

NEW SNOW METRICS FOR DECISION SUPPORT

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Snow Affects Multiple Sectors

Water
Energy
Forest



Fish & wildlife

Agriculture



Reservoir operations

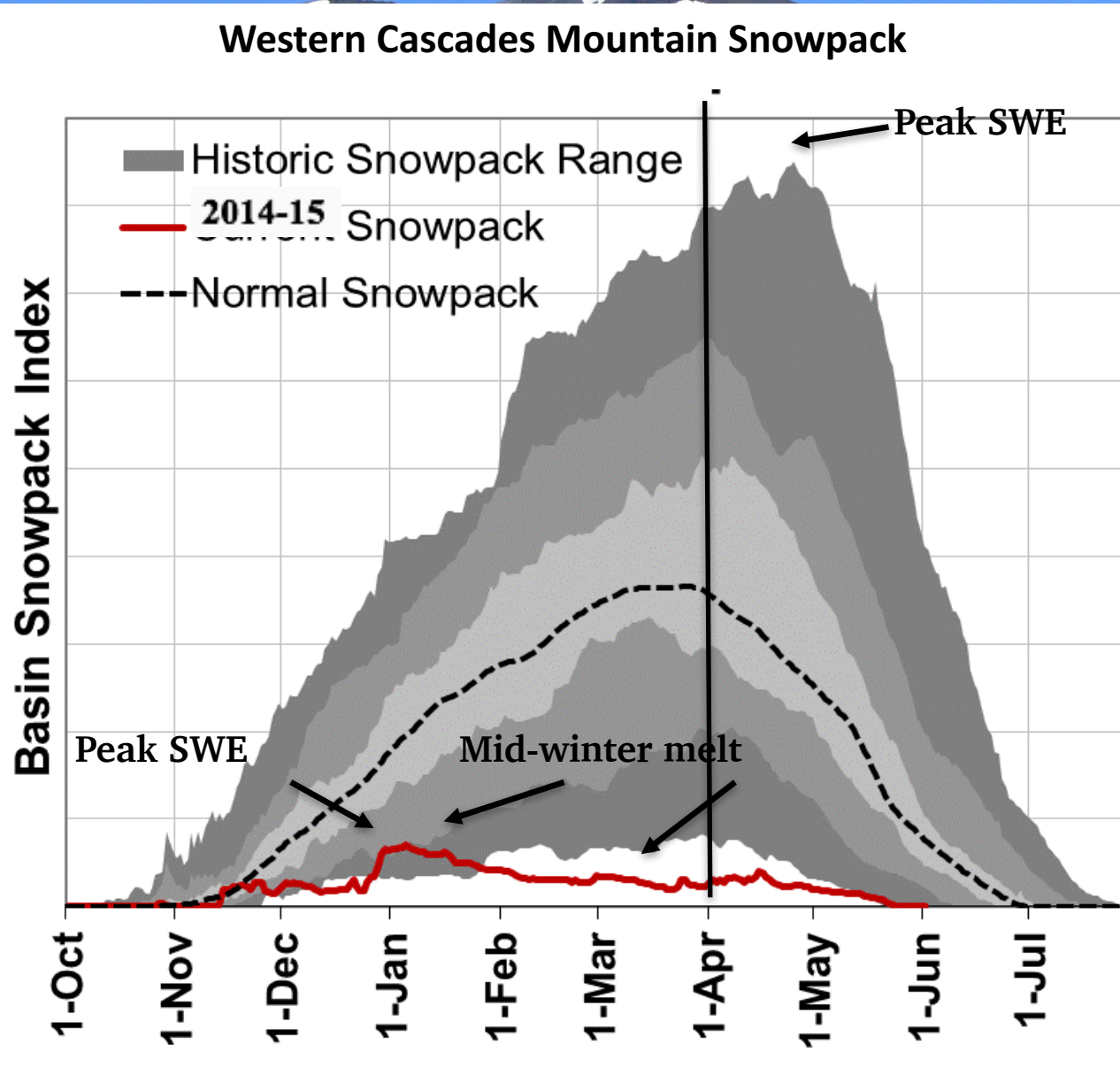
Transportation



Recreation & tourism



Why do we need new snow metrics?



New Snow Metrics - 1

- **Snow Cover Frequency (SCF)**
- **Snow Disappearance Date (SDD)**

Created using 500-m, daily data from NASA's Moderate Resolution Spectroradiometer (MODIS) for the period 2000-present, globally

New Snow Metrics - 2

- **Snowstorm
Temperature**

Created using daily data
from NRCS Snow
Telemetry sites (SNOTEL)
800+ sites across the
western US, 1984-
present



For each site, we identify the min/max/mean daily temperature on days when the snow pillow indicates that snowfall has occurred

New Snow Metrics - 3

- **Frequency of a Warm Winter (FWW)**

- **“At-risk” Snow**

Created using daily, downscaled climate model output and the VIC hydrologic model

- 4 km spatial resolution
- 40+ climate models
- historical and future periods (1950 – 2099)

- lower-48 US only

At-risk snow

- Safe snow
- At-risk snow
- Negligible snow

rcp85 - historical

- Safe (unchanged)
- Safe to at-risk
- At-risk (unchanged)
- Safe to negligible
- At-risk to negligible
- Negligible snow (unchanged)

How we do this: Cloud-computing and Visualization using Google Earth Engine

Google Earth Engine

[FAQ](#)

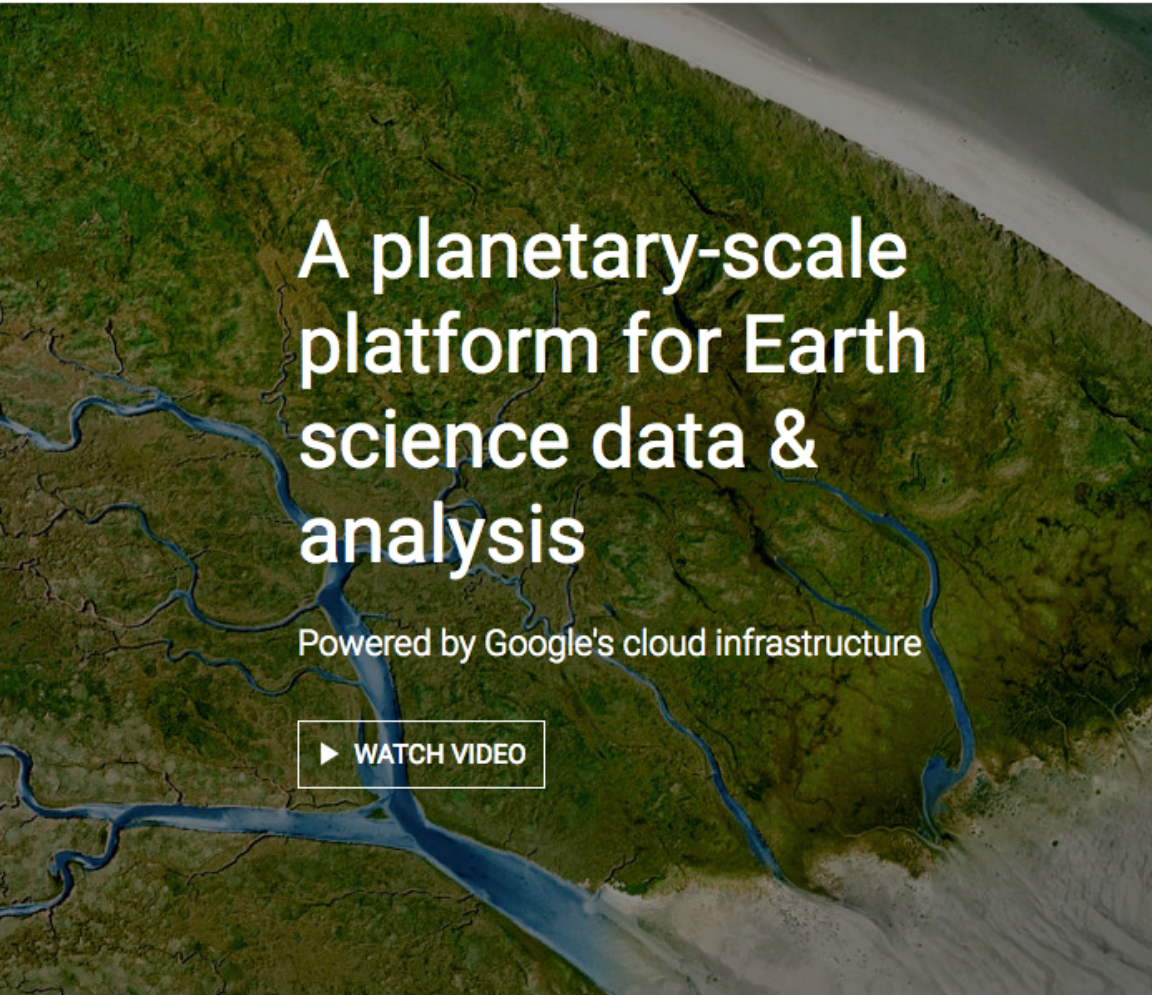
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A satellite map of a river delta, likely the Amazon, showing a complex network of blue water channels branching out from a larger body of water into a green, forested land area. The text 'A planetary-scale platform for Earth science data & analysis' is overlaid in white. Below it, in smaller white text, is 'Powered by Google's cloud infrastructure'. At the bottom left, there is a white rectangular button with a play icon and the text 'WATCH VIDEO'.

