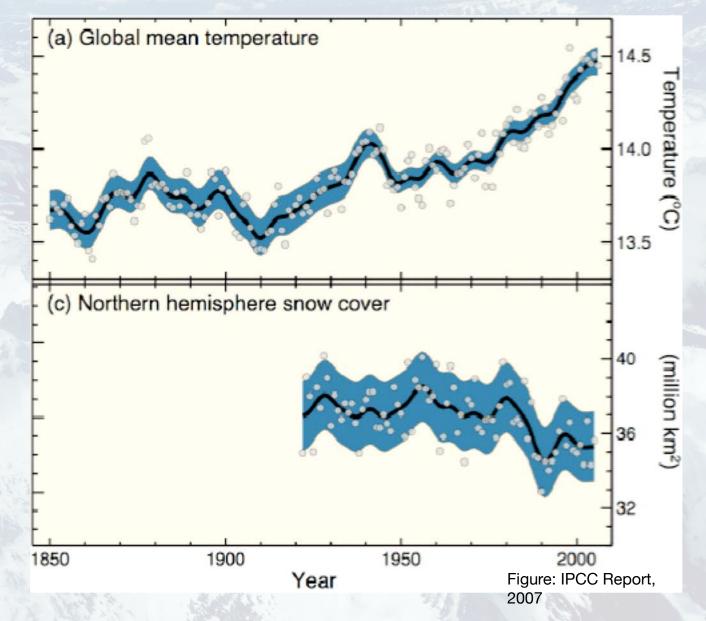
WHEN DOES UNCERTAINTY MATTER WHILE MODELING CLIMATE [CHANGE] IN MOUNTAIN HEADWATERS?

Contrasting Model Resolution and Complexity in an Alpine Catchment





CLIMATE CHANGE = ↑ Temp, ↓ Snowpack



MOUNTAIN SNOWPACK CRITICAL FOR WATER SUPPLIES

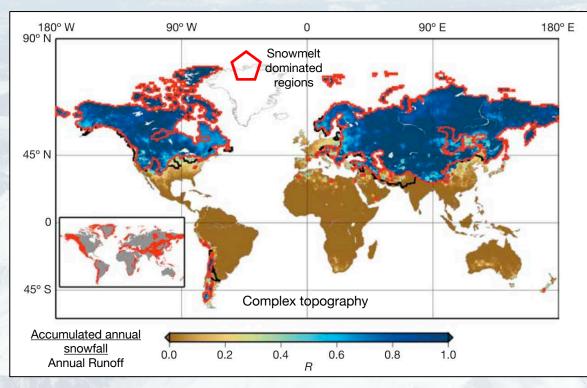


Figure (Barnett et al 2005): More than 1/6th population depends on surface water supplies from snowmeltdominated systems.

MOUNTAIN SNOWPACK CRITICAL FOR WATER SUPPLIES

38°N-

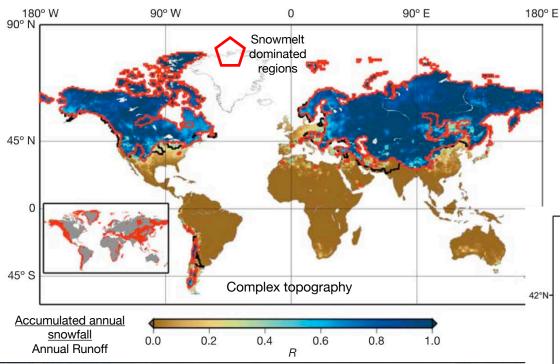


Figure (Barnett et al 2005): More than 1/6th population depends on surface water supplies from snowmeltdominated systems.

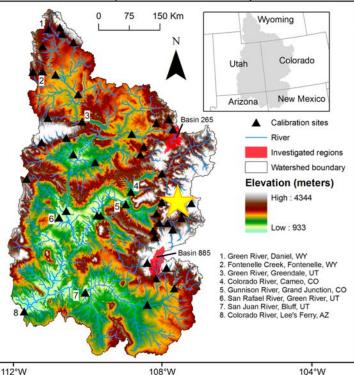


Figure (Ficklin et al 2013): More than 85% of upper Colorado R. Streamflow (main supply for Southwestern United States) generated from snowmelt in Rocky Mountain Headwaters.

