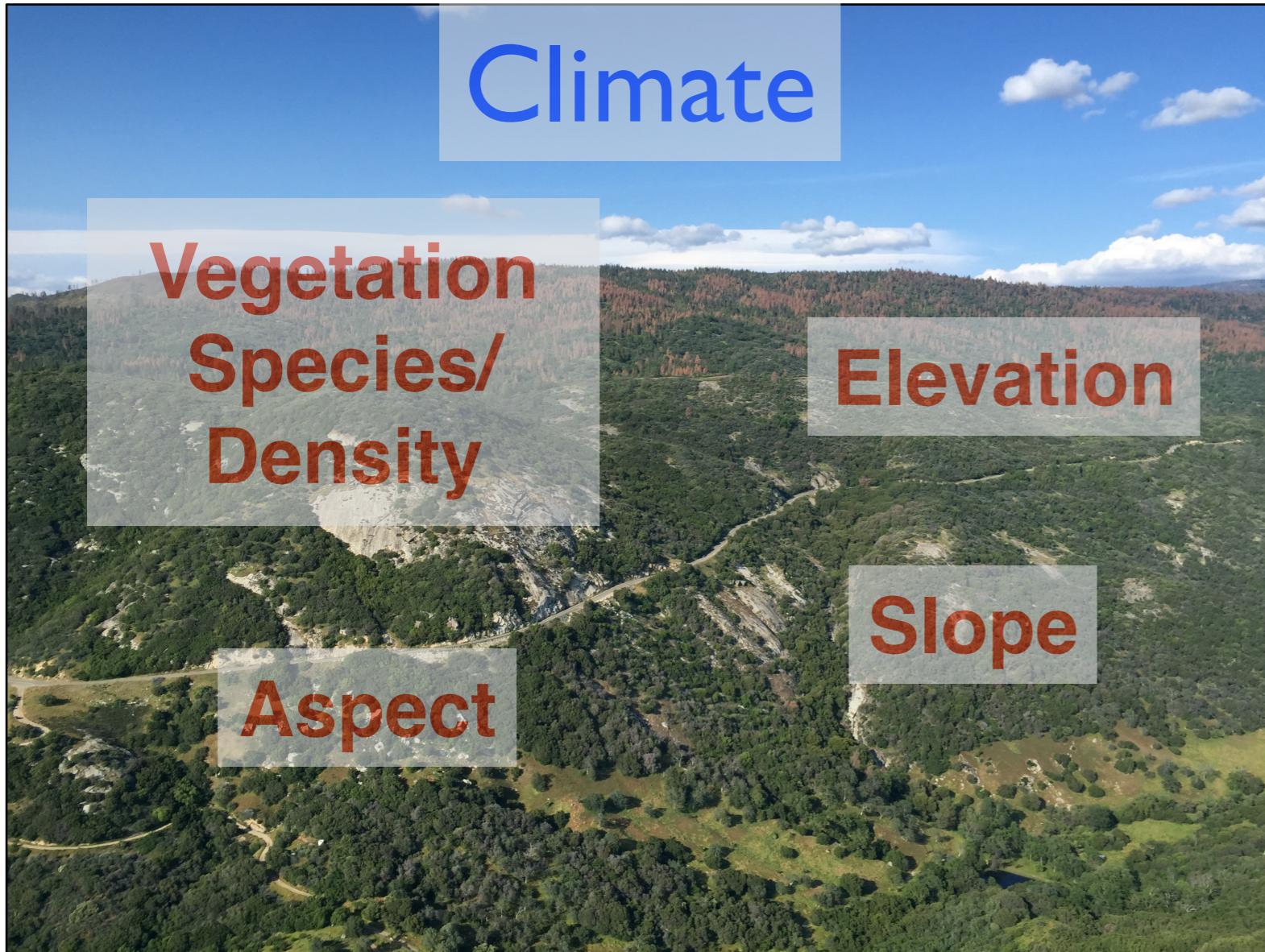
Comparison of ET Deviation from Control Year across Drought Types



- Low P, riparian trees initially benefit until 4th year of drought, and snow buffers this

Site
U
R

Fuel treatment impacts - over time on water, carbon, and fire - what matters? (Burke et al., 2021)



Southern Sierra Watershed



Type of treatment -
biomass removal,
rotation...