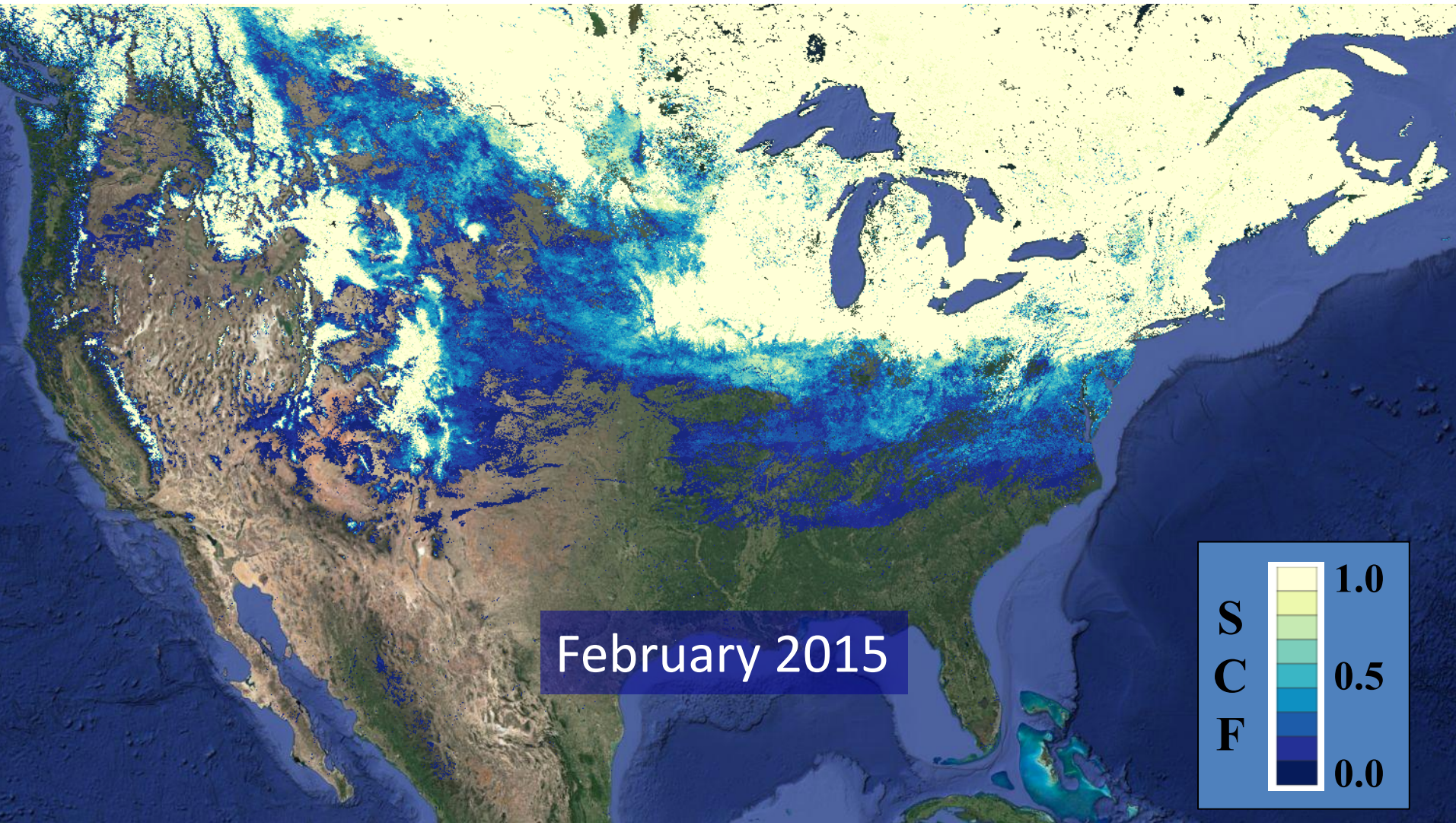
A planetary-scale
platform for Earth
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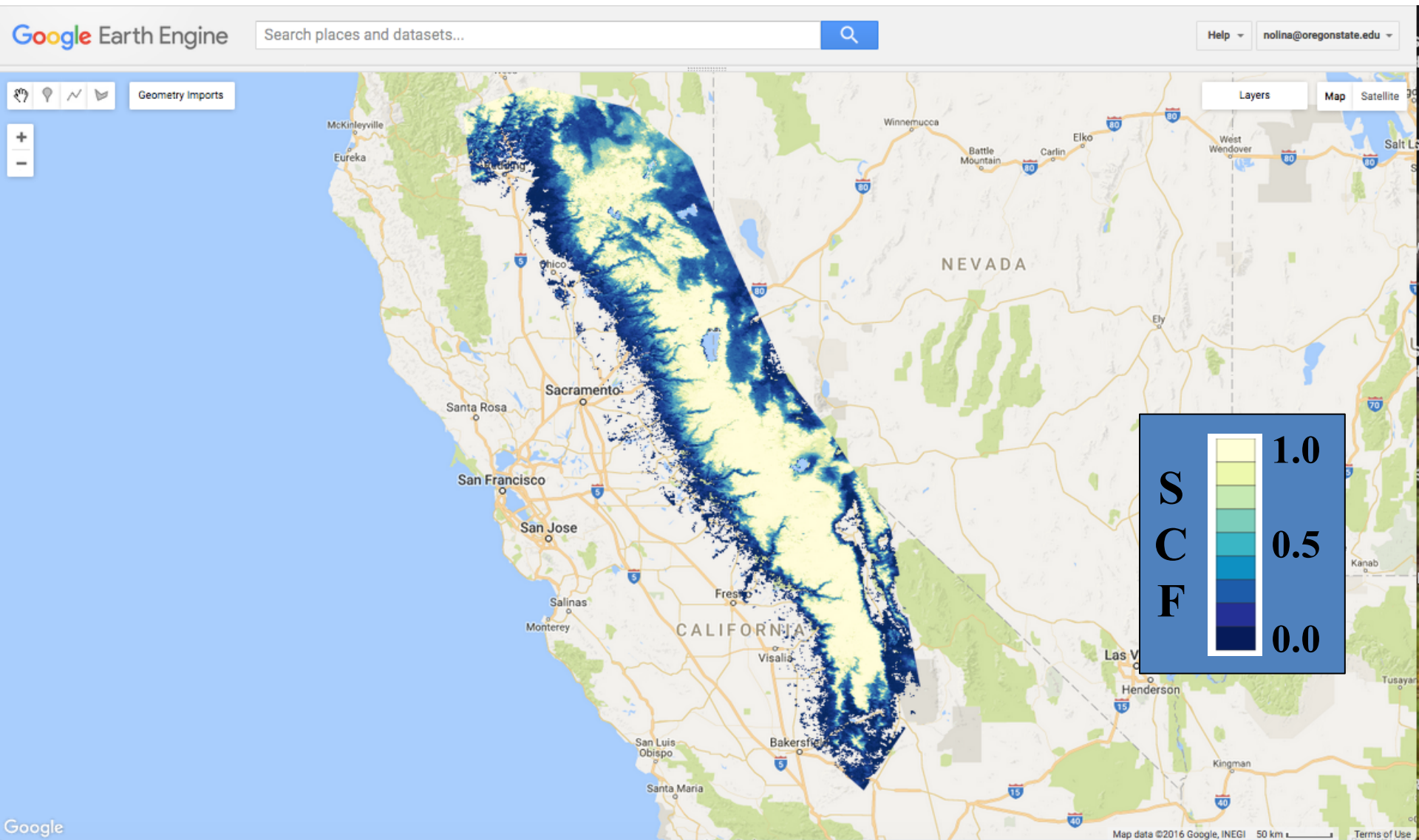
▶ WATCH VIDEO

