

Snow depth spatial structure from hillslope to basin scales

Jeffrey S. Deems

NASA JPL Airborne Snow Observatory
CIRES National Snow and Ice Data Center
CIRES/NOAA Western Water Assessment
University of Colorado

Ian W. Bolliger

College of Natural Resources
UC Berkeley



University of Colorado
Boulder

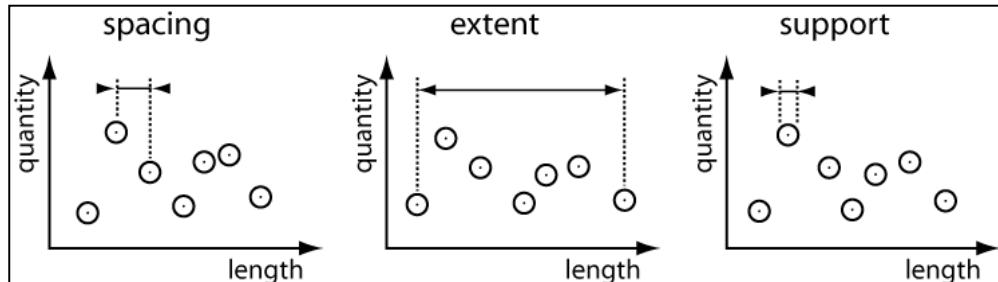


ENERGY & RESOURCES GROUP

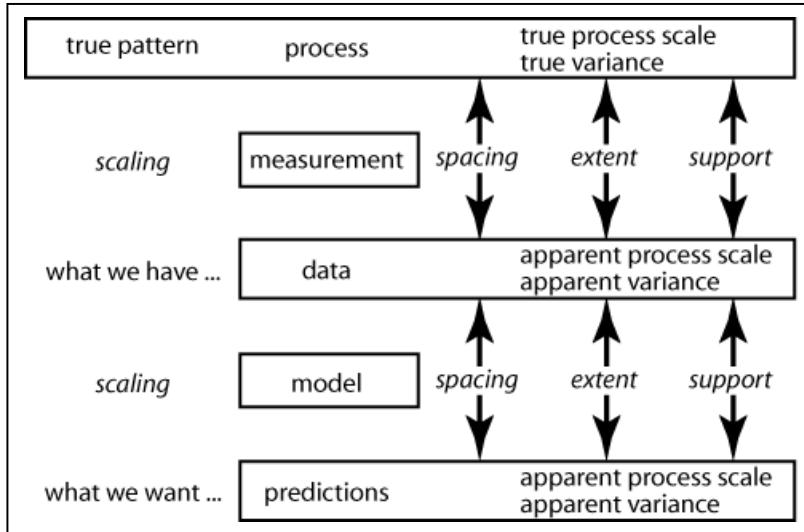
snow depth variability & scale

knowledge of process scales is key to:

- understanding impacts of measurement & model scale choices
- identifying optimal scales of measurement & modeling

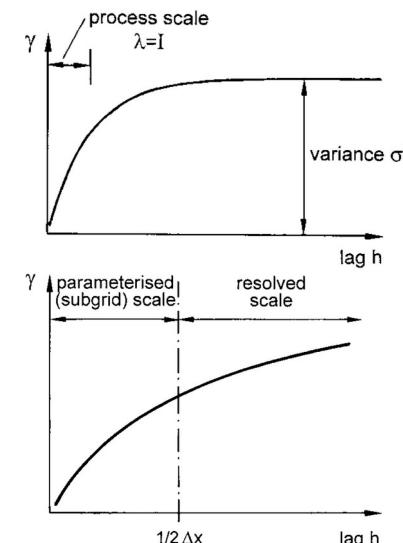


from Blöschl (1999)

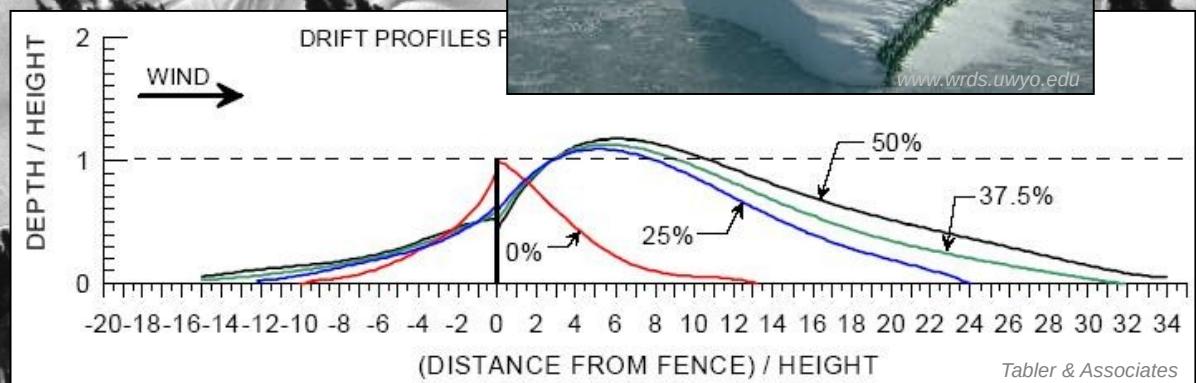
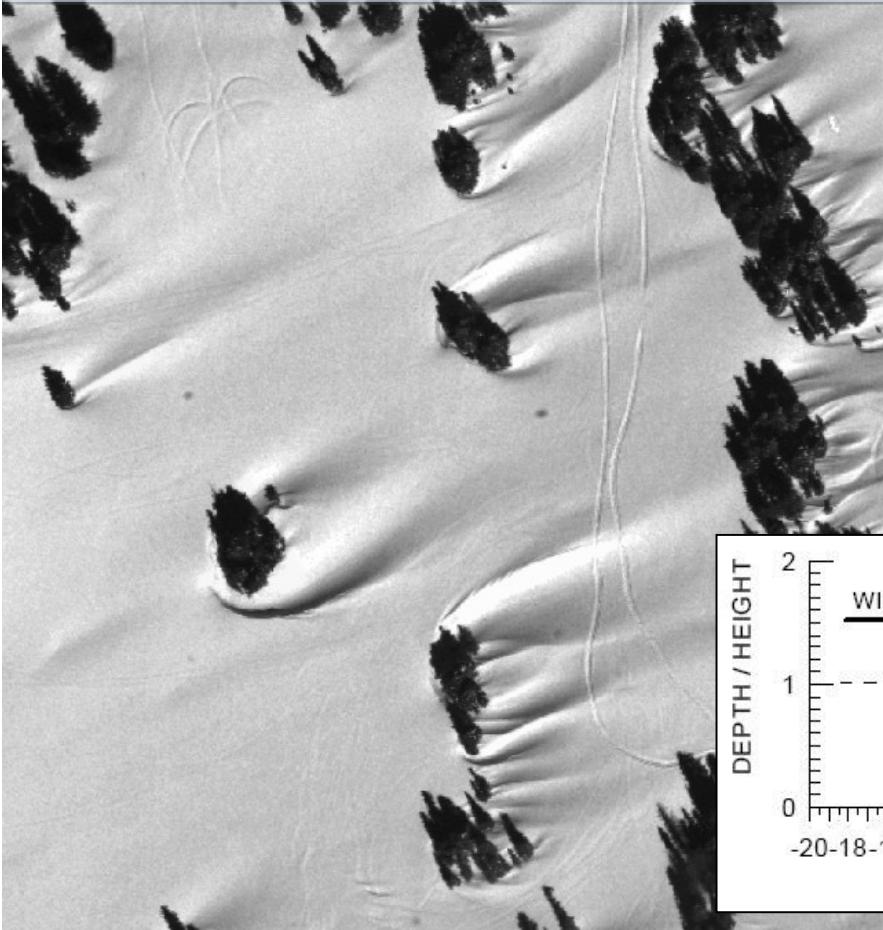