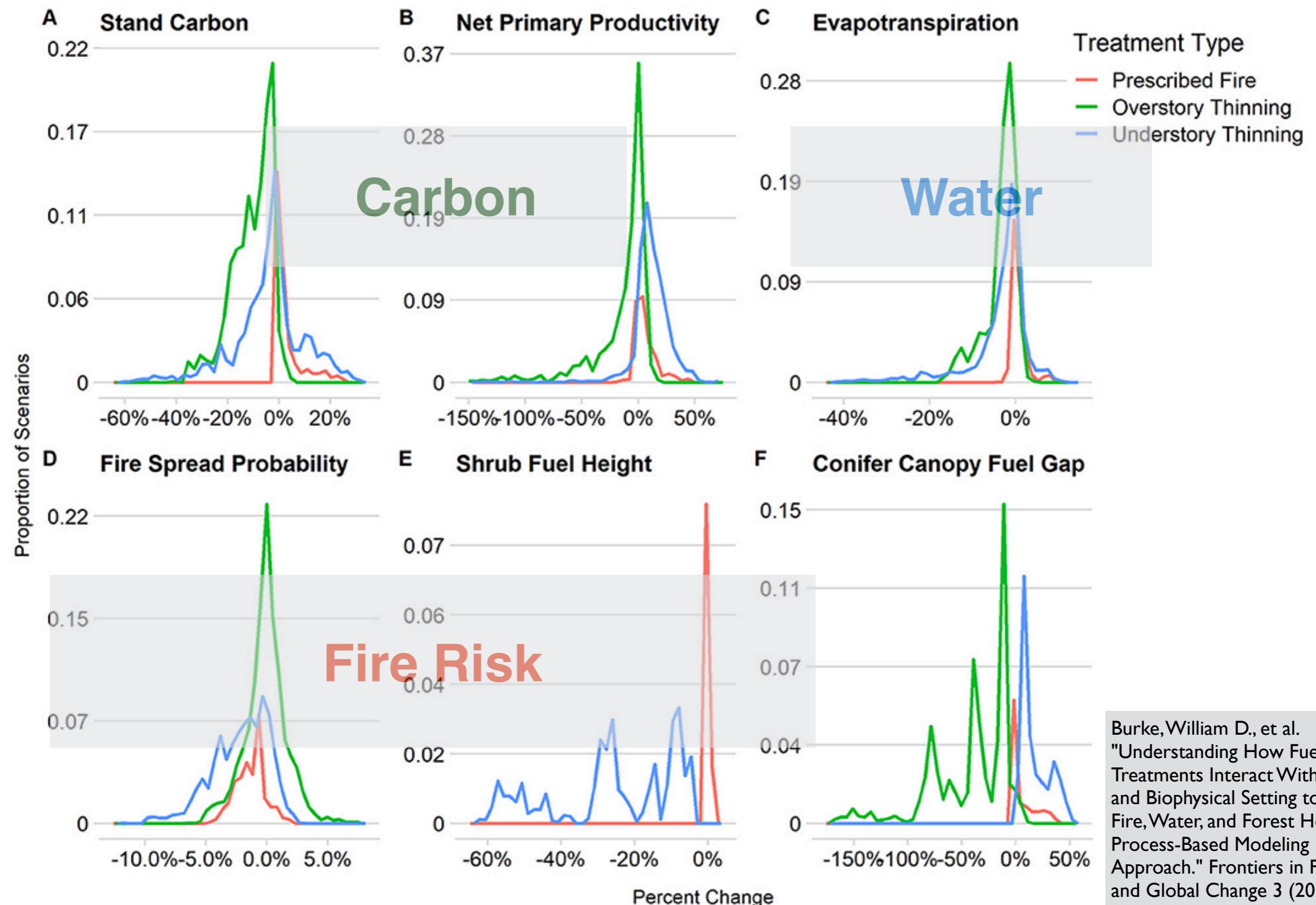
RHESSys-Fire: Context dependence of fuel treatment effectiveness - even within the same watershed

TABLE 1 | Summary of fuel treatment scenario parameters.

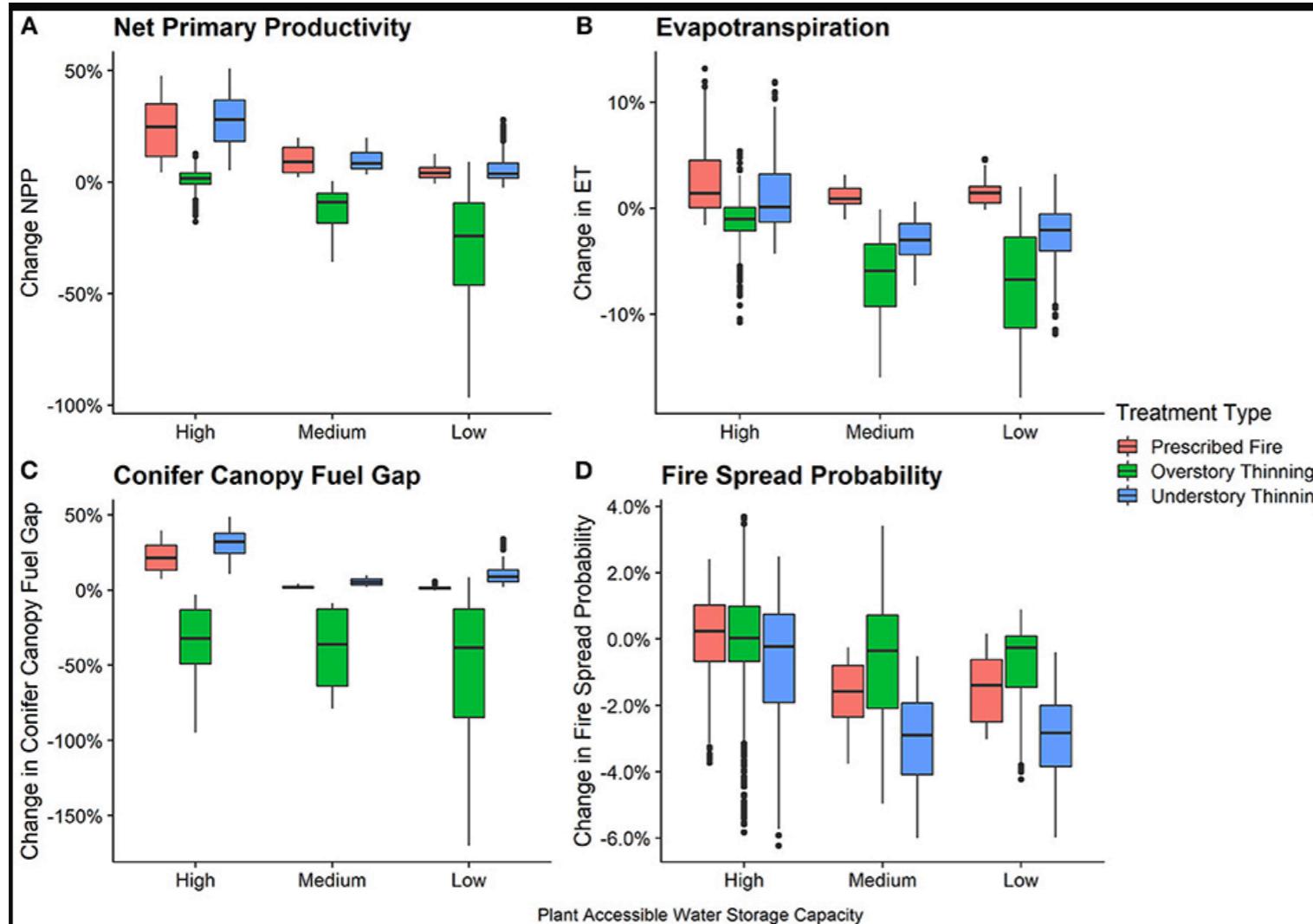
Fuel treatment scenarios	
Treatment method and intensity	10
Understory thinning + prescribed fire: high, med, low	3
Overstory thinning, with/without slash: high, med, and low	6
Prescribed fire	1
Treatment Frequency: 5, 10, and 30 years	3
No treatment	1
Site characteristics	540
Vegetation: shrub, conifer, and shrub/conifer mix	3
Aspect: north, south	2
Plant accessible water storage capacity: low, med, and high	3
Aridity: dry, variable, and wet	3
Climate warming: baseline, + 2°C	2
Root sharing coefficients: 0, 0.25, 0.5, 0.75, and 1	5
Total (incompatible combinations removed)	13,500

Bold values highlight the major subcategories of scenario variation.