Mountains to warm more quickly (NCC 2017)

NATURE CLIMATE CHANGE DOI: 10.1038/NCLIMATE2563

REVIEW ARTICLE

Table 1 | Results from studies that investigated elevational gradient in warming rates (updated from ref. 25).

	Observations			Models		
Elevational gradient in the warming rate	T _{min}	T _{max}	Tavg	T _{min}	T _{max}	T _{avg}
Increases with elevation	Annual ^{23,8a,86} Winter ^{23,70g,45} Spring ⁴⁷ Autumn ^{47,83e}	Annual ^{23,81} Summer ⁴⁷	Annual ^{76,5e,80b,10c,11,75} All seasons ⁴⁷ Winter ^{83e}	Annual ⁴⁵ Winter ^{45,43,72,74} Spring ^{45,43}	Winter ⁴³ Spring ⁴³ Autumn ⁷²	Annual ⁷⁹ Winter ^{30,69,14,79} Spring ^{30,14,79} Summer ⁸⁶
Decreases with elevation	Winter ²³	Winter ⁴⁷	Annual ^{77,84} Winter ^{30g} Autumn ^{30g}	Summer ⁸⁴		Annual ^{73f} Spring ^{73f} Autumn ^{73f}
No significant gradient	-	Annual ⁶⁸	Annual ^{30,85,71d,82} All seasons ^{44,85}	-		Annual ³⁰
No significant gradient but largest warming rates at an intermediate elevation	-	Annual ⁷⁸	Annual ^{9,43c} Spring ^{30g}	-		Spring ³⁰

Superscript letters accompanying references indiciate: *No significant gradient but greater warming at higher elevations relative to regions between 0–500 m; ^bradiosonde data, clearest signal in the tropics; ^c65% of the regional groups examined showed fastest trends at highest elevations and 20% showed fastest trends at intermediate elevations; ^dhigh-elevation trends based on borehole data; ^esatellite-derived temperature estimations; ^freanalyses; ^ggridded data.

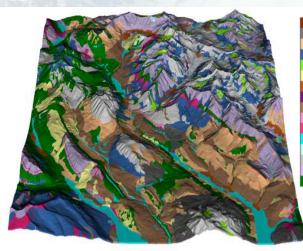
Mountains are sensitive but process-based understanding is limited by complexity

Range of elevations Steep Temperature Gradients Variable Precipitation

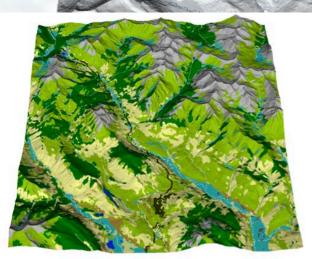


Mountains are sensitive but process-based understanding is limited by complexity

Range of elevations Steep Temperature Gradients Variable Precipitation Heterogeneous Geology & Landcover



Subsurface **Debris Flow** Undifferentiated Talus **Rock Glaciers** Landslide Fan Volcanic Glacial Carbonate **Course Grained Sedimentary** Alluvial Colluvium - Crystalline Lower Body - Mancos Shale Main Body - Mancos Shale



Grasslands
Barren
Evergreen Needleleaf
Water Bodies
Permanent Wetlands
Deciduous Broadleaf
Open Shrublands
Cropland
Built-up

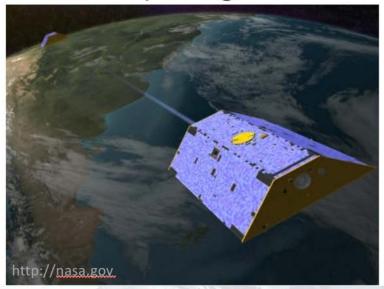
∆1418m

All methodologies simplify the real world...

• Observations:

Local measurements are Remote sensing can't difficult to scale see everything





All methodologies simplify the real world...

