Now You See It Now You Don't: A Case Study of Ephemeral Snowpacks and Soil Moisture Response in the Great Basin U.S.A.

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1 Abstract

Ephemeral snowpacks, or those that persist for <60 continuous days, are challenging to observe and model because snow accumulation and ablation occur during the same season. This has left ephemeral snow understudied, despite its widespread extent. Using 328 site years from the Great Basin, we show that ephemeral snowmelt causes a 70 day earlier soil moisture response than seasonal snowmelt. In addition, deep soil moisture response was more variable in areas with seasonal snowmelt. To understand Great Basin snow distribution, we used MODIS and Snow Data Assimilation System (SNODAS) data to map snow extent. Estimates of maximum continuous snow cover duration from SNODAS consistently overestimated MODIS observations by >25 days in the lowest (<1500 m) and highest (>2500 m) elevations. During this time period snowpack was highly variable. The maximum seasonal snow cover over water years 2005-2014 was 64 % in 2010 and at a minimum of 24 % in 2014. We found that elevation had a strong control on snow ephemerality, and nearly all snowpacks over 2500 m were seasonal except those on south facing slopes. Additionally, we used SNODAS-derived estimates of solid and liquid precipitation, melt, sublimation, and blowing snow sublimation to define snow ephemerality mechanisms. In warm years, the Great Basin shifts to ephemerally dominant as the rain-snow transition increases in elevation. Given that snow ephemerality is expected to increase as a consequence of climate change, physics-based modeling is needed that can account for the complex energetics of shallow snowpacks in complex terrain. These modeling efforts will need to be supported by field observations of mass and energy and linked to finer remote sensing snow products in order to track ephemeral snow dynamics.

2 Introduction

Seasonal snowmelt supplies water to one-sixth of the world's population, which supports one-fourth of the global economy (Barnett et al., 2005; Sturm et al., 2017). Seasonal snowpack provides predictable melt timing and volumes in the spring, which influences streamflow timing, surface water and groundwater availability (Berghuijs et al., 2014; Jasechko et al., 2014; Stewart et al., 2005). Reliable spring snowmelt also provides a strong control on vegetation phenology and productivity in many ecosystems (Parida and Buermann, 2014; Trujillo et al., 2012). Despite the importance of seasonal snow to water supplies,

much of the world's snow is ephemeral (or intermittent), which means it melts and sublimates throughout the snow cover season instead of having one consistent period of snowmelt. Even small shifts from seasonal to ephemeral snowpacks due to regional warming could disrupt snowmelt rates and timing. A shift from seasonal to ephemeral snowpacks will also have negative implications for the winter tourism that requires continuous snow cover, as well as water management and hydropower

5   that relies on the predictability of snowmelt from mountain 'reservoirs' (Schmucki et al., 2017; Sturm et al., 2017). The hydrological impacts of ephemeral snowpacks have received little study.

Snowmelt influences a variety of terrestrial hydrological processes and states, particularly soil moisture dynamics in areas with low summer precipitation (Harpold and Molotch, 2015; Seyfried et al., 2009). Snowmelt-derived soil moisture is a primary control on streamflow generation and timing and ecosystem productivity in many semi-arid systems (Jefferson, 2011;

10   McNamara et al., 2005; Schwinning and Sala, 2004; Stielstra et al., 2015; Trujillo et al., 2012). Although few studies have isolated their hydrological importance, ephemeral snowpacks modify the intensity and duration of precipitation inputs to soil by storing and releasing water in a less predictable way than seasonal snow. For example, McNamara et al. (2005) described five predictable phases of soil moisture evolution in semi-arid watersheds with seasonally dominant snowmelt: (1) a summer dry period, (2) a transitional fall wetting period, (3) a winter wet, low-flux period, (4) a spring wet, high-flux period, and (5)

15   a transitional late-spring drying period. Soil moisture response to ephemeral snow melt is likely to sit between the predictable timing and rates of seasonal snow and the stochastic nature of rainfall, but few observations across this gradient exist. Despite the hydrological and ecological importance of ephemeral snow, we lack widely accepted methodologies to classify, map, and model snow ephemerality.

One commonly used snowpack classification system by Sturm et al. (1995) divides snowpack into six categories and defines

20   ephemeral snowpacks as those persisting for less than 60 consecutive days, are less than 50 cm depth, and have less than three different snow layers (Sturm et al., 1995). While it is arbitrary, using the 60-day threshold allows for comparisons between the extent of ephemeral snow to previous studies and among different areas. The Sturm et al. (1995) classification system is also incorporated into physical snowpack models, such as SnowModel (Liston and Elder, 2006), to separate seasonal and ephemeral snowpacks into different modeling domains. Models often make this separation because the energetics of ephemeral

25   snowpacks are much more sensitive to basal melt from ground heat flux. Additionally, cold content varies more rapidly through time in shallow ephemeral snowpacks. Most physics based models (e.g. Liston and Elder 2006), are optimized for seasonal snow, and produce less accurate results over ephemeral snow (Kelleners et al., 2010; Kormos et al., 2014).