Variation in fuel treatment impacts across site parameters, climate, treatments



Forest growth, water use and fire indicators as responses to fuel treatment



Note parameter interactions - impact of treatment type depends on subsurface water storage

Fire Severity Risk

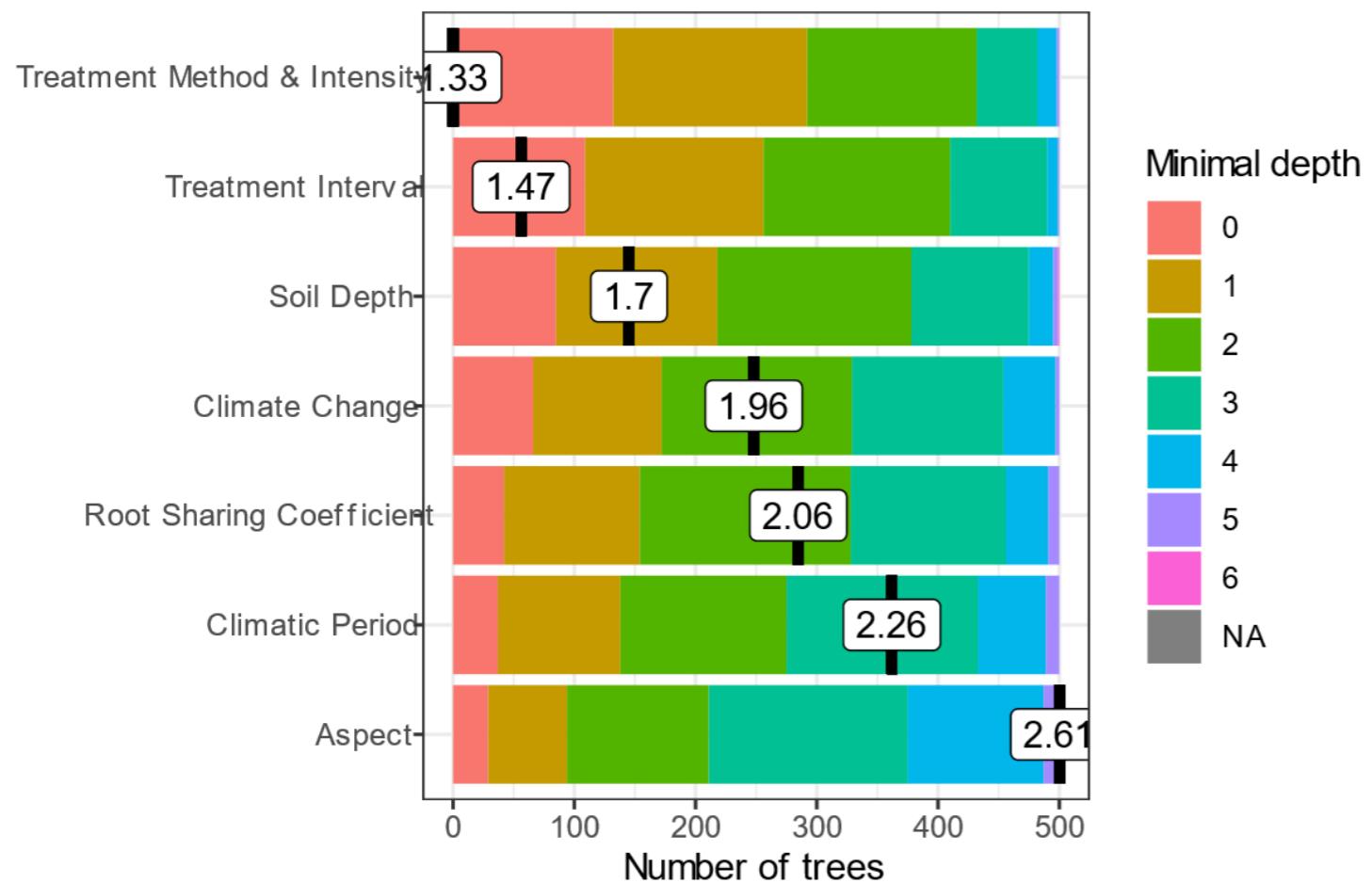


TABLE 1 | Summary of fuel treatment scenario parameters.

Fuel treatment scenarios

Treatment method and intensity

Understory thinning + prescribed fire: high, med, low
Overstory thinning, with/without slash: high, med, and low
Prescribed fire