Models

 Coarse resolution models to make decisions/predictions
 -> run quickly, missing feedbacks

Rasmussen et al. I I SRES A2 I (2011)	2-, 6-, 18-, and 36-km grids	Pseudo-global warming approach	RCM runoff
--	---------------------------------	--------------------------------------	------------

• Fine resolution models are computationally expensive

		2			0			
Rasmussen et al. (2011)	1	1	SRES A2	1	2-, 6-, 18-, and 36-km grids	Pseudo-global warming approach	RCM runoff	
	eres esta de	18-, and m grids		Pseudo-global warming approach		RCM runoff		
Meko et al. (2007)	-	-	-	1	II chronologies, upper basin	-	Proxy reconstructions	
McCabe and Wolock (2007)	Estimate 2°C'	-		2	62 HUC8s	-	Percentage adjust- ment based on TWB model and proxy reconstructions	
USBR (2011a) (approach 8°)	-	-	-	1244 and 1000 traces ²	II chronologies, upper basin	-	Proxy reconstructions	

TABLE I. Details of studies used in evaluating future Colorado streamflow.

Emission

scenarios

SRES AIB

CMIP5

RCP8.5

SRES AIB

ACPI BAU

SRES A2

and BI

SRES A2

and BI

SRES A2.

AIB, and BI

SRES AT

Total

projections

49

43

24

3

22

4

112

Spatial

resolution

~2° lat-lon

(~200 km)

~2° lat-lon

(~200 km)*

~2° lat-lon

(~200 km)

1/8° lat-lon

(~12 km)

1/8º lat-lon

(~12 km)

1/8° lat-lon

(~12 km)

1/8° lat-lon

(~12 km)

50-km grids

Type

downscaling

BCSD

BCSD

Constructed

analogs

BCSD

Dynamical

Land surface

representation

GCM P-E

GCM P - E and runoff

GCM runoff

VIC hydrologic model

VIC hydrologic model

VIC hydrologic model

VIC hydrologic model

RCM runof

No. of

RCMs

-

-

-

_

_

No. of

GCMs

19

16^b

12

1

11

24

16

Seager et al. (2007)

Seager et al. (2013)

Milly et al. (2005)

Christensen et al.

Christensen and

Lettenmaier (2007)

Cayan et al. (2010)

USBR (2011a)

(approach 3°)

Gao et al. (2011)

(2004)

. but when does it matter?

Background - Parameter Scaling - Climate Change and Resolution - Conclusions

Using high resolution enabled by super computing to inform low resolution models... bridge the gap



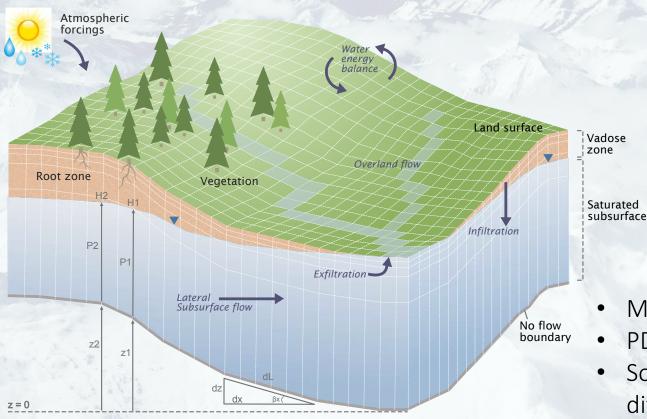
WATER RESOURCES RESEARCH, VOL. 47, W05301, doi:10.1029/2010WR010090, 2011

Hyperresolution global land surface modeling: Meeting a grand challenge for monitoring Earth's terrestrial water

Eric F. Wood,¹ Joshua K. Roundy,¹ Tara J. Troy,¹ L. P. H. van Beek,² Marc F. P. Bierkens,^{2,3} Eleanor Blyth,⁴ Ad de Roo,⁵ Petra Döll,⁶ Mike Ek,⁷ James Famiglietti,⁸ David Gochis,⁹ Nick van de Giesen,¹⁰ Paul Houser,¹¹ Peter R. Jaffé,¹ Stefan Kollet,¹² Bernhard Lehner,¹³ Dennis P. Lettenmaier,¹⁴ Christa Peters-Lidard,¹⁵ Murugesu Sivapalan,¹⁶ Justin Sheffield,¹ Andrew Wade,¹⁷ and Paul Whitehead¹⁸ Received 6 October 2010; revised 21 January 2011; accepted 24 February 2011; published 6 May 2011. High resolution in both SPACE and TIME can bridge observational gaps

- 2. Insight into physical mechanisms driving changes
- 3. Inform predictive and decisionmaking models

We use the integrated hydrologic model ParFlow, coupled to land surface model CLM