- Allows users to interactively explore spatial and temporal snow metrics
- Can export the data in non-proprietary data formats (geotiff images and csv files)

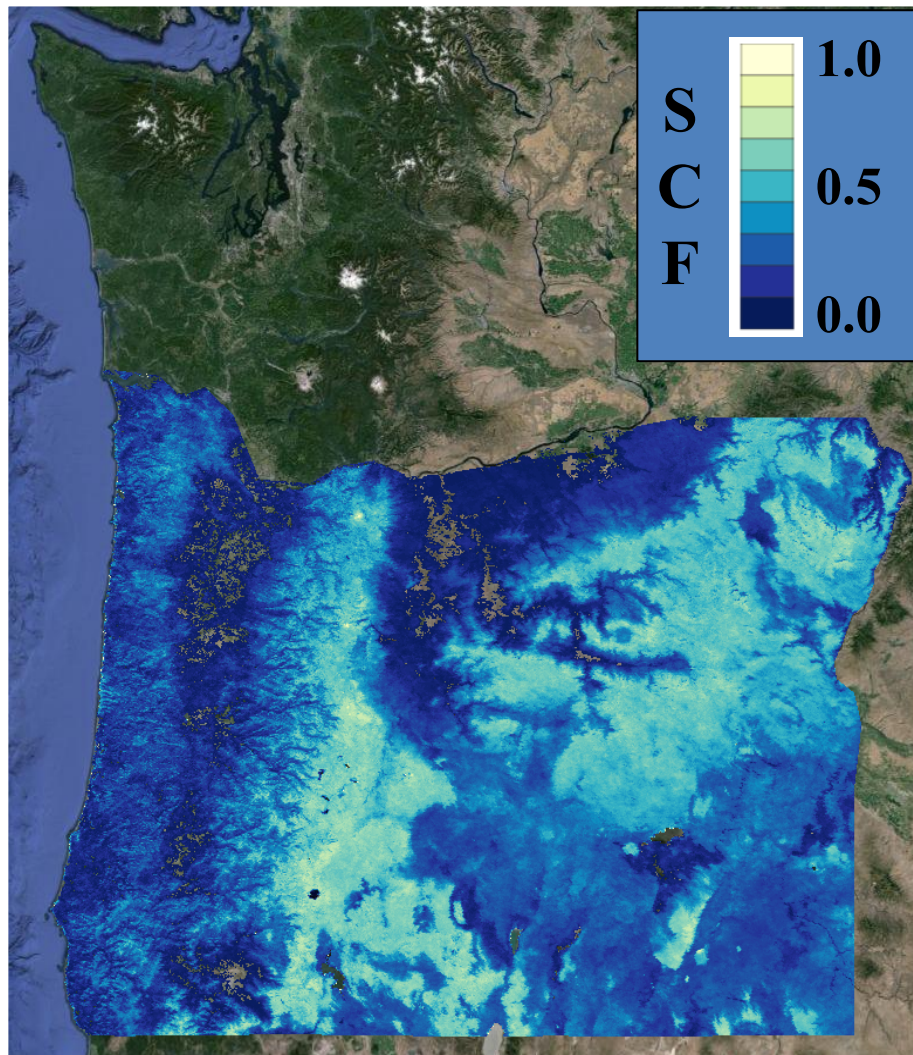
Example of Snow Cover Frequency over a 28-day period



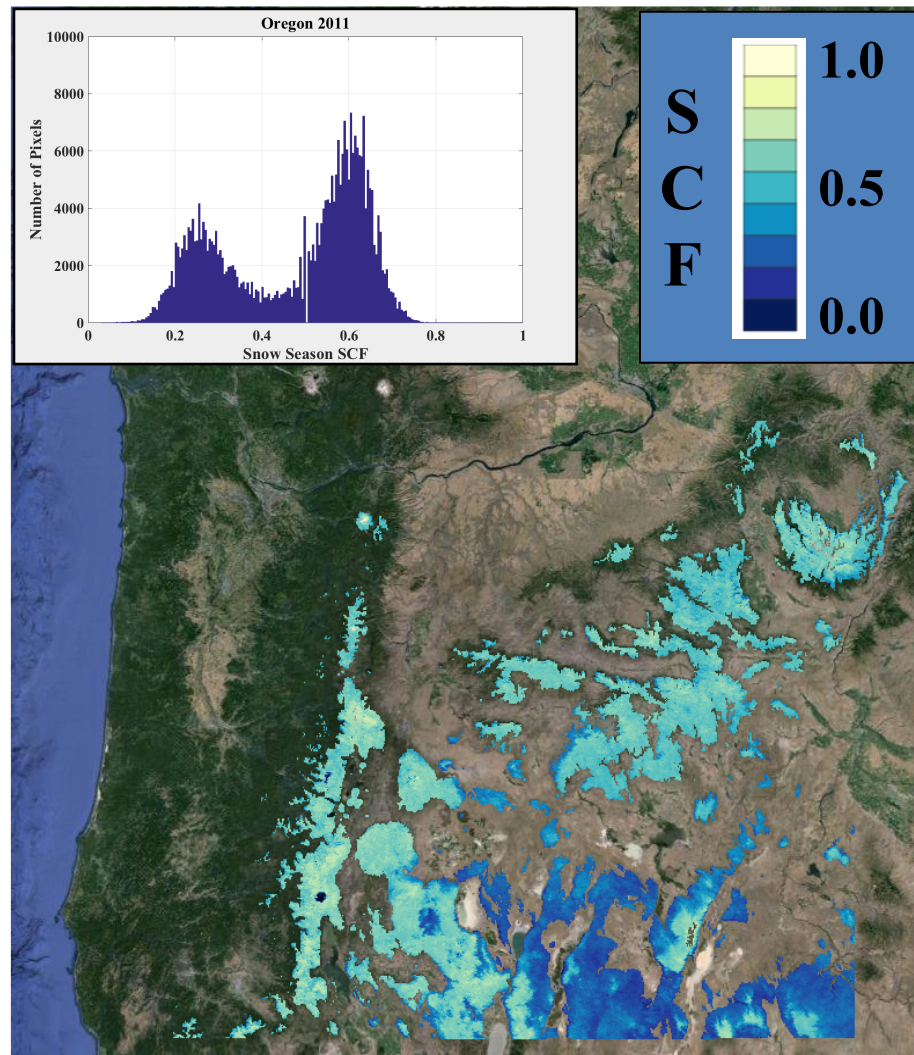
Polygon subsetting: draw (or import shapefile)