from Blöschl (1999)

- scale triplet describes the scale characteristics of the measurement or model
 - process scales should inform measurement/model designs/choices
- measurement/model scale choices often dictated by data availability, technology, \$\$
- process scale knowledge is needed to assess bias/error due to scaling



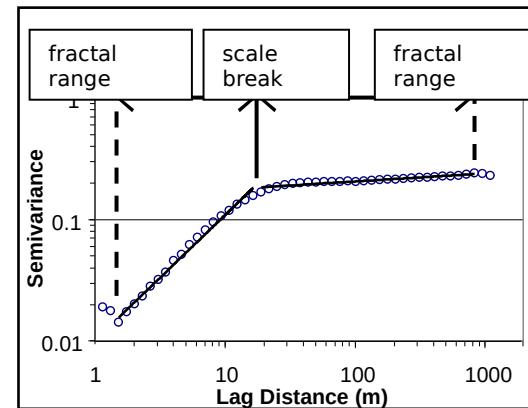
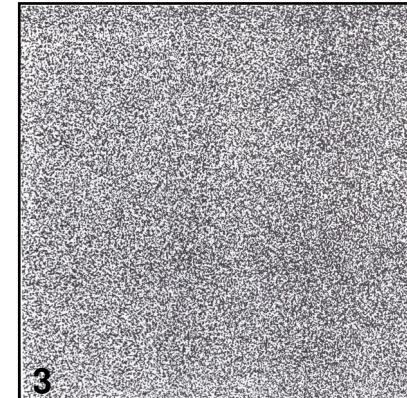
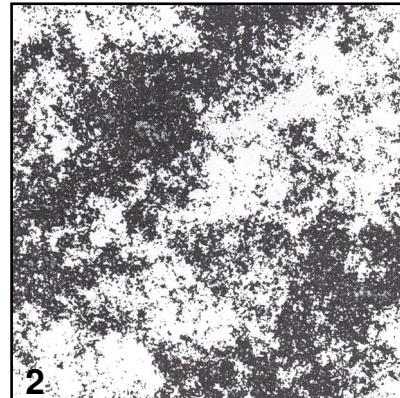
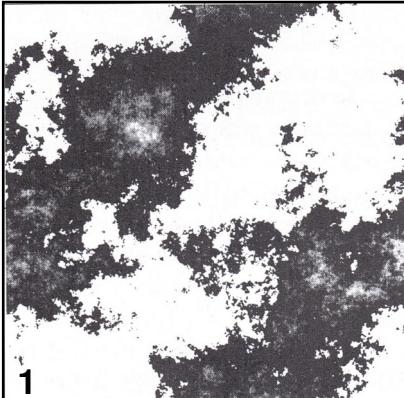
snowdrift geometries across scales



spatial complexity & fractals

- self-similarity
 - similar structure or statistical properties across scales
- geometric vs. statistical fractals
 - natural: specific scale ‘windows’
- fractal dimension
 - consistent with Euclidean/topological dimensions
 - measure of relative roughness, complexity
 - describes balance of small & large scale patterns/processes

*fractal geometry can characterize
spatial variability of the snow cover*



fractal ranges:

- scale ranges of consistent process relationships

scale break:

- change in driving processes or interaction

scale ramifications:

- process
- measurement
- model

1. $D = 2.1$

Some small-scale detail, but generally smooth variation;
large scale processes dominate

2. $D = 2.5$

Rich complexity; balance of small scale & large scale processes

3. $D = 3.0$

White noise – spatially random;
no autocorrelation

prior examples

Shook & Gray 1996

Variogram fractal structure of depth transects

Blöschl 1999

Fractal geometry of SCA over wide range; D as variability metric

Kuchment & Gelfan 2001

Fractal character of long depth transects

Deems et al. 2006, 2008

Variogram fractal depth distributions over 1m-1km range; wind redistribution/vegetation linkage; interannually consistent.

Trujillo et al. 2007, 2009

Spectral scaling exponents of lidar depth transects; wind redistribution/vegetation linkage; interannually consistent.