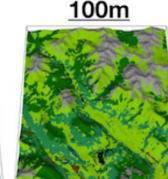


Maxwell (2013); Kollet and Maxwell (2008); Kollet and Maxwell (2006); Maxwell and Miller (2005); Dai et al. (2003); Jones and Woodward (2001); Ashby and Falgout (1996)

- Multi-physics
- PDE-based system
 - Solving the nonlinear diffusion and wave equations
 - Globally
 - Implicitly
 - In parallel

At 1km patterns of landcover, elevation, geology, and soils are decimated

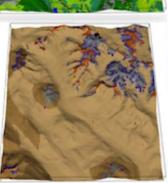
<u>1km</u>



Color Landcover Type Grasslands

Grasslands Evergreen Needleleaf Barren Deciduous Broadleaf Open Shrublands Impermeable Water Bodies Mixed Forest Natural Vegetation Small differences (<5%) in landcover and geology type

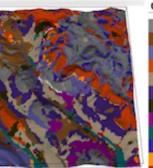






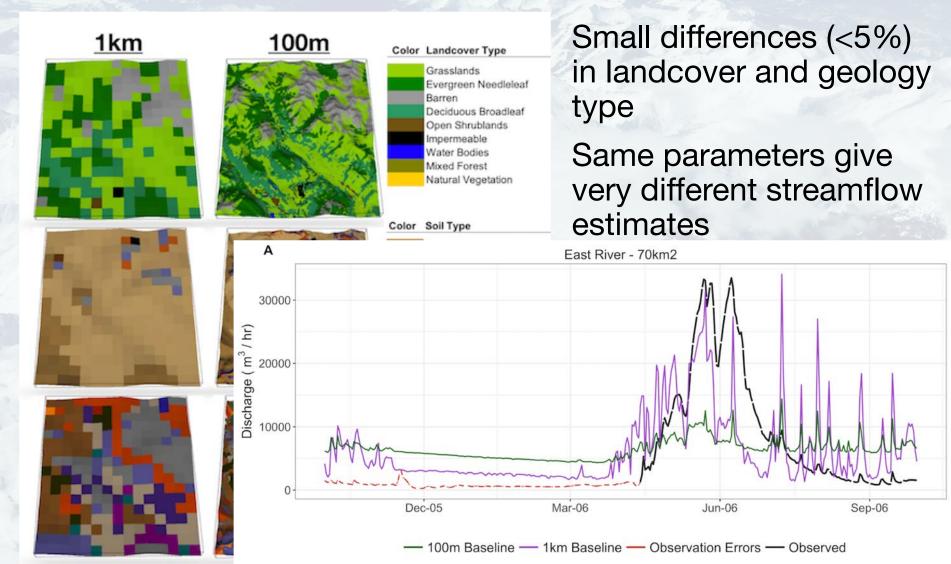
Sandy Loam Loam Clay Loam Outcropped Geology



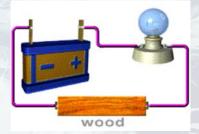


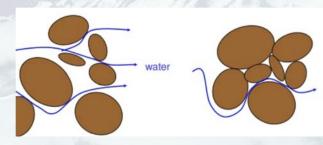
Color Geology Type Crystalline Mancos Shale Talus Uplifted Sedimentary Unconsolidated Glacial Unconsolidated General Landslide Deposits Debris Flow Alluvial Deposits Rock Glaciers Entrada Sandstone Unconsolidated Fan