Oregon Oct 1 – Apr 1, 2011



1500 m – 2500 m elev. band



Examples of Applications

1. Streamflow forecasting (SCF + hydro model)
2. Post-fire vegetation recovery and snow (SCF)
3. Identifying temperature thresholds for snow vs. rain (Snowstorm temperature)
4. Potential impacts of future climate change on watersheds (“At-risk” snow, FWW)
5. Potential impacts of future climate change on ski areas (FWW)

Streamflow forecasting in a snow-dominated watershed using SCF as input to a simple, statistical model

La Laguna, Chile is a snow-dominated but drought-prone region



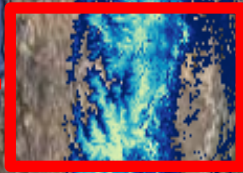
High value agriculture depends on irrigation



Sproles et al. 2016
Water Resources Management

SCF for the Chilean Andes

30°S



Input data

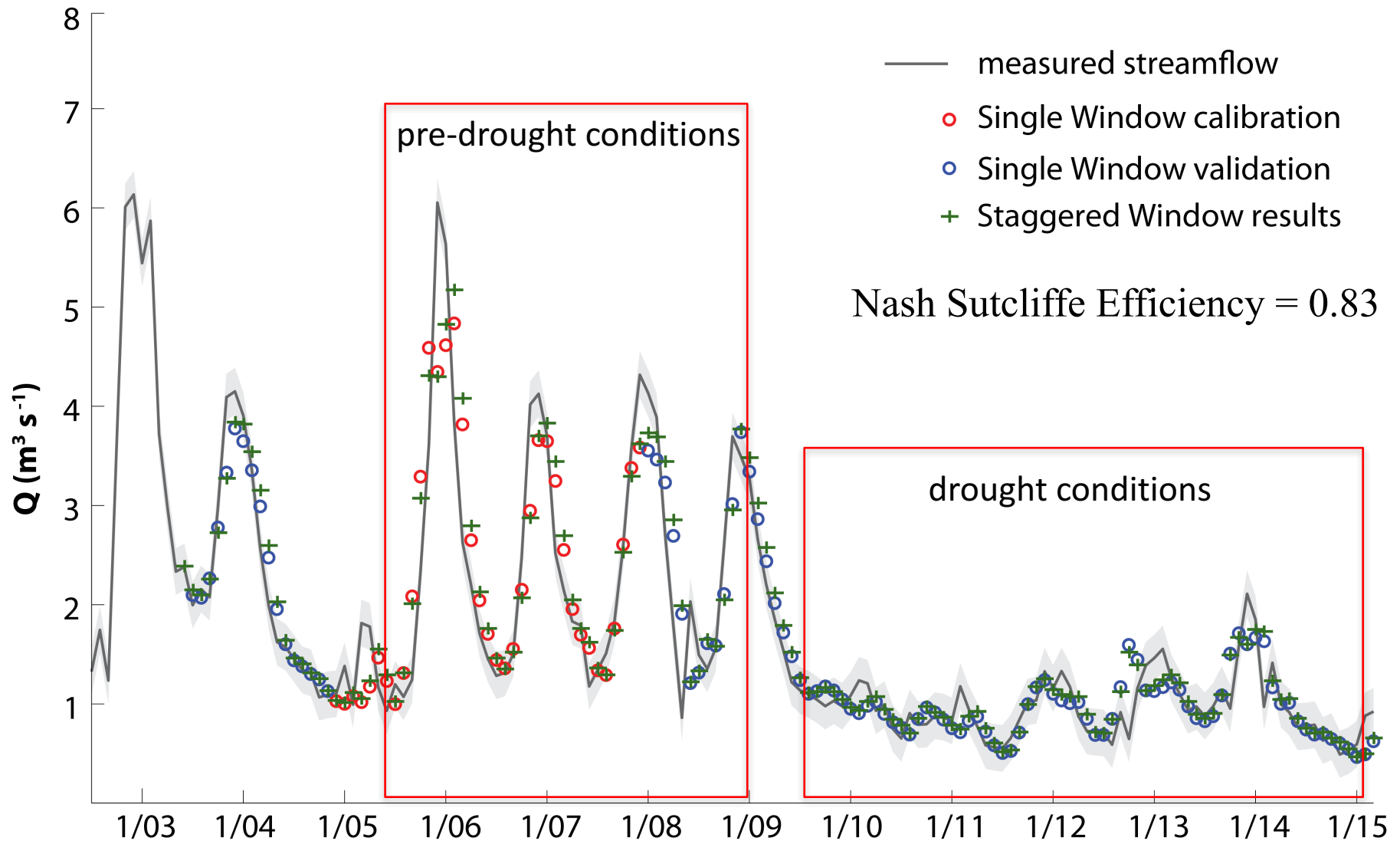
- Prior monthly mean streamflow
- Prior monthly **SCF** for the watershed

Output data

- Next month's mean streamflow

La Laguna, Chile

Measured and Modeled Monthly Mean Streamflow



See: <http://mountains.ceoas.oregonstate.edu/snowcloud/>

<http://mountains.ceoas.oregonstate.edu/snowcloud/>

The OSU Mountain Hydroclimatology Research Group

Snow – Ice – Climate – Hydrology – Communities

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■ SnowCloud

¡Haga clic aquí para la versión en español!