Schirmer & Lehning 2011

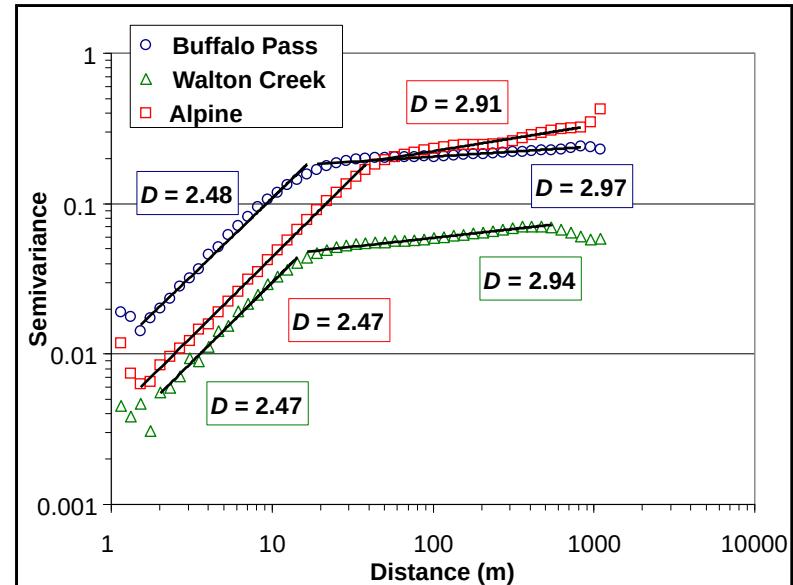
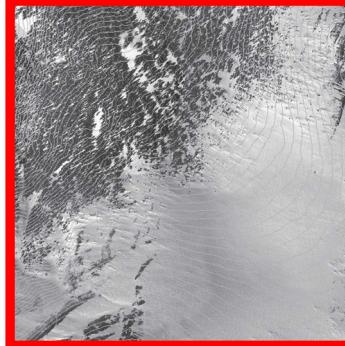
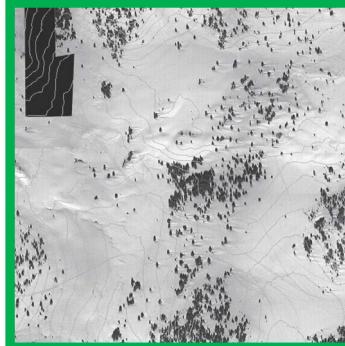
Variogram fractal depth distributions over 0.1m-0.5km range; wind redistribution linkage; temporal evolution; interannually consistent.

Lehning et al. 2011

Fractal dimension as roughness index & predictor variable

Mott et al., 2011

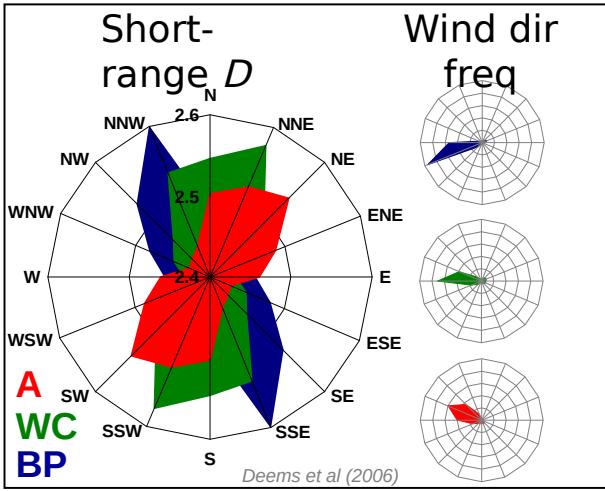
Fractal dimension variation with wind direction in lidar & simulated snow depth



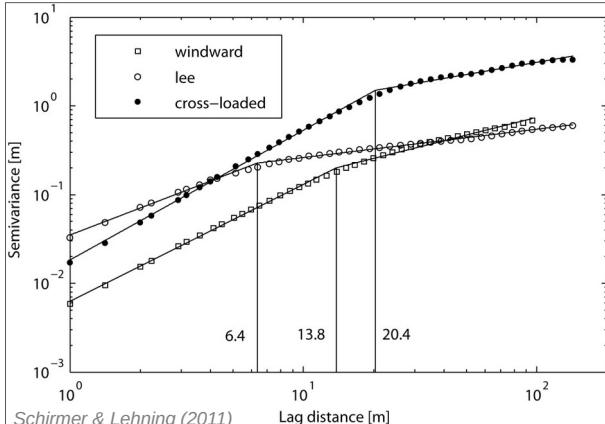
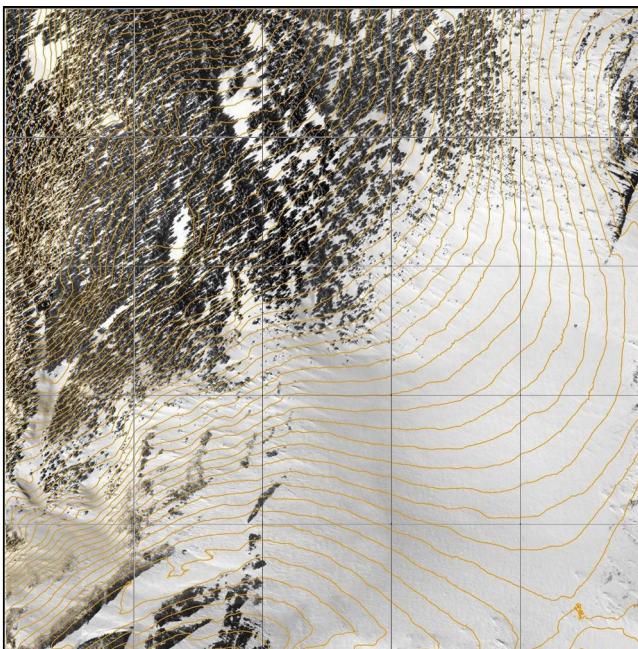
Deems et al (2006)

- short-range D values ~ 2.5 indicate balance in process scales
- scale breaks:
 - WC: 15 m
 - BP: 16 m
 - A: 40 m
- high D beyond scale break – bulk of spatial structure is occurring at shorter distances
- directional variograms suggest process link to wind redistribution