At 1km patterns of landcover, elevation, geology, and soils are decimated



Hydraulic conductivity- critical parameter for estimating streamflow



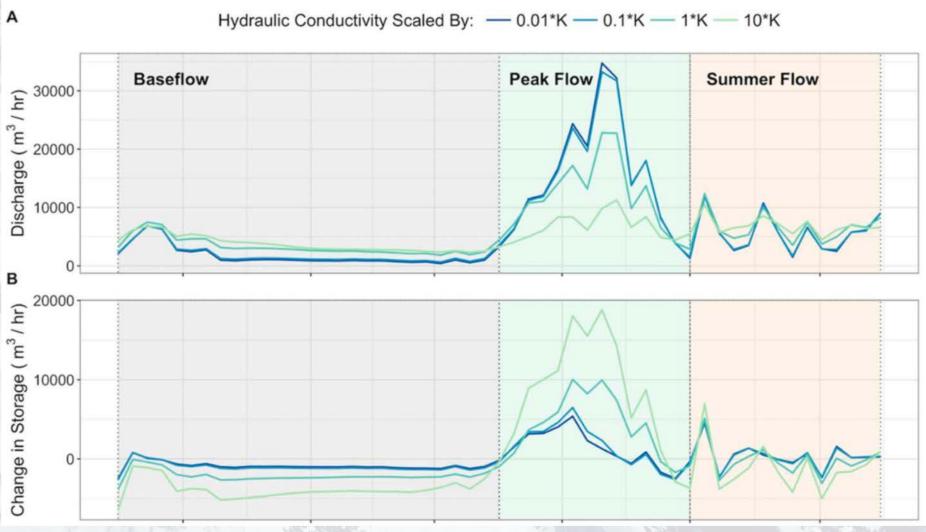


Electrical Conductivity: inherent property of substance explaining how conducive to FLOW Hydraulic Conductivity: inherent property of rock explaining how easily water flows through it.

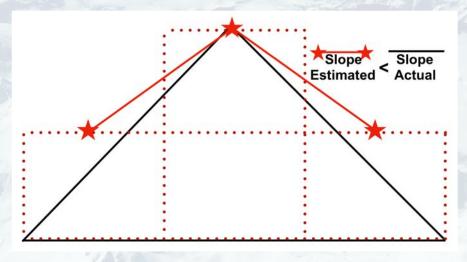
Sensitivity Experiment

_	Hydraulic Conductivity Scaling Factor								
Layer	0.01	0.1	1	10					
Soils	s.01K	s.1K	baseline	s10K					
Geology	g.01K	g.1K	baseline	g10K					
Basement	b.01K	b.1K	baseline	b10K					

Hydraulic conductivity acts as a moderator between streamflow and subsurface flow...

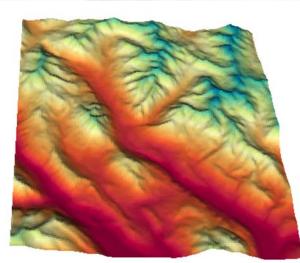


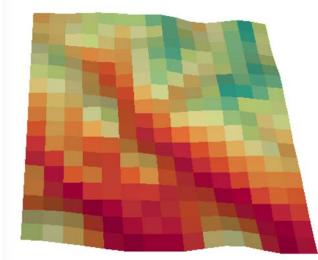
So then what changes between resolutions to cause different flow?



Simple Hydrology:

- Water flows downhill (GRAVITY)
- Resisted by friction (1/HYDRAULIC CONDUCTIVITY)

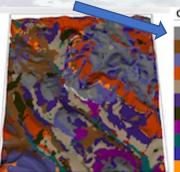




Combine uncertainty in *K* with loss of gradient to make *effective K*

Measured Crystalline K (m/hr):





3.6E-11 < K < 1.08

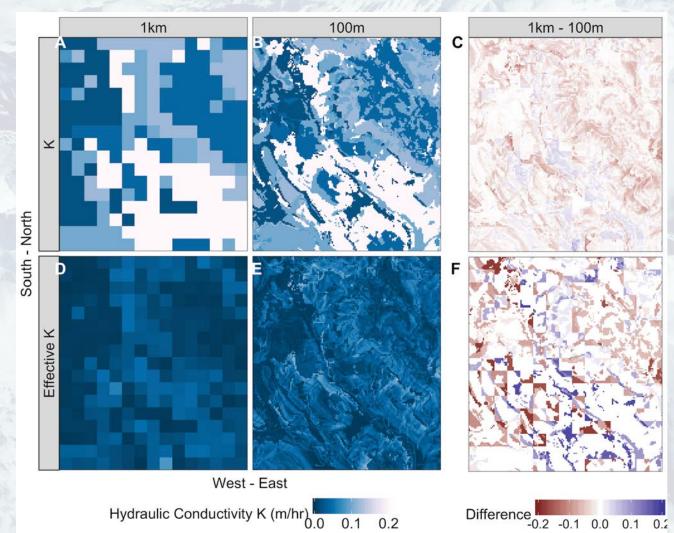


•

Topographic loss of 191m
of elevation... reduces
gravity term in 1km model.
Hydraulic Conductivity (K)
is a highly variable (10 OM)
measured parameter