Treatment Frequency: 5, 10, and 30 years

No treatment

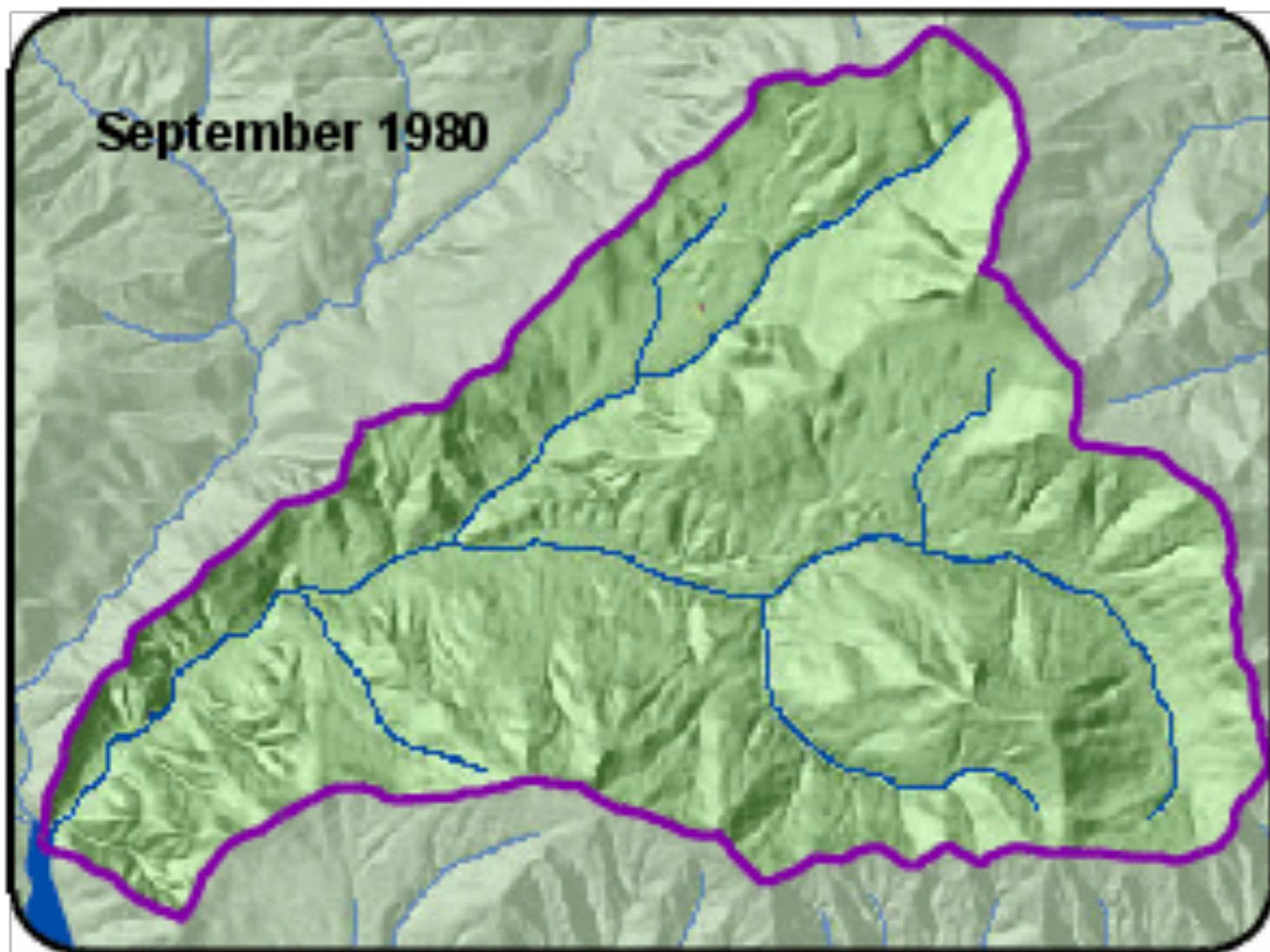
Site characteristics

Vegetation: shrub, conifer, and shrub/conifer mix
Aspect: north, south
Plant accessible water storage capacity: low, med, and high
Aridity: dry, variable, and wet
Climate warming: baseline, + 2°C
Root sharing coefficients: 0, 0.25, 0.5, 0.75, and 1

Total (incompatible combinations removed)

Bold values highlight the major subcategories of scenario variation.

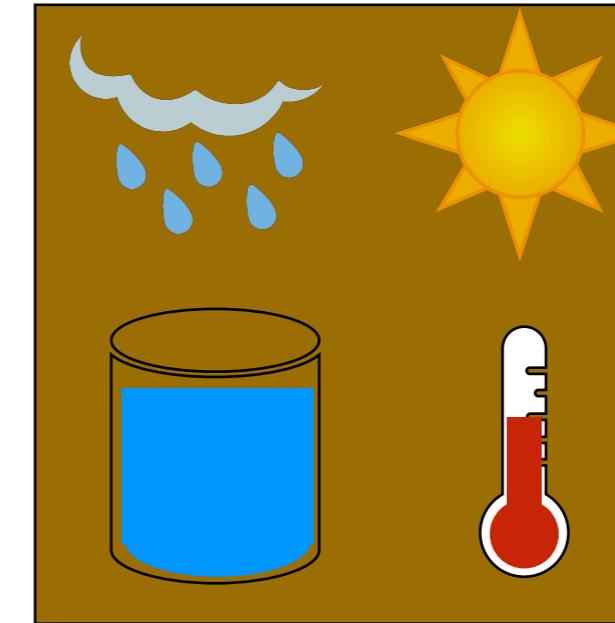
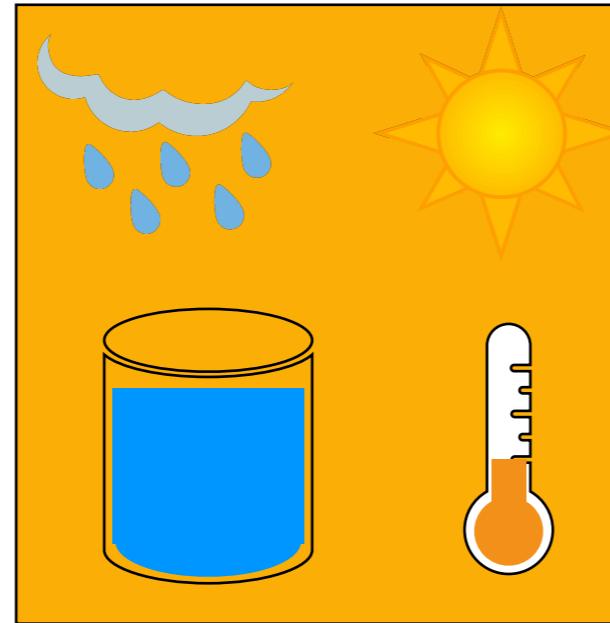
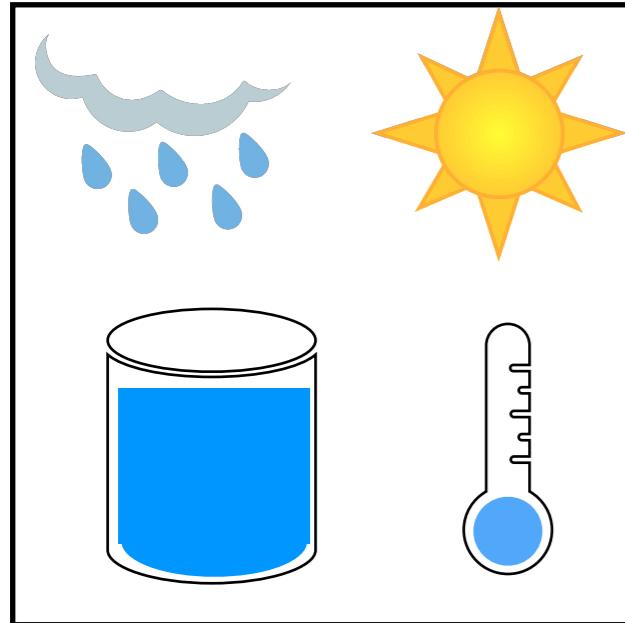
What about fire....RHESSys-Fire