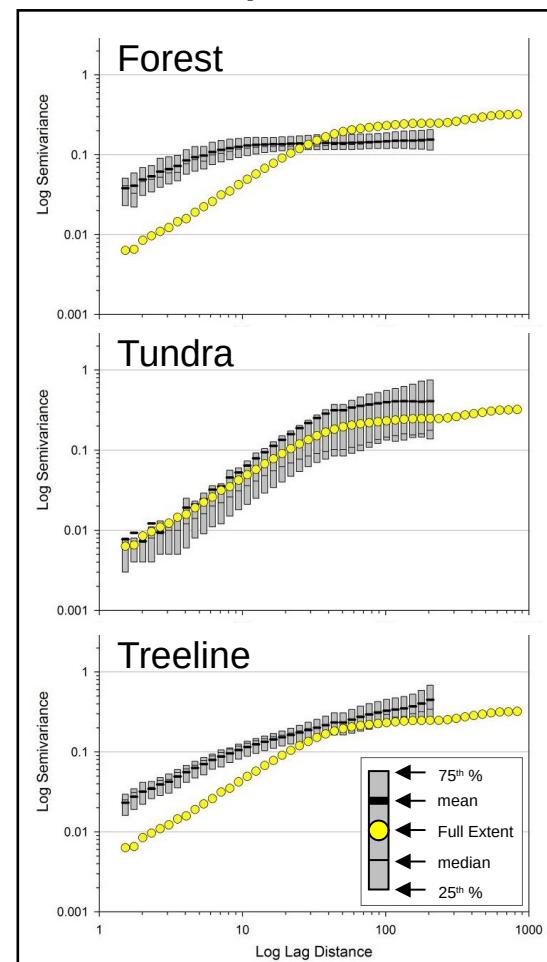
linkage to physical processes



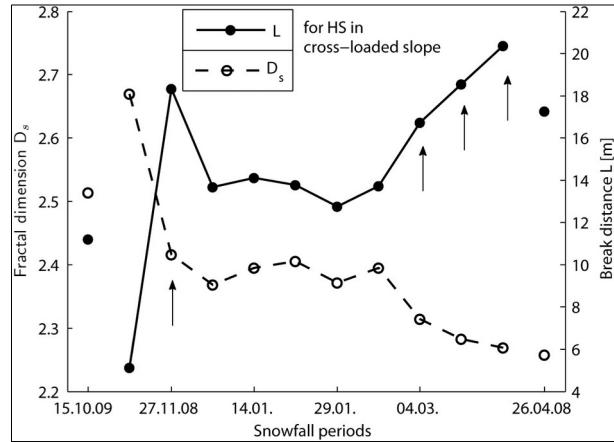
- forest: shorter scale breaks than tundra areas
- tundra subsets dominate full extent pattern
- treeline subsets show complex mix of patterns



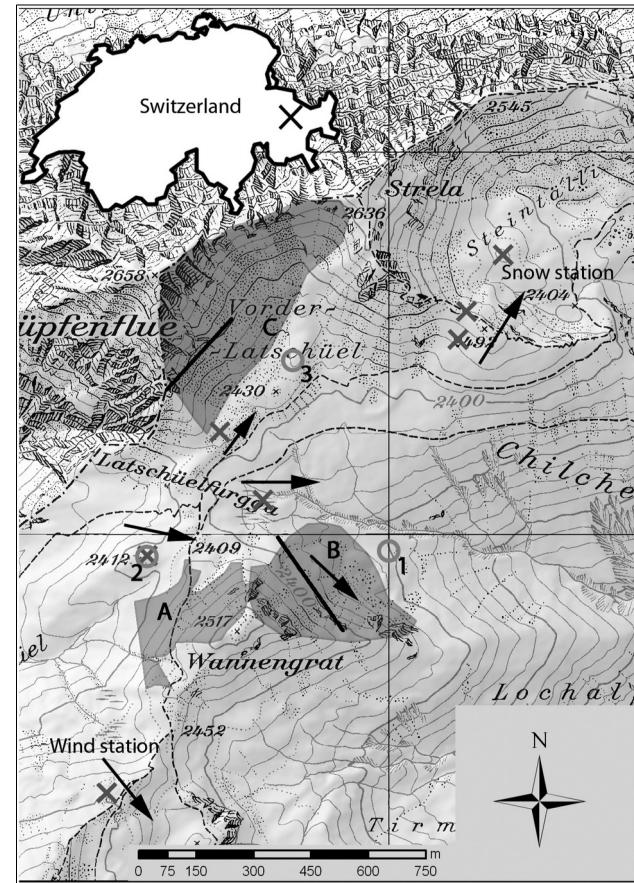
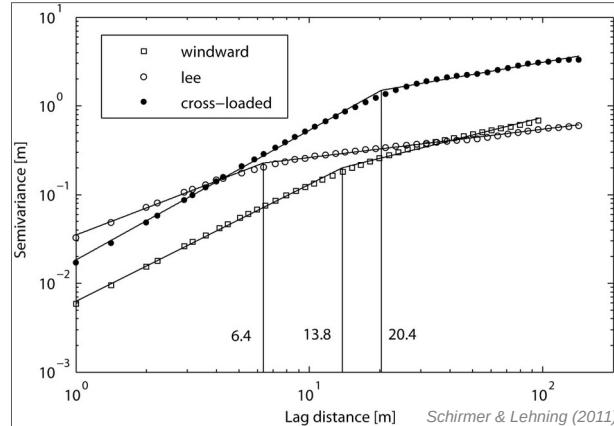
200 m Alpine Subsets



fractal dimension & scale break as metrics



- D characterizes balance of process over a specific scale range
- SB bounds that scale range
- are these parameters stationary across the landscape &/or through time?



Airborne Snow Observatory: Tuolumne river basin

- 1200 km²
- 1100-4000m ASL
- montane to alpine
- maritime snow climate

ASO data

- lidar DEM & tree height
- CASI land cover class
- lidar snow depth maps:

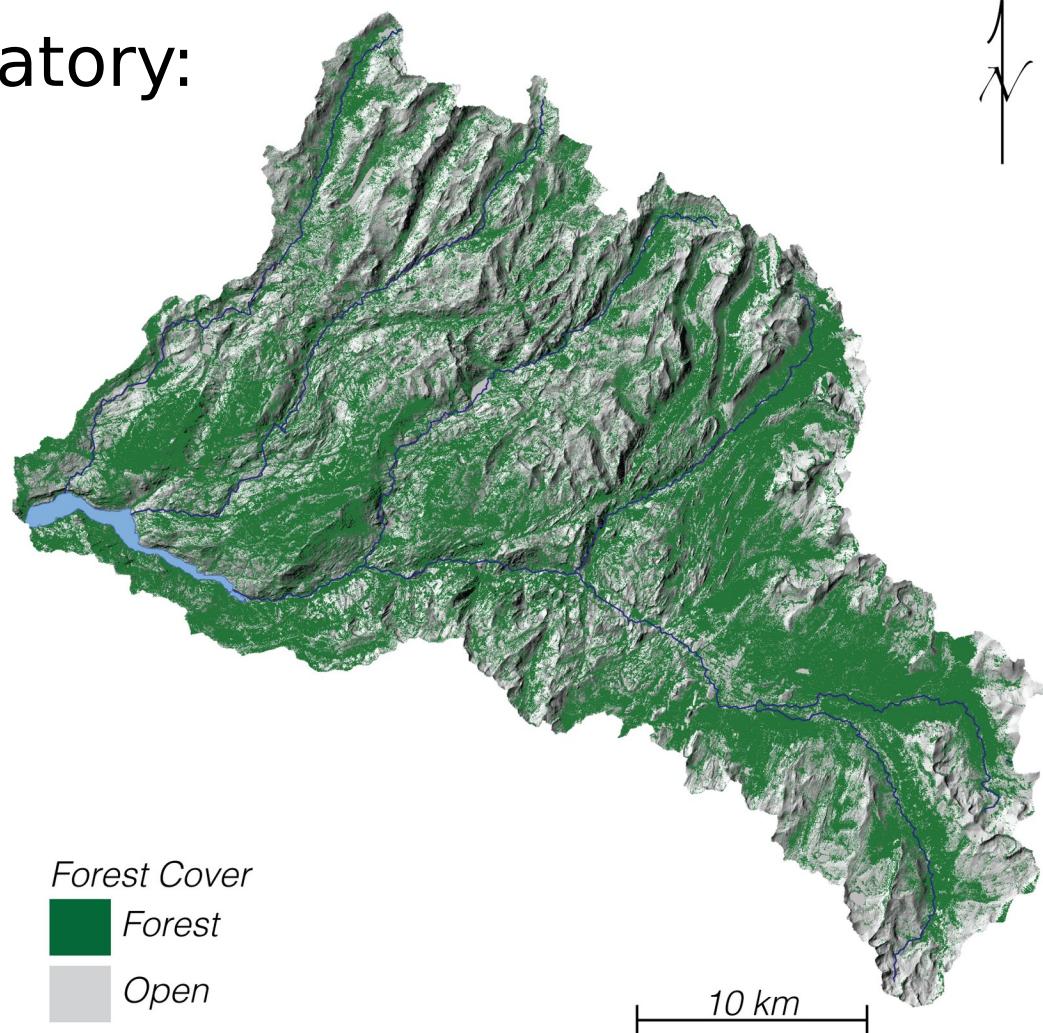
2013: 6 flights

2014: 11 flights

2015: 10 flights

2016: 12 flights

2017: 8 flights

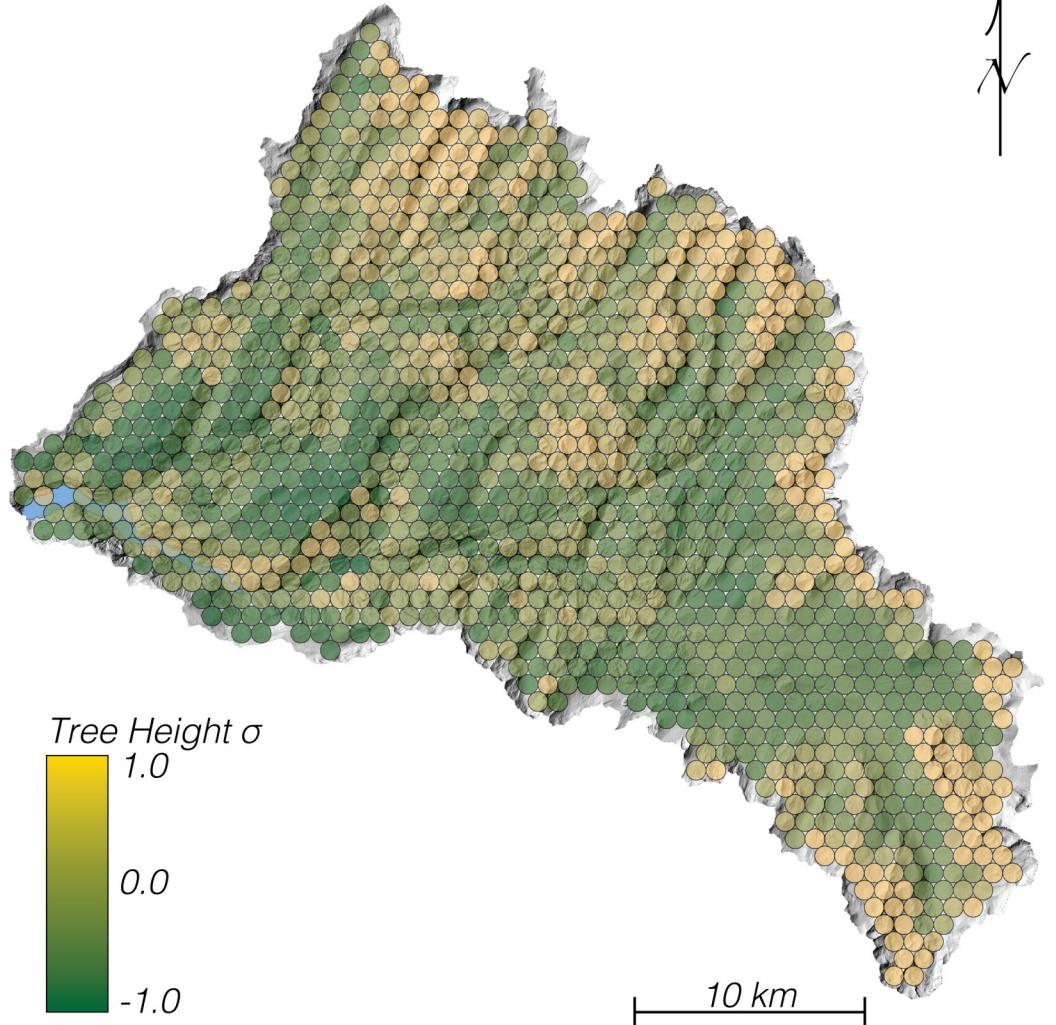


methods

N

- fill basin with 1km circles
- static physiographic parameters
 - DEM (μ, σ)
 - slope (μ, σ)
 - northness (μ, σ)
 - tpi (μ, σ)
 - veg height (μ, σ)
- snow depth variograms
for each circle, for each ASO flight
 - **zero depths not included
- calculate scale break SB &
fractal dimensions D
for each circle
 - segmented power law fit