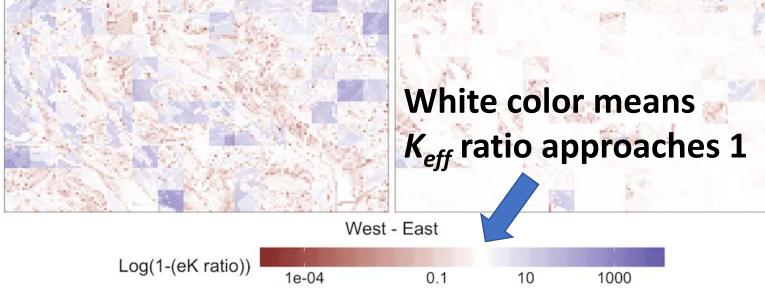
 $K_{eff} = sin\theta K$

Much larger difference between 1km and 100m *effective K* than *K*

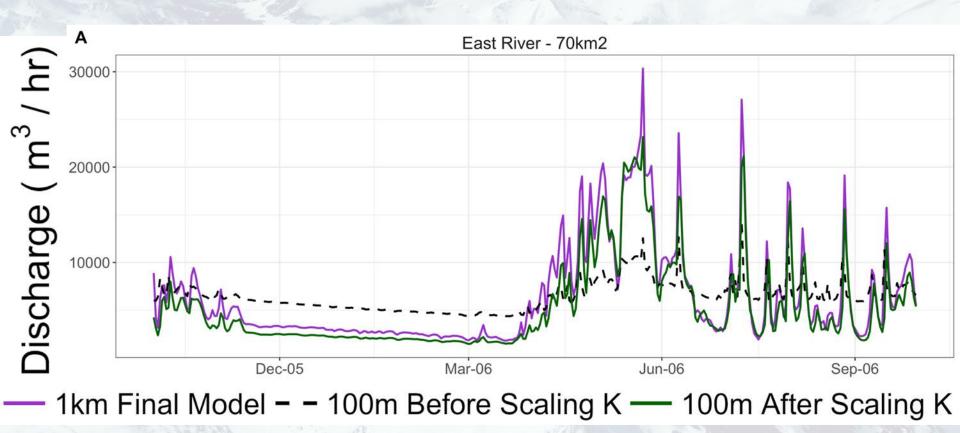


Next step to parameter matching is minimizing the *effective K* ratios between resolutions BEFORE SCALING Minimize K_{eff} ratio



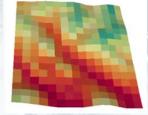


The improvement to matching streamflow between resolutions is dramatic

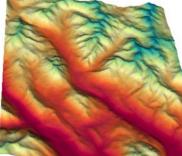


This method can help parameterize hyperresolution models where traditional calibration procedures are limited by computational demand

Tune simple model parameters



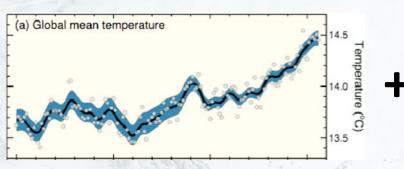
Parameter scaling method

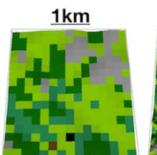


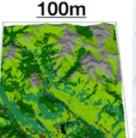
Accurate fine to hyper scale model

= ???

Now we have matching fine and coarse-scale models to examine climate change impacts...







Global climate models and regional hydrologic models are known to perform poorly in the Rockies.