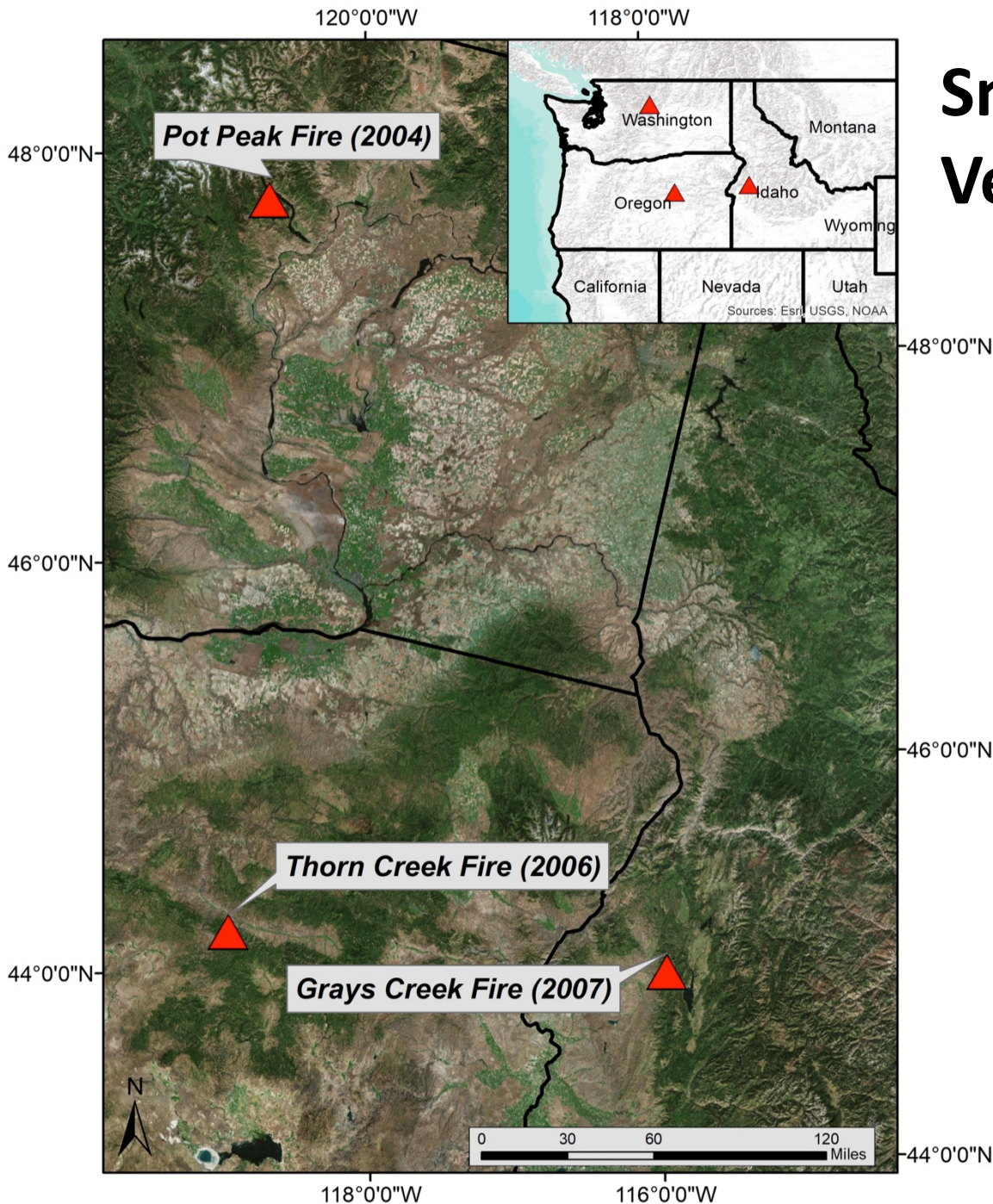
How do you predict streamflow in the absence of field data?

This challenge is particularly vexing in snowy mountain regions. Critical data are often absent and prediction efforts are hampered by lack of timely snow data. Predicting low flows is particularly important where water resources are limited.

Remote sensing, cloud computing, and interactive web-based mapping tools offer a new paradigm for delivering key streamflow data to water resource managers.

LATEST NEWS

- Nolin to join UNR as Director of Hydrologic Sciences program
- March 17-25 Nolin and Cosgrove complete snow surveys in Alaska
- Nolin hosts Mountain Sentinels workshop in Bend, Oregon
- Chris Cosgrove receives a



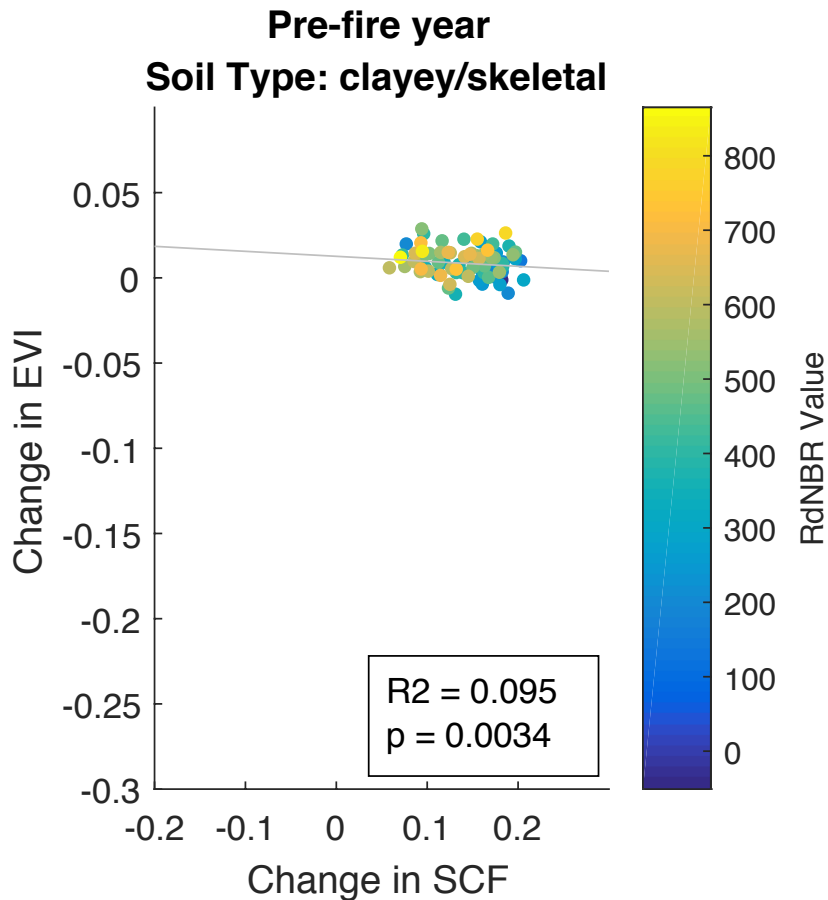
Snow and Post-fire Vegetation Recovery

Thorn Creek (2006)
Grays Creek (2007)