$$\gamma(r) = a r^b \quad D = 3 - \frac{b}{2}$$

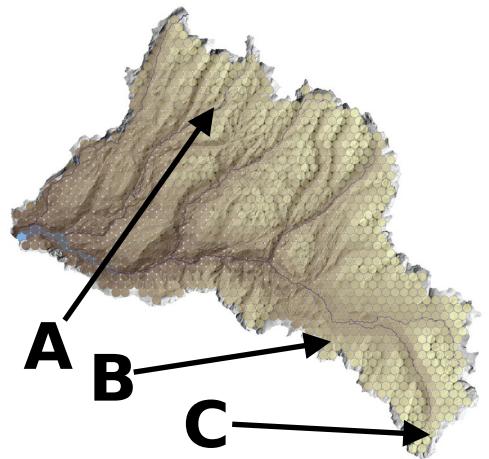


example results: 2017

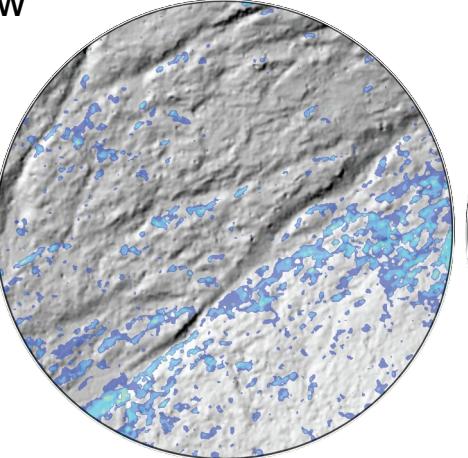
A: forested; shallow snow

B: treeline; strong
geomorphic controls

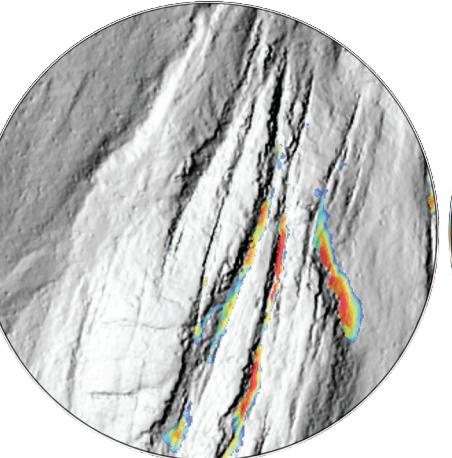
C: alpine; deep snow



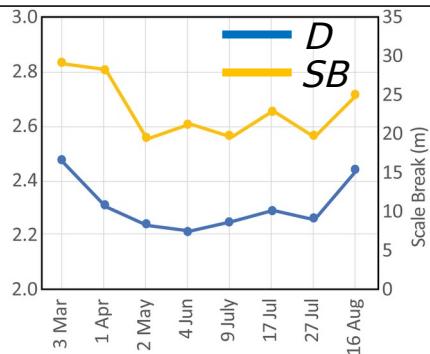
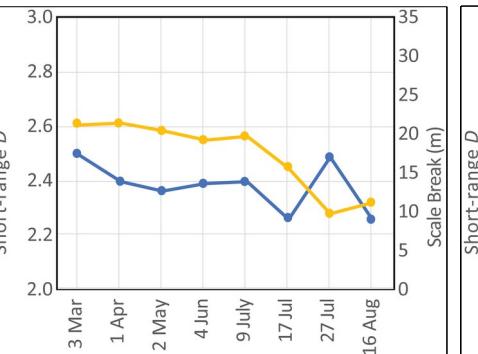
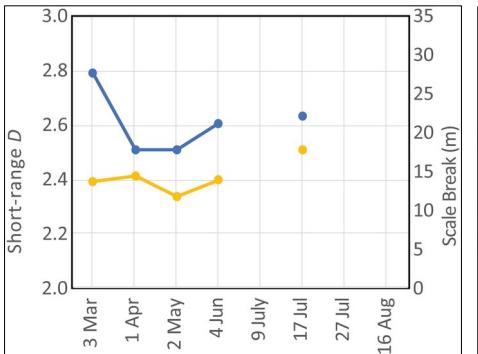
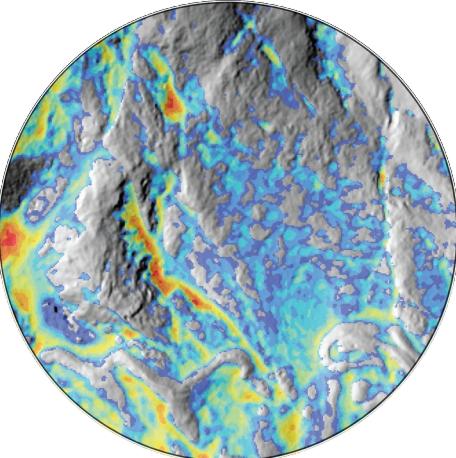
A



B



C



A vertical colorbar titled "Snow Depth (m)" with a scale from 0.2 (blue) to 12+ (red). The color gradient transitions through green and yellow towards red.

temporal patterns

Short Range D

- decrease in spring
- smoother/less-variable depth as melt progresses
- stronger effect in wetter years