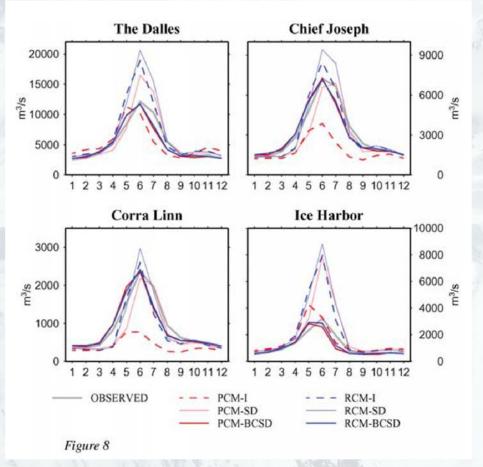
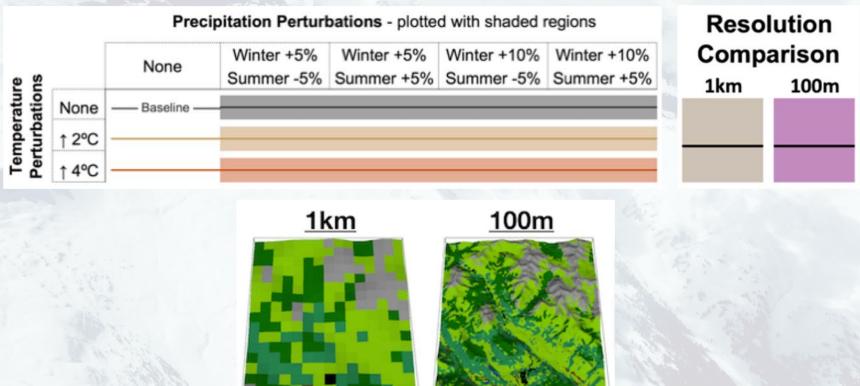


Figure: simulated streamflow for different downscaling methods on PNW snowmelt driven rivers. (Wood et. al. 2003)

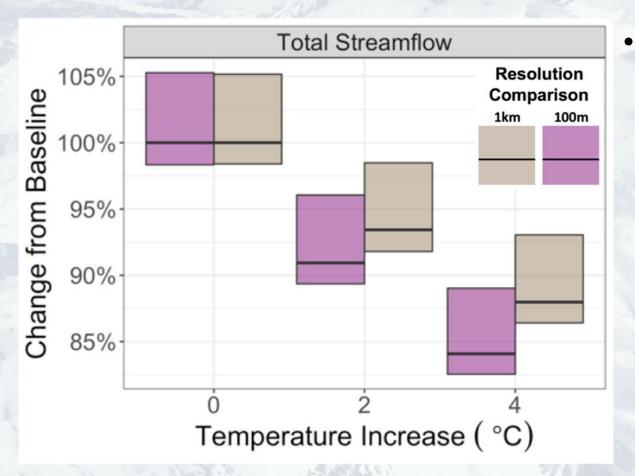
 Begs the question... if our models are more uncertain than climate change... are we able to predict climate impacts?

We compared climate variability with variation in model resolution

30 climate scenarios from Rocky Mountain projections



Results suggest that the coarseresolution models used today may underestimate climate impacts



100m model predicts
a 18% decrease in
headwater
streamflow after 4
degrees of warming...
1km model only
predicts a 12%

To learn more...

UNCERTAINTY MATTERS when MODELING CLIMATE CHANGE in MOUNTAIN HEADWATERS

Lauren Foster¹, Kenneth Williams², Reed Maxwell¹

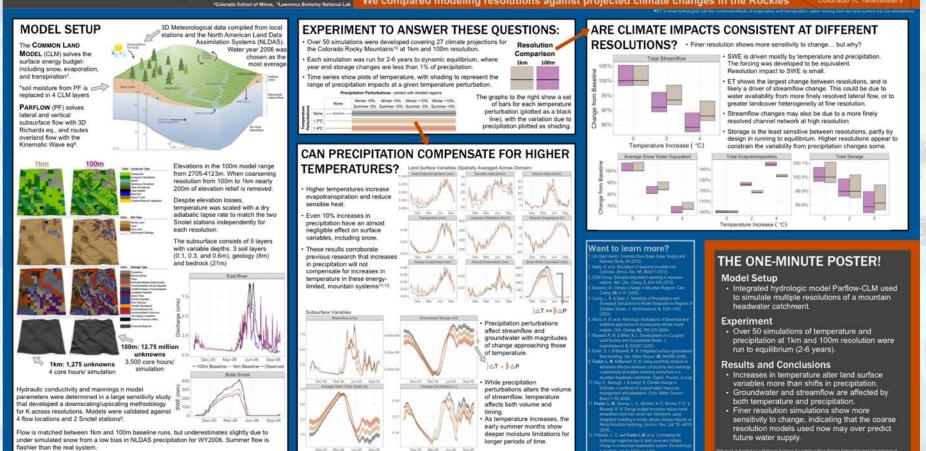
MOTIVATION

- 1 in 10 Americans source water from the Colorado River¹, 85% of which is generated in mountai headwater catchments²
- There is consensus that mountains are especially sensitive to climate and environmental changes³
 Nevertheless, plobal climate models and regional hydrologic models are known to perform poorly in the
- regions⁺, casting doubt on water supply forecasts for the next century