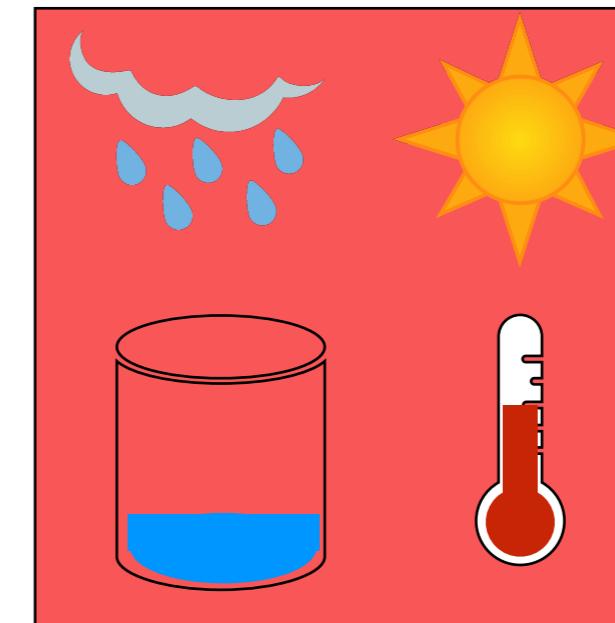
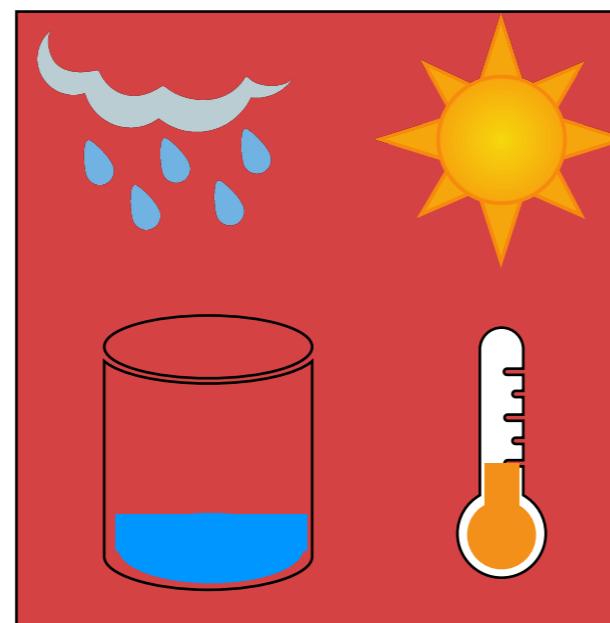
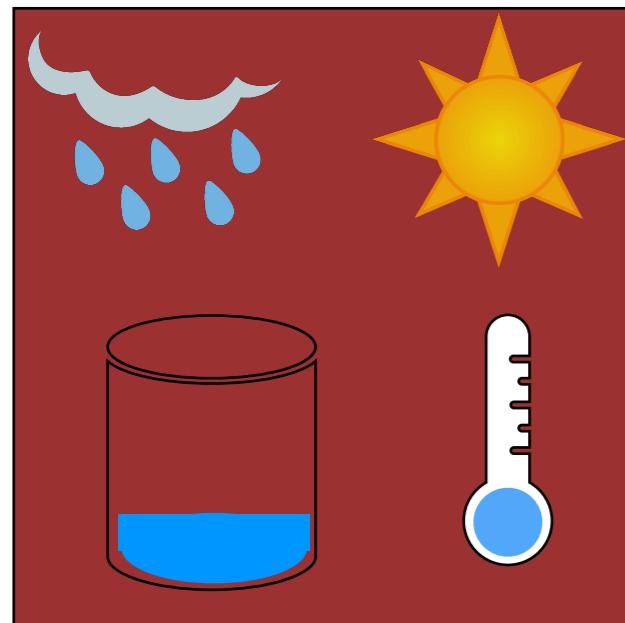
Kennedy et al. (2017)
International Journal of Wildland Fire



What happens to fire regimes if droughts are more frequent? If droughts are warmer?