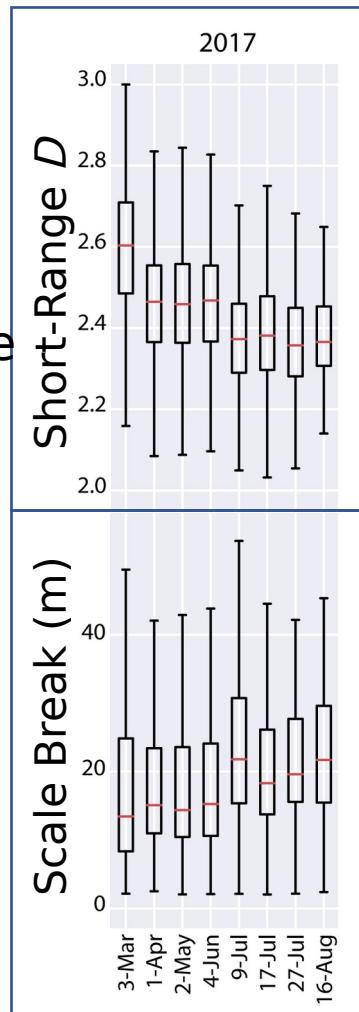
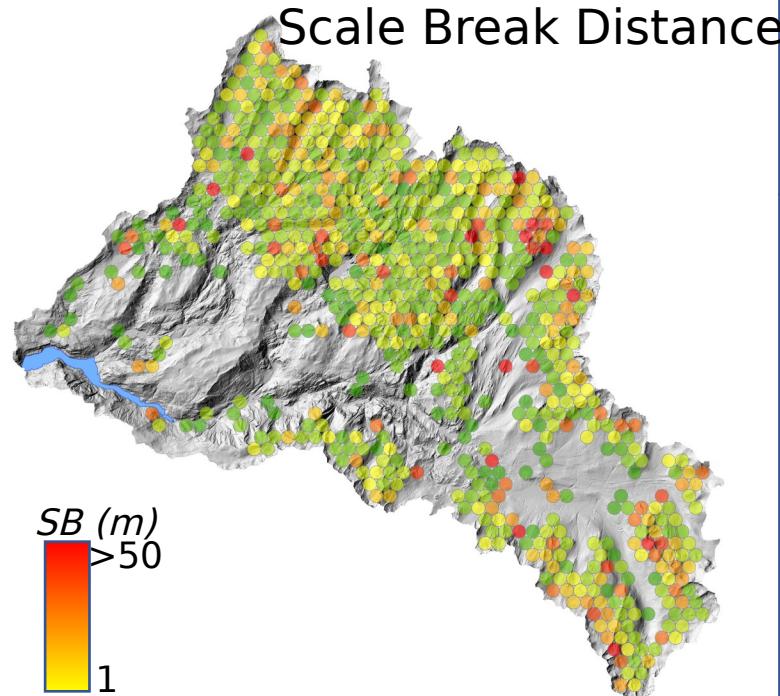
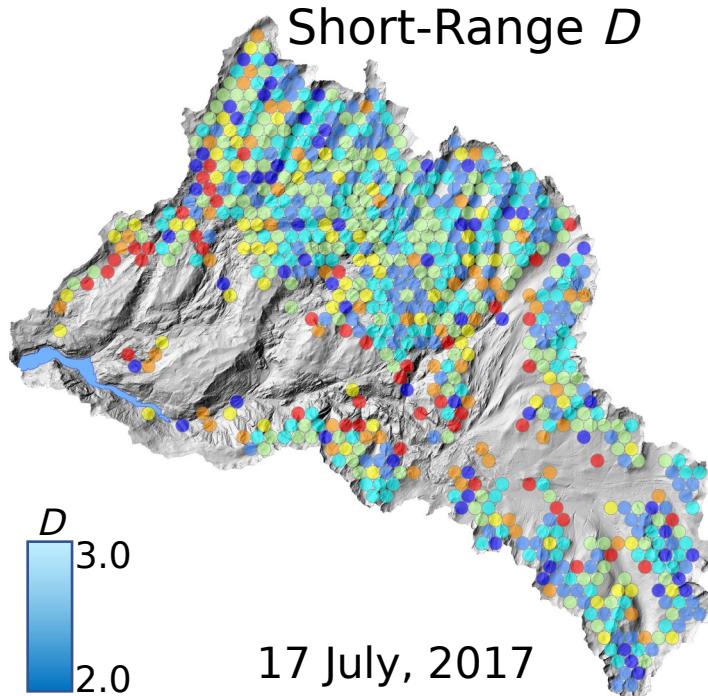
Scale Break

- increase through season
- melt processes consistent over wider range than accumulation processes?



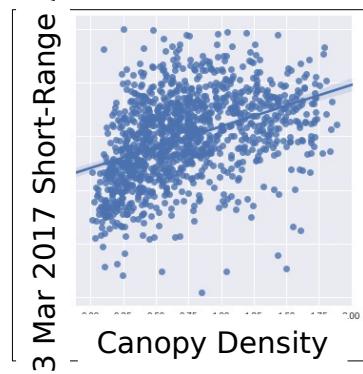
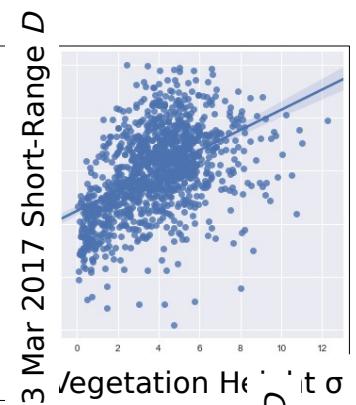
spatial & temporal patterns

- distribution of Short-range D & scale break
- $D \downarrow$ & $SB \uparrow$ over time
- not spatially uniform trajectories

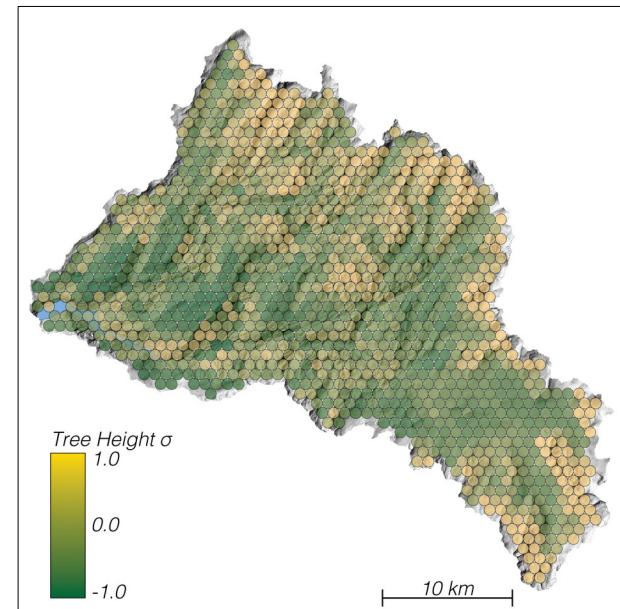
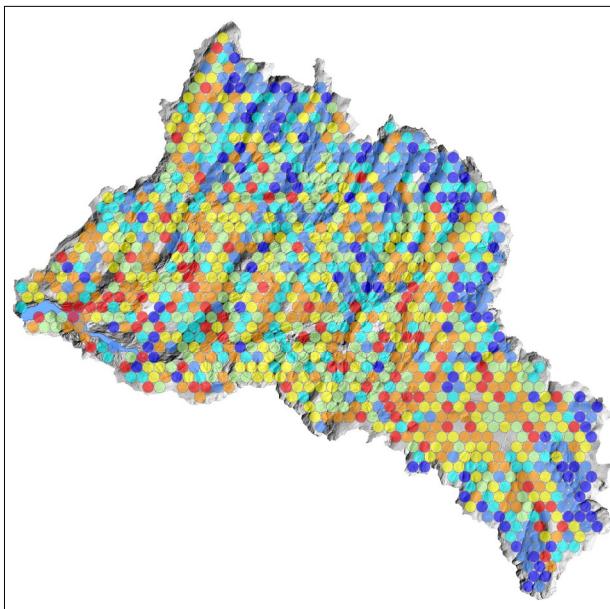