Yikes, but what to do?

hese models are at coarse resolutions (1-50km) that flatten topography and simplify complex geology... → in mountains, these simplifications alter predictions of snowpack, groundwater, streamflow, and ET*

We compared modeling resolutions against projected climate changes in the Rockies



he models are here, nea

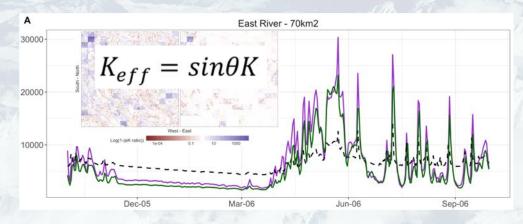
that is representative of

Colorado R. headwaters

Crested Butte, CO. The Ea

Background - Parameter Scaling - Climate Change and Resolution - Conclusions

Why systems like BW? Computational Demand...



CAN PRECIPITATION COMPENSATE FOR I

AINTY MATTERS when LING CLIMATE CHANGE in

MOUNTAIN HEADWATERS

Parameter scaling study:

- 36 simulations
- 60,000 core hours

Climate uncertainty study:

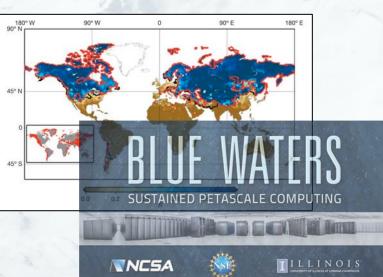
- 54 simulations
- 90,000 core hours

Total: 150,000 hours (not counting mistakes or experiments that were not included in papers)

ARE CLIMATE IMPACTS

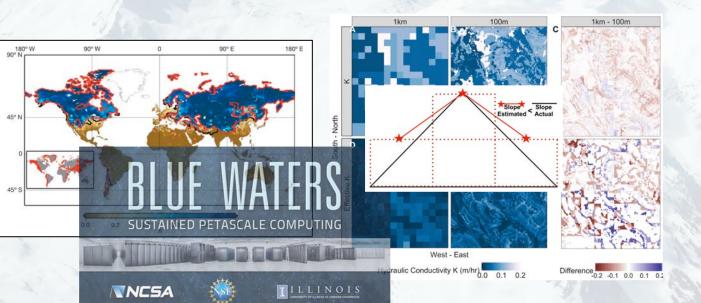
Conclusions

 Hyper-resolution models and HPC systems can help us understand important, complex systems like mountains



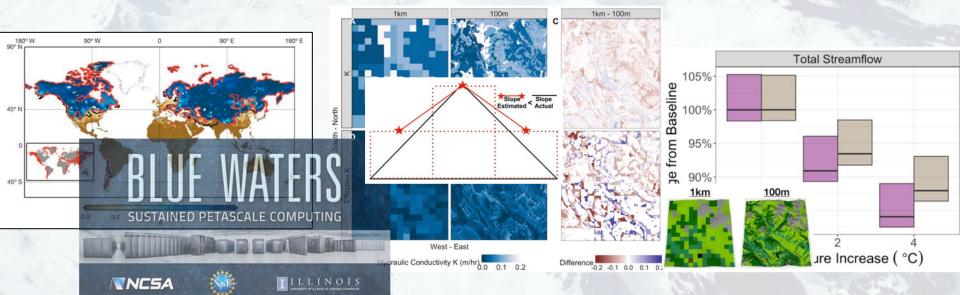
Conclusions

- Hyper-resolution models and HPC systems can help us understand important, complex systems like mountains
- Model interrogation and development are critical to getting the right answers for the right reasons... i.e. model sensitivity and parameter estimation!



Conclusions

- Hyper-resolution models and HPC systems can help us understand important, complex systems like mountains
- Model interrogation and development are critical to getting the right answers for the right reasons... i.e. model sensitivity and parameter estimation!
- How we build and use our models is as important as the climate changes they are built to detect, so we must be thoughtful about our results and their implications.



Thank you!!

an sel

Watershed

Function

Scientific



COLORADOSCHOOLOFMINES



Climate [change]?



Questions?