Ground-based and remote sensing observations have their own strengths and weaknesses for observing ephemeral snowpacks and soil moisture response. Most ground-based snow measurement stations (e.g. the National Resource Conservation Snow

30   Telemetry, NRCS SNOTEL) in the Great Basin, and the Western United States, are built to observe seasonal snow (Fig. 1). This is because sites are typically placed in topographically sheltered forest gaps that retain snow longer than nearby terrain. This improves the skill of streamflow forecasting, the primary goal of the SNOTEL network, but means that most SNOTEL sites only have ephemeral snow cover in exceptionally dry or warm years (Serreze et al., 1999). Only two of the 131 SNOTEL stations in the Great Basin experience an ephemeral snow season on average (Fig. 1) each water year from 2005-2014. The scarcity

35   of ground-based ephemeral snow and soil moisture data has changed slightly in recent years with additional measurements

at NRCS Soil Climate Analysis Network (SCAN) (Fig. 1) and increased deployment in research watersheds (Anderton et al., 2002; Jost et al., 2007). On average, 26 out of 39 SCAN stations in the Great Basin experienced ephemeral snow cover each year (Fig. 1). However, the lack of field observations from ephemeral snowpacks with co-located soil moisture has limited previous investigations (e.g. Sturm et al. 2010).

5   Spectral remote sensing collects observations over all cloud-free areas, but has its own sets of advantages and challenges for observing ephemeral snow. One issue is that there are multiple methods to define the start and end of the observed snow covered period. Often, it is defined as the date of the first and last remotely sensed observations of snow cover (e.g Choi et al. 2010; Kimball et al. 2004; Nitta et al. 2014). Because this approach does not account for intermittent snow free periods, it tends to overestimate snow duration and miss important ephemeral dynamics (Thompson and Lees, 2014). Snow persistence

10   thresholds can be used to define snow ephemerality, but no standard persistence threshold exists (e.g. Gao et al. 2011; Karlsen et al. 2007). Given the intermittent nature of ephemeral snow, observations must be daily or finer to capture its dynamics (Wang et al., 2014). Consequently, products like Landsat that has a 16-day overpass and Sentinel that has 5-10 day overpass do poorly at estimating snow seasonality compared to products like the MODIS that have twice daily overpass, but they offer untapped potential for merged products with higher spatial and temporal resolution. Moreover, high cloud cover reduces observation

15   frequency, and limits the ability to observe ephemeral snow events. Like with ground-based snow research, some remote-sensing based studies exclude ephemeral events altogether (e.g. Sugg et al. 2014). Only a limited number of algorithms have been developed to handle ephemeral snow specifically. For example, the algorithm developed by Thompson and Lees (2014) uses daily MOD10A1 data and accounts for snow absences in the middle of the snow season, but their study was challenging to verify and applied only in a small area of Australia. Given the current lack of ground-based observations (Fig. 1), there is

20   great potential to use finer-scale satellite products and employ more refined methods targeted at areas with ephemeral snow.

There are a variety of underlying processes that cause ephemeral snowpack and challenge snow models. Based on previous classification systems, we define three mechanisms causing ephemeral snowpacks: 1) Rainfall limiting the accumulation of snowpack, 2) Snowpack ablation from melt or sublimation, and 3) Wind scour removing snowpacks. All of these mechanisms have a variety of underlying atmospheric and snowpack processes that challenge prediction with snow models. At rain-snow

25   transition elevations, even small temperature variations and other atmospheric variables can alter the mixture of rainfall and snowfall (Harpold et al., 2017b; Jefferson, 2011; Klos et al., 2014). Complete snow water equivalent (SWE) removal from melt or sublimation is also another common cause of snow ephemerality (Clow, 2010; Leathers et al., 2004; Mote et al., 2005; Sospedra-Alfonso and Merryfield, 2017). Typically, physics based models overestimate modeled SWE in ephemeral snowpack, due to neglect or underestimation of ground heat flux and the challenges of tracking cold content in shallow snowpacks (Cline,

30   1997; Hawkins and Ellis, 2007; Kelleners et al., 2010; Kormos et al., 2014; Tyler et al., 2008; Şensoy et al., 2006; Slater et al., 2017). Models parameterize energy fluxes differently, which can lead to differences in model estimates of sublimation and melt (Essery et al., 2009; Sospedra-Alfonso et al., 2016; Schmucki et al., 2014). Removal of snowpack from wind scour is an important control on snow accumulation in alpine regions, but is often neglected in models altogether (e.g. Mernild et al. 2017; Pomeroy 1991; Winstral et al. 2013). Widespread evidence exists that wind redistribution of snow can cause ephemeral

35   snowpacks that are consistent from year to year because of topography and dominant wind directions (Hood et al., 1999).