physiographic influences – bulk properties

- positive relationship between short-range D & veg height σ
- similar with canopy density

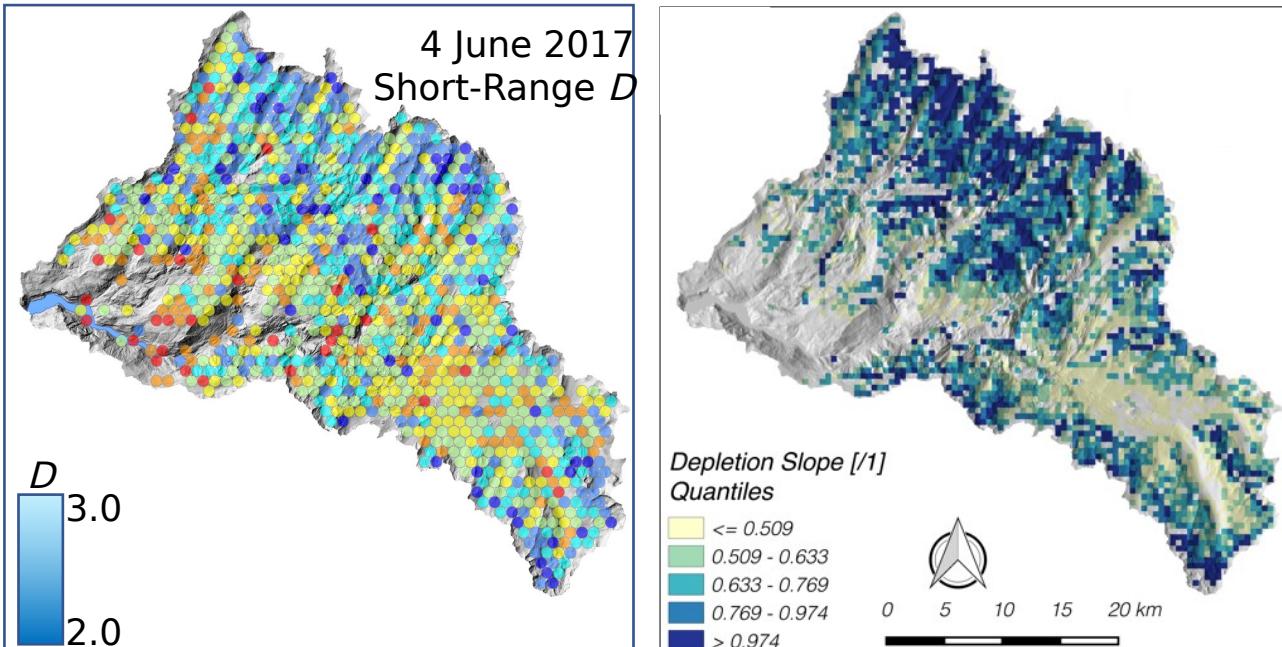
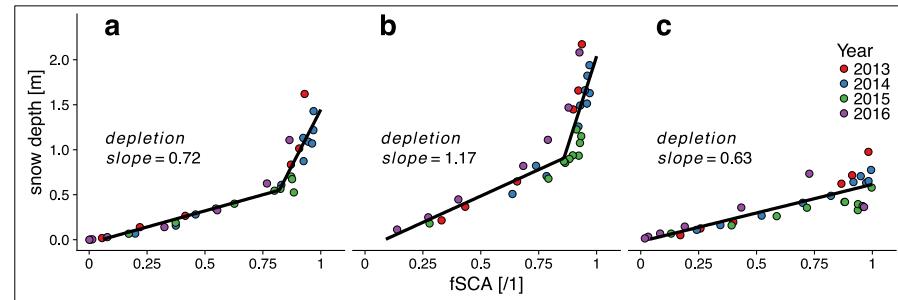


3 Mar 2017 Short-Range D Vegetation Height σ



fractal dimensions, depletion curves, & terrain roughness

- snow depth/fSCA depletion curve slopes vary with terrain roughness
- roughness variation occurs over a range of length scales
 - 10^{-2} m – rocks, talus
 - 10^0 m – trees, outcrops
 - 10^1 - 10^3 m – geomorphic features
- D describes balance of roughness scales



Schneider, D., N.P Molotch, J.S. Deems, T.H. Painter (2017) Analysis of topographic controls on depletion curves derived from airborne lidar snow depth data, *Geophys. Res. Lett.*, *in review*.

summary & ongoing work

initial results

- D , SB vary across the domain
 - relationships to bulk physiographic parameters; vegetation strongest
- temporal evolution through the snow season
 - accumulation & melt processes seem to show different scale properties

ongoing

- comparison to wind parameters & physiographic scaling
- other sites with different physiography, vegetation, snow climate
- different max variogram extents
- SCA thresholds

Friday sessions: *Quantifying Spatial and Temporal Variability of Snow and Snow Processes*
C51E & C52B – Oral C53B - Posters

Ian Bolliger C53B-1023: *Spatiotemporal Variability in Topographic and Vegetative Controls on Basin-Wide Snow Distribution in the Tuolumne River Basin*