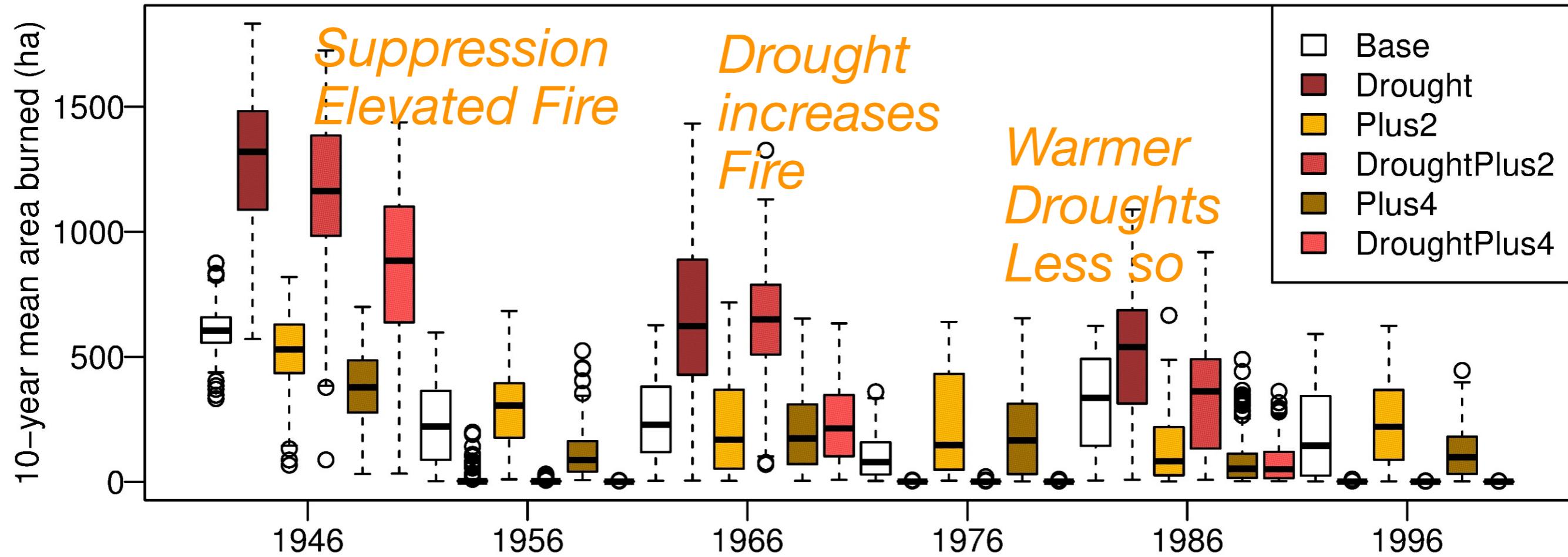


- Base
- Drought
- Plus2
- DroughtPlus2
- Plus4
- DroughtPlus4



2010's drought
repeated every
10 years

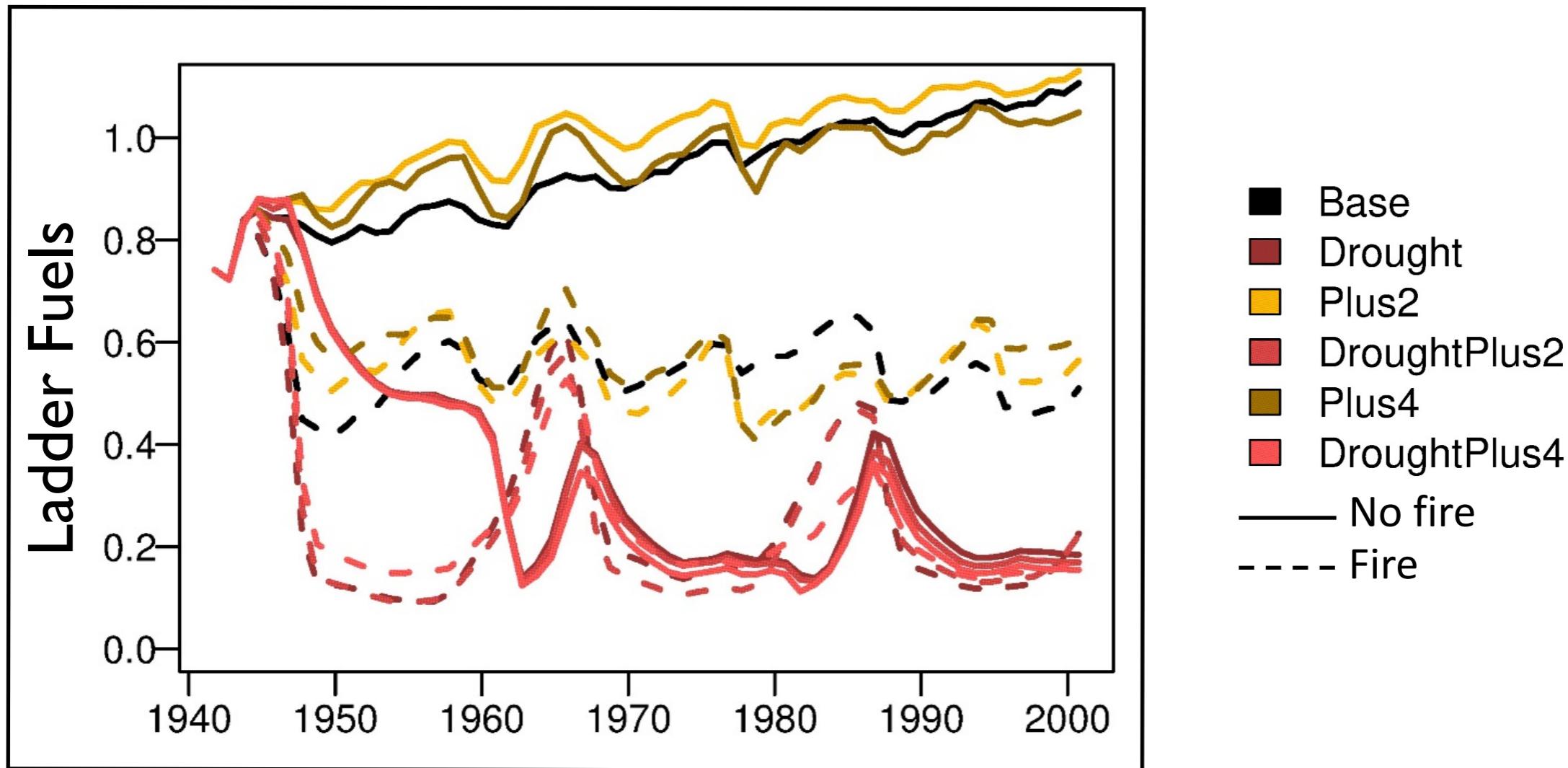
Results of the model experiment



Baseline Climate Time Period

Kennedy et al., (2021) Does hot and dry equal more wildfire? Contrasting short and long-term climate effects on fire in the Sierra Nevada, CA, *Ecosphere* 12(7): e03657

Fire vegetation over time



- Warm droughts:
reduce vegetation productivity and regrowth - leading to fuel limitation - less fire
- Regular drought:
increase in fire - weather dominates
- Varies WITHIN watersheds

Related studies in Idaho - suppression vs. climate change for last decades



RHESSys-Fire scenarios - spatial patterns of burn probably with/without historic fire exclusion and climate change (since 1980)

Burn probability and impacts suppression and recent climate change vary with location's aridity

Our answers always are ‘it depends’.....on

- the weather you get
- the soil
- when the fire occurs
- what species come back

How do we communicate this complexity?