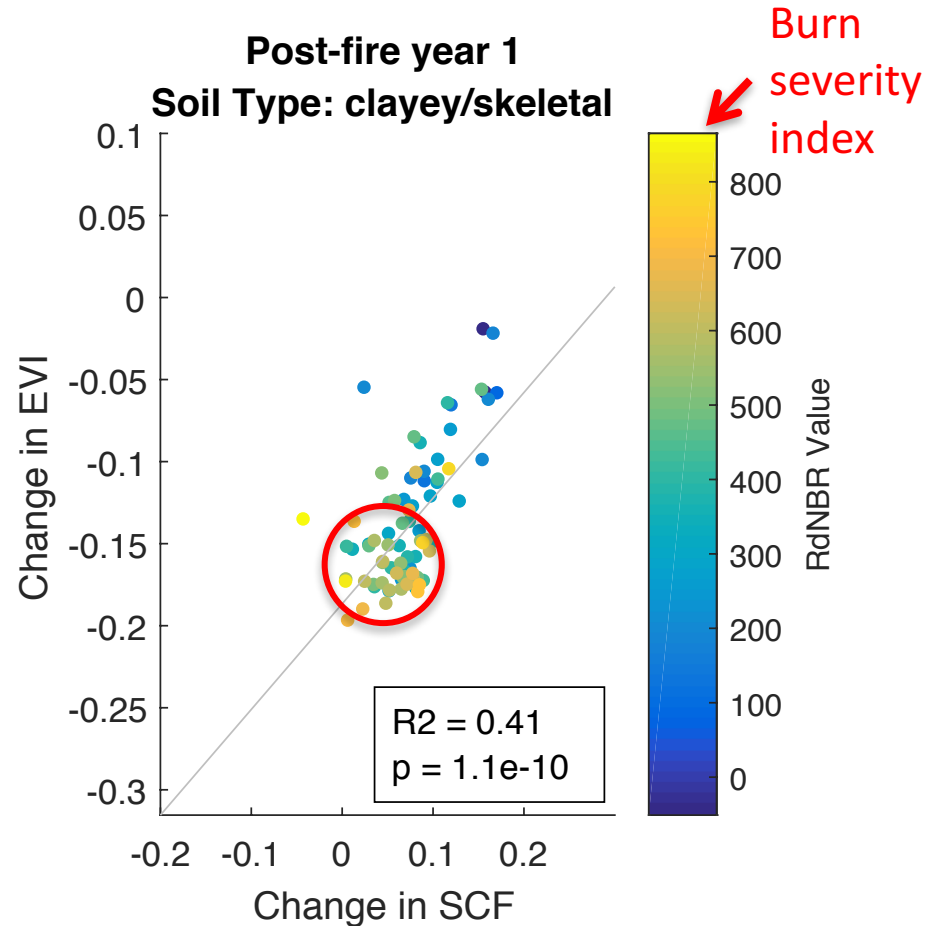
Variables:

- Max summer greenness (from MODIS EVI)
- Prior winter SCF
- Soil type
- Land cover type
- Burn severity