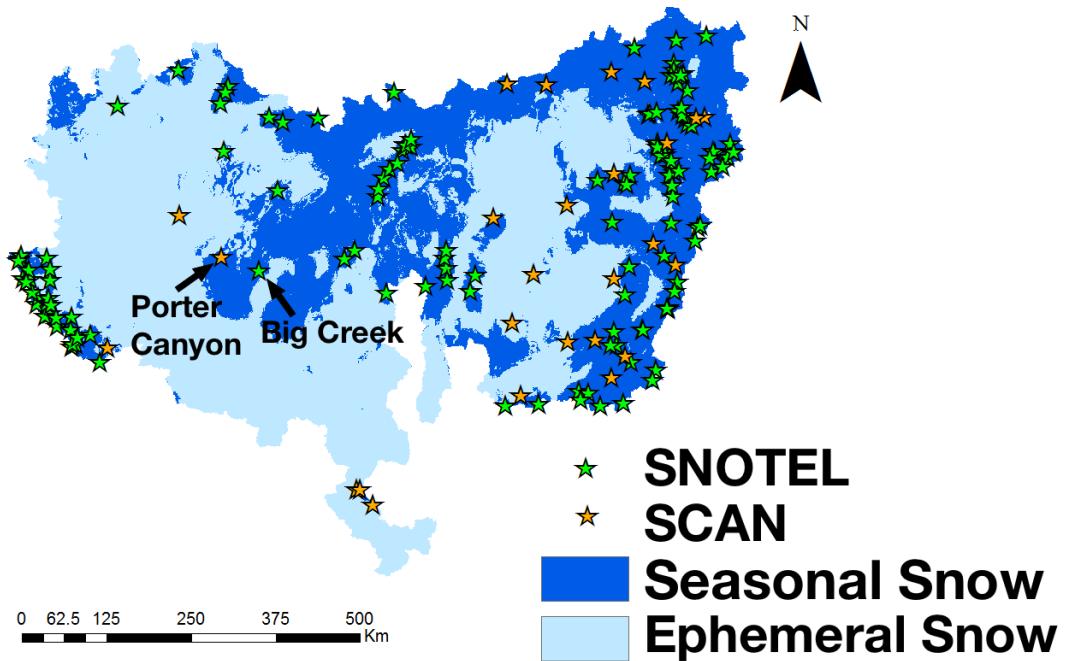



Figure 1. Locations of SNOTEL and SCAN stations in the Great Basin, USA in ephemeral and seasonal snow as defined by <60 or ≥60 days of maximum consecutive snow duration, respectively.

The three mechanisms causing ephemeral snow (i.e. rain-snow transition, ablation by sublimation and melt, and wind scour) have fundamentally different underlying causes, with variable and poorly quantified sensitivities to climate and land cover variability.

  The goal of this paper is to use the Great Basin as a case study to estimate the distribution and mechanisms causing ephemeral
5 snow to better constrain their impact on soil moisture and hydrological response. We adapt the classification from Sturm et al. (1995) to map snow across the Great Basin, compare remotely sensed and modeled estimates of ephemeral snow, and develop our own metrics to further classify snow seasonality. The Great Basin is ideal for this investigation because it spans dramatic gradients of elevation and hydroclimatology with large areas of both seasonal and ephemeral snow. This prototypical area depends disproportionately on mountain snowpack for water supplies, contains few ground-based observations, and there is
10 relatively little winter cloud cover to limit spectral remote sensing techniques. Three research questions guide our analyses of ephemeral snowpacks in the Great Basin: 1) What are the implications for soil moisture from seasonal to ephemeral snowmelt? 2) How does topography affect snow seasonality? and 3) What mechanisms cause ephemeral snowpacks and how does that vary with climate? We find that ephemeral snow originates from melt and shifts to lower elevation rain-snow transitions during warm winters, which leads to fundamentally different soil moisture response than from seasonal snowmelt.