Eco-hydrology Modeling for Integration

Many have advocated for using models for:

- eco-hydrology hypothesis testing
- behavioral modeling
- dialog between experimentalist and modelers
- placing new conceptual models into context

(Clark, et al., 2011, *WRR*; Schaeefli, 2011, *HESS*,)

But we don't do this very often...

And when we do we mistrust results...

Why?

Barriers: Making use of Eco-hydrology Modeling for Integration

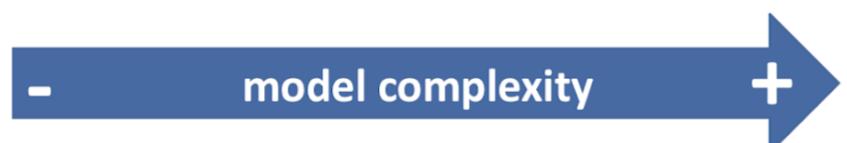
Black Box
Model

Deep dive
into papers
or model
code



*Is just too hard to know what
complex models do*

Review of models used to estimate how intentional changes in forest structure impact water cycle



Factor	No Recovery	Statistical Recovery	Physical-constant Recovery	Physical-dynamic Recovery
time: stand recovers through time	No	Yes	Yes	Yes
resources: availability of light, water, nutrients changes through time	No	No	No	Yes
canopy structure: stem density, age class	No	No	No	Yes
species composition: evolution through time	No	No	Yes	Yes
fire regime: change in fire regime through time	No	No	No	Yes
mortality risk: change in mortality risk through time	No	No	Yes	Yes
example	Statistical-Dynamical Ecohydrology Model	FVS	Biome-BGC	NA

Conclusion:
Diversity is HIGH;
Transparency LOW

Challenging to integrate new science (particularly on forest adaptation to climate change) into these tools

Eco-hydrology Models: Revealing what is hidden

Example: Future Mountain interface for RHESSys

Ethan Turpin, David Gordon, & human-computer interface researchers.



Eco-hydrology Models: Revealing what is hidden

Example: Future Mountain interface for RHESSys

Ethan Turpin, David Gordon, & human-computer interface researchers.





Cube Comparisons:

Simple warming impacts: less snow

Wildland Art Show- Ethan Turpin

Search 

 WESTMONT RIDLEY-TREE
MUSEUM of ART

VISIT PROGRAMMING GET INVOLVED PEOPLE STORE PERMANENT COLLECTION WESTMONT HOME



WILDLAND: Ethan Turpin's Collaborations on Fire and Water

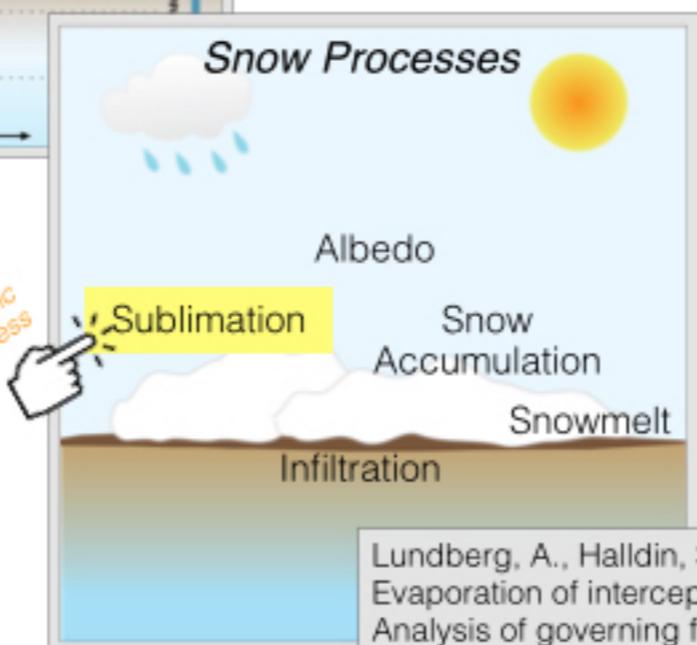
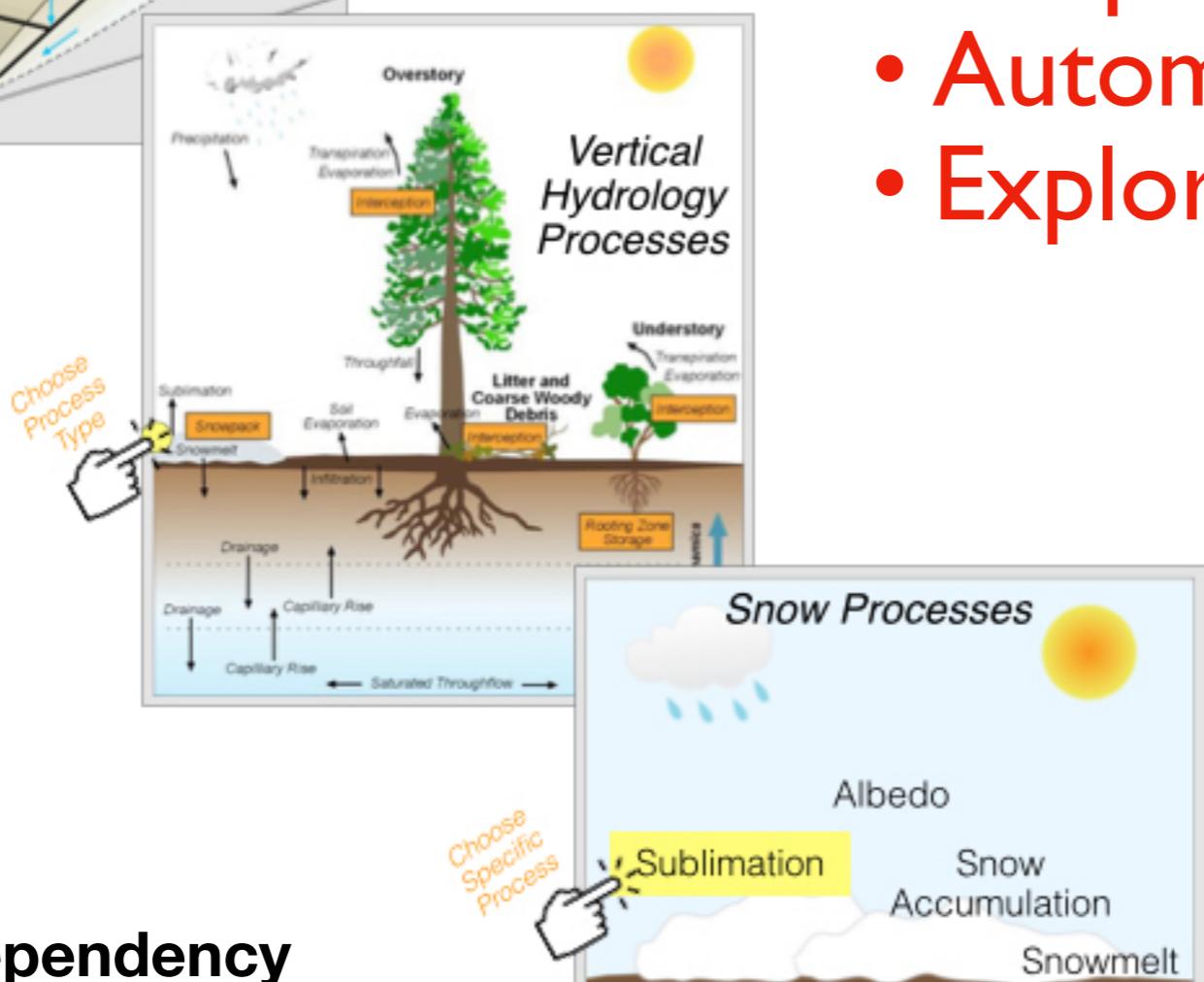
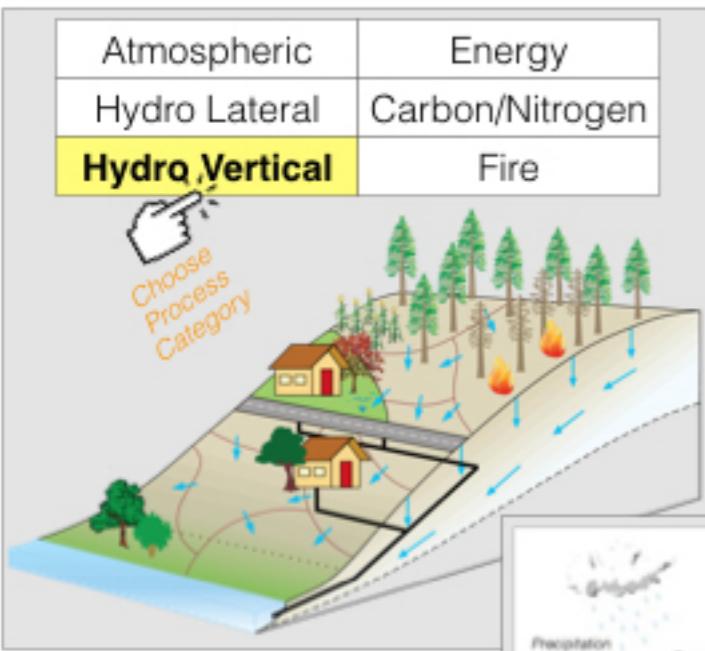