Grays Creek Fire: Δ Greenness vs. Δ SCF

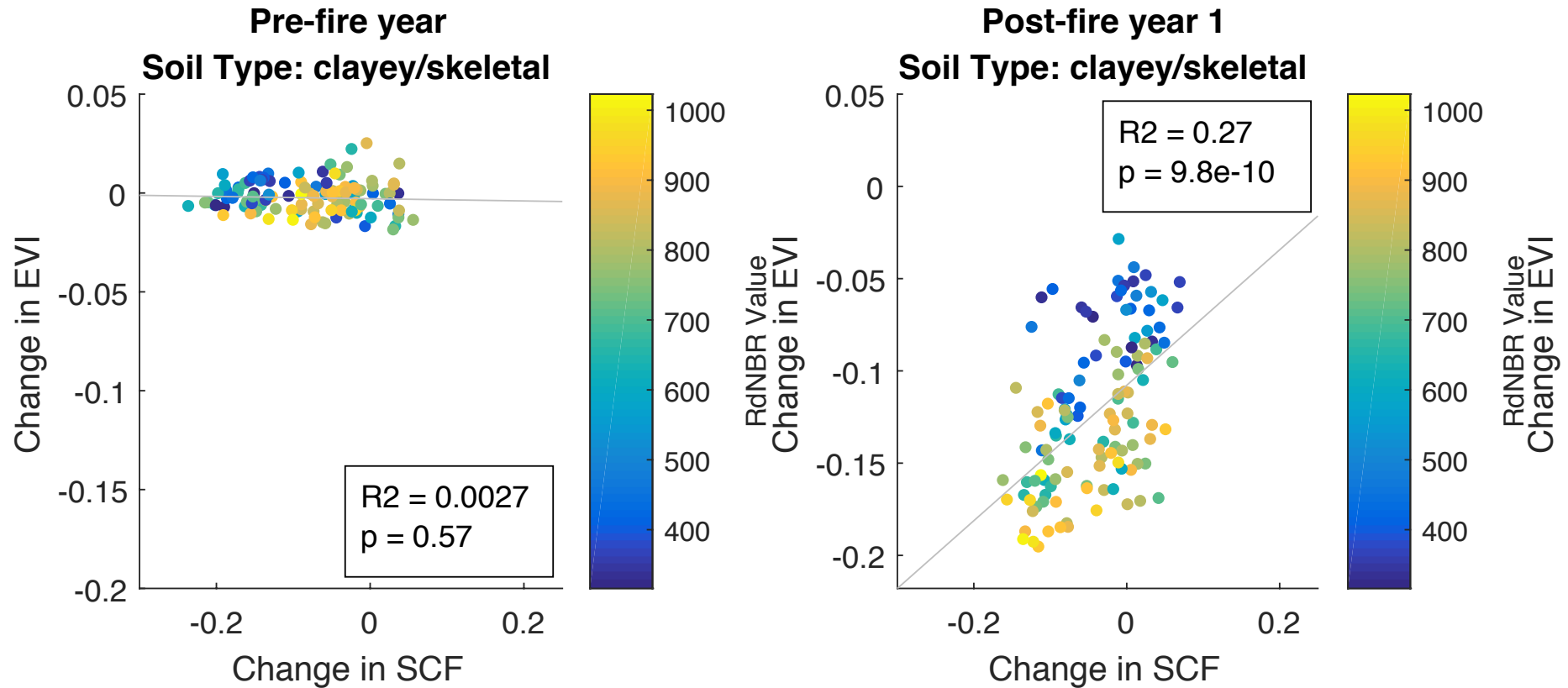


Pre-fire Δ EVI vs. Δ SCF; the relationship is not significant



Year 1 post-fire Δ SCF explains **41%** of variance in Δ EVI

Thorn Creek Fire: Δ Greenness vs. Δ SCF

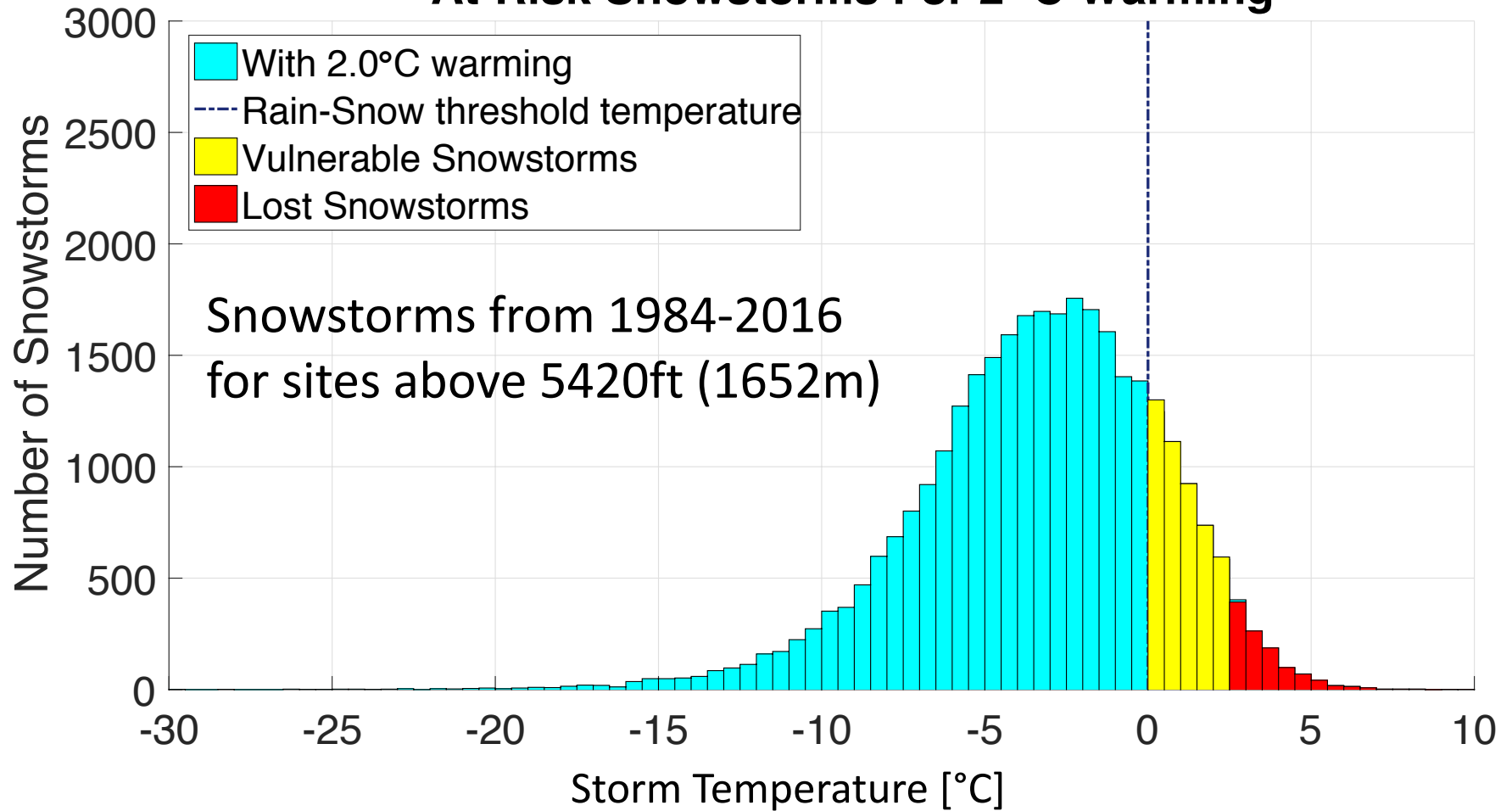


Pre-fire Δ EVI vs. Δ SCF; the relationship is not significant

Year 1 post-fire Δ SCF explains **27%** of variance in Δ EVI

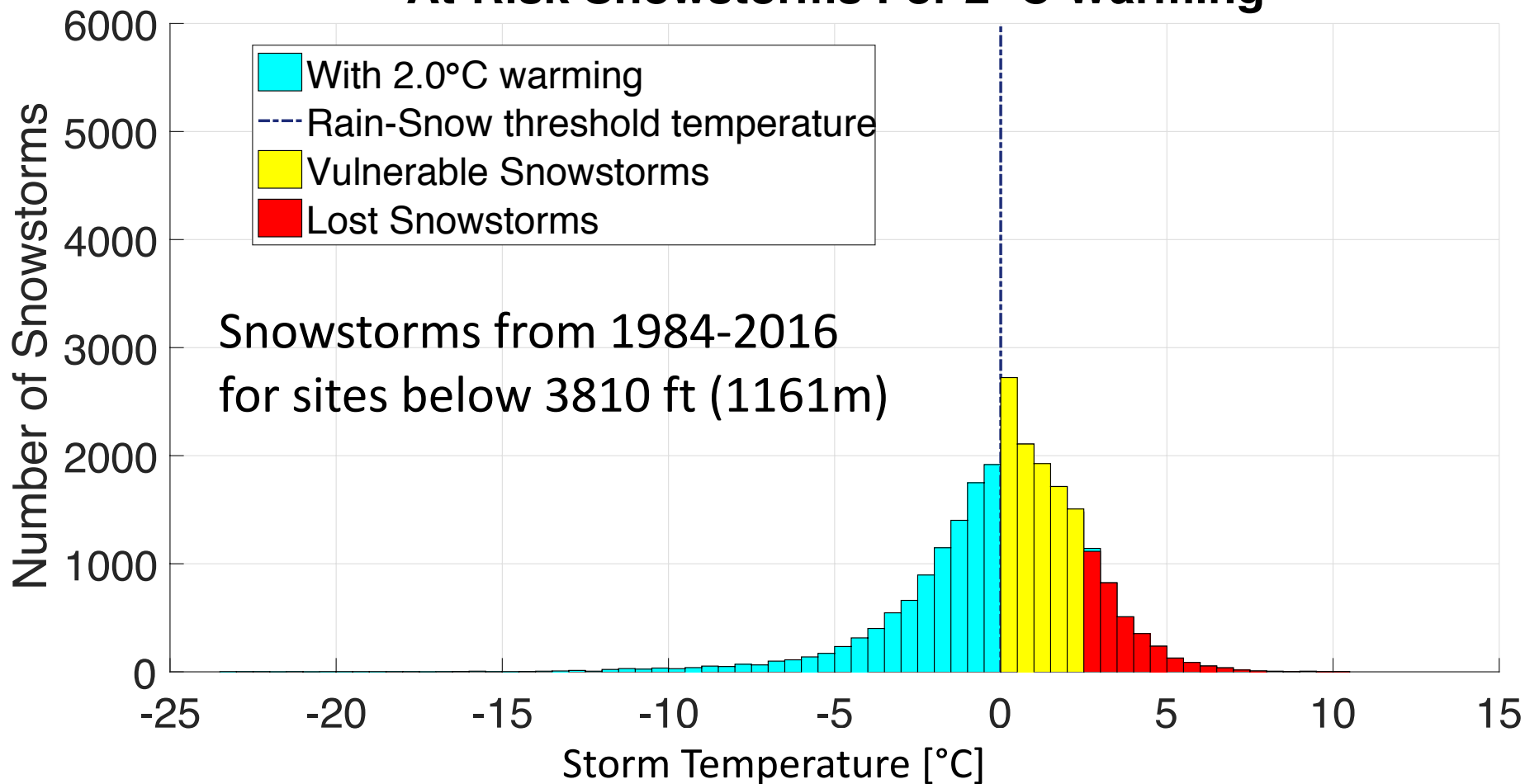
Snowstorm Temperatures: PNW High Elevation