3 Study Area

The Great Basin is the closed basin between the Wasatch and southern mountain ranges in Utah and the eastern slope of the Sierra Nevada mountain range in California. The region is known for having "internal drainage," which means that none of the waterways travel to the ocean (Svejcar, 2015). The climate is semi-arid and the ecosystem is shrub-dominated (Svejcar, 2015; West, 1983). We defined the Great Basin region based on the Hydrologic Unit Code (HUC) Region 16 adapted from Seaber et al. (1987) by the United States Geological Survey (USGS) (Fig. A1). Precipitation in the Great Basin varies widely between <10 cm in many of the lower elevations to >100 cm on many of the high elevation mountains (Fig. A2). Overall, the Great Basin has a mean winter (defined as Dec 1st to Apr 1st) precipitation of 12 cm and a mean winter temperature of 0.4 °C (Fig. A2; Abatzoglou 2012).

4 Methods

In order to compare the effect of snow ephemerality on soil moisture patterns, we first investigated snow and soil moisture response for SNOTEL and SCAN stations within the Great Basin. To evaluate how soil moisture varies based on snowpack parameters during a drought year (water year 2015) and a non-drought year (water year 2016), we chose two SNOTEL stations: Porter Canyon (ID: 2170, Elevation 2191 m) and Big Creek Summit (ID: 337, Elevation 2647 m) that differ in elevation but are in close proximity. We used average snow water equivalent (SWE) data from snow pillows to determine snow cover. We categorized each day as snow covered if continuous SWE was greater than 0.1 cm. We then designated site years as seasonal or ephemeral depending on if continuous snow cover was greater or less than 60 days respectively. For these stations, we compared percent soil moisture, at 5 and 50 cm soil depth along with snow depth, and SWE. We then also acquired soil moisture and SWE data at 5 and 50 cm for all the SNOTEL and SCAN stations in the Great Basin in water years 2014-2016 and categorized site years from those stations as ephemeral or seasonal. We discarded years and stations containing more than seven days of continuous missing data or soil moisture values that were 0 %. To compare the timing of snow and peak soil moisture, we then took the difference between the day of last snow and the day with peak median 10 day soil moisture for each year at each site. It should be noted that ablation on the snow pillow may be impacted by differences in ground heat flux and co-location issues with the soil moisture sensors. We also calculated the coefficient of variation (one standard deviation divided by the mean) of soil moisture for each year at each station.

We mapped ephemeral snow across the Great Basin using two methods: spectral remote sensing with MODIS data and modeled SNODAS data. We used Google Earth Engine to analyze the data, which is a cloud-based computing platform optimized for mapping large datasets (Gorelick et al., 2017). The MODIS dataset we used was the 2010 MODIS/Terra Snow Cover Daily L3 Global 500 m Grid (MOD10A) and we used the Normalized Difference Snow Index (NDSI) with parameters outlined in Hall et al. (2006) to find fractional snow covered data. The equation for calculating NDSI in MOD10 is:

$$NDSI = \frac{Band4 - Band6}{Band4 + Band6} \tag{1}$$

A pixel is then mapped as containing fractional snow using the NDSI value, as long as the reflectance in Band 2 is >10 % (Hall et al., 2001). We classified all pixels with a snow fraction of 30-100 as Snow, pixels with snow fractions between 0 and 30 as No Snow, and pixels that had all other designations as Other. We also used an algorithm derived from Thompson and Lees (2014) to minimize the impact of cloud cover in our MODIS data. The algorithm 'grows' the boundaries of all areas containing snow and reclassifies pixels that were classified as Other to Snow if the corresponding pixels in the previous image were classified as Snow. It also reclassifies pixels that were classified as Other to No Snow if the corresponding pixels in the previous image were No Snow.

To determine the number of ephemeral and seasonal snow events, we used a Google Earth Engine function to note the day of the water year when snow appeared (when a pixel went from classified as No Snow in the previous day to classified as Snow in the current day) and when snow disappeared (a pixel went from classified as Snow in the previous day to being classified as No Snow in the current day), and determined the length of snow cover by subtracting the day of snow appearance from the day of snow disappearance. If the length of snow cover was <60 days, then the snow event was classified as ephemeral. Otherwise, if the length of snow cover was ≥60 days, the snow event was categorized as seasonal. In addition to these metrics, we derived a snow seasonality metric (SSM) to quantify a MODIS pixel's tendency to have ephemeral or seasonal snow, rather than a binary metric like <60 days. The SSM is depicted in Eq. (2) and it works by classifying every day where there was seasonal snow present as 1 and every day where there was ephemeral snow present as -1, and then averaging all -1 and +1 values. This created a -1 to 1 scale, where -1 signifies that all the snow covered days in a given pixel within one water year were ephemeral and +1 signifies that they were all seasonal.

$$SSM = \frac{Days_{Seasonal} - Days_{Ephemeral}}{Days_{Total}} \tag{2}$$

Additionally, we discarded all instances where snow was absent for one day only from the overall record of snow disappearance and appearance because we found numerous artifacts from the MOD10A NDSI processing that lead to single day snow disappearance during