Artist & Collaborator Lecture

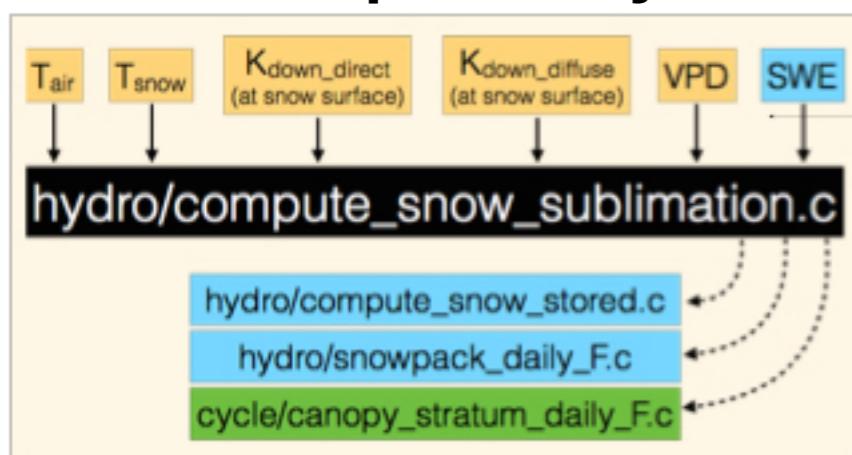
When: January 30, 2025 at 5:30pm

Where: Porter Theatre, Westmont College

Ethan Turpin and Naomi Tague will present a lecture on their collaborations for *WILDLAND*. Naomi Tague is a researcher and professor of ecology at the University of California, Santa Barbara's Bren School of Environmental Science & Management. Join us for "Beyond Data Visualization: Art-Science Collaboration in an Era of Global Environmental Change."



Process Dependency



- Searchable
- Graphical
- Automatically Generated
- Explorable..

Process sub-model

Lundberg, A., Halldin, S. (1994)
Evaporation of intercepted snow:
Analysis of governing factors, Water
Resources Research 30(9): 2587-2598.
doi.org/10.1029/94WR00873

$f(K_{down\,direct}, K_{down\,diffuse}, vpd, ga_{snow}, T_{air}, T_{snow}, SWE)$



the **tague team** lab
ECOHYDROLOGY+INFORMATICS

Funding

Tague Team Lab:

Ryan Bart (post-doc)

Louis Graup (PhD)

Will Burke (PhD)

Sloane Stevenson (PhD)

Janet Choate (Lab Manager)

Maureen Kennedy (UW)

Erin Hanan (Univ. Nevada)

Tamir Klein (Weizmann
Institute)

Lab Blog: tagueteamlab.org

About: fiesta.bren.ucsb.edu/~rhessys/

Code: github.com/RHESSys/RHESSys



Bren School
of Environmental Science & Management
UNIVERSITY OF CALIFORNIA, SANTA BARBARA



National Science Foundation: US-Israel
Binational Science Foundation