At-Risk Snowstorms For 2° C Warming

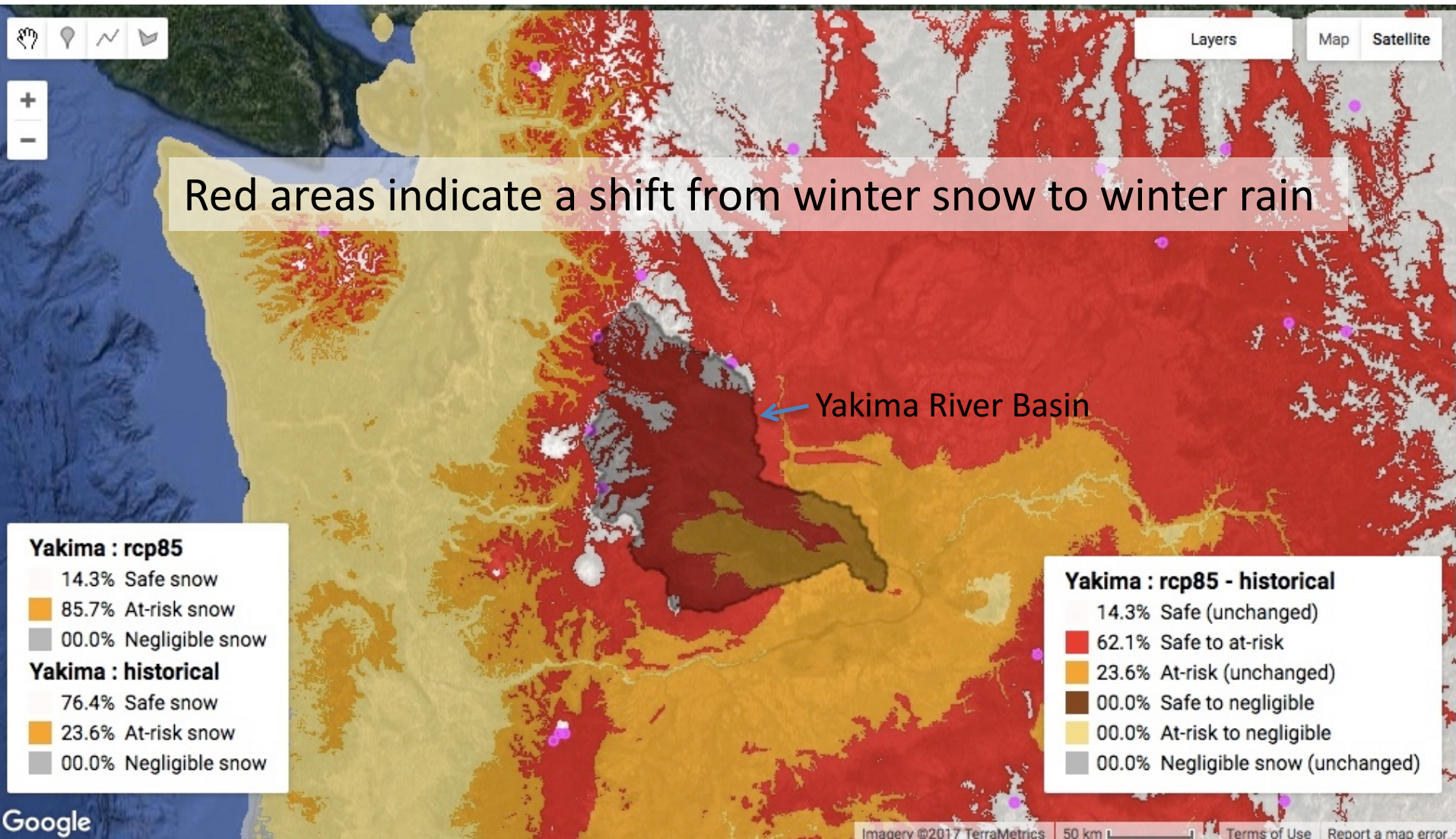


Snowstorm Temperatures: PNW Low Elevation

At-Risk Snowstorms For 2° C Warming



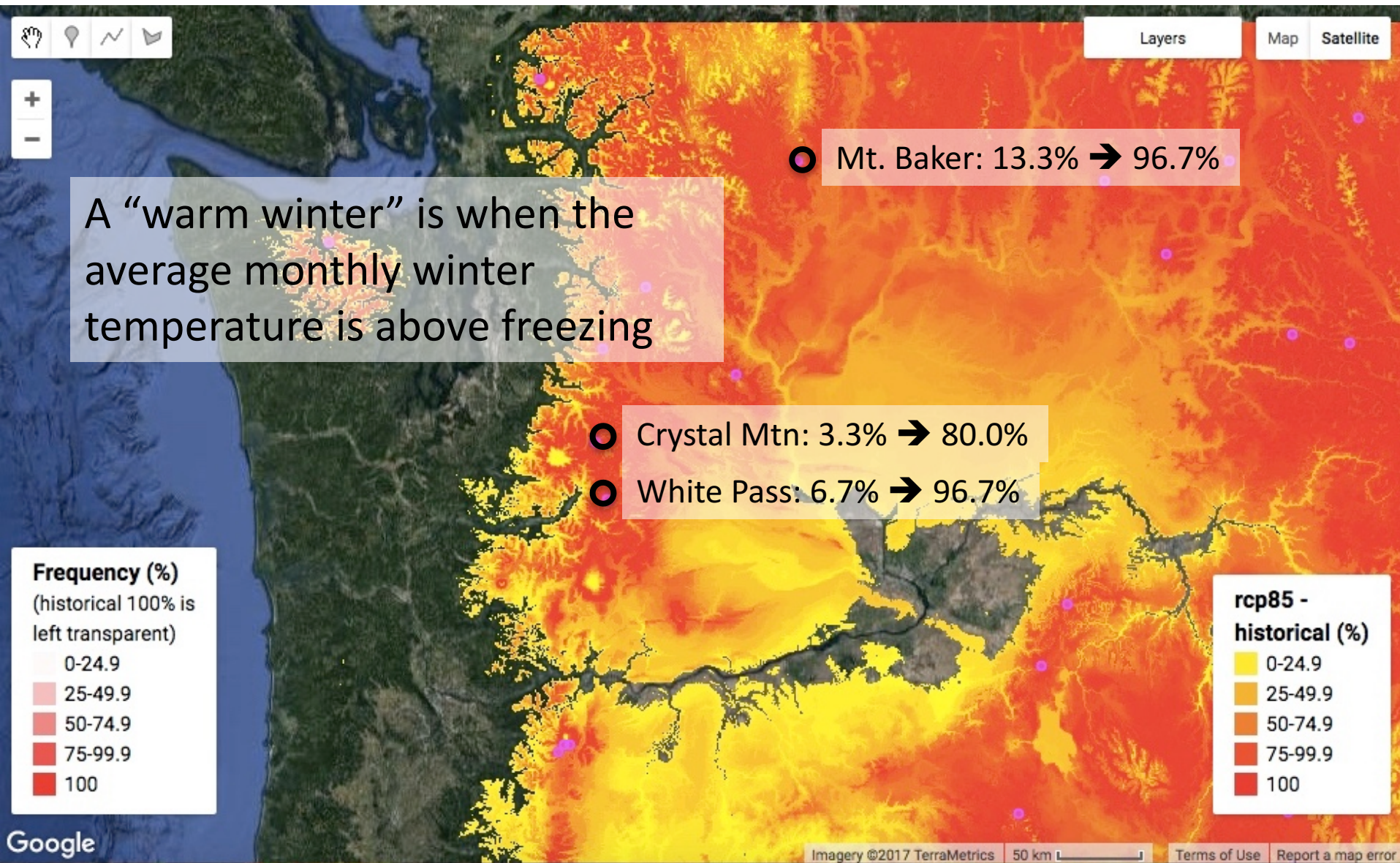
Historical and Future: Modeled “At-risk” Snow



RCP 8.5 represents the high end of climate change

Historical period is for 1970-1999; Future period is for 2079-2099

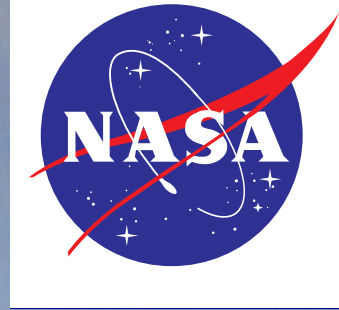
Historical and Future: Modeled Freq. Warm Winters



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Thank you.
Questions, please



College of Earth, Ocean, and Atmospheric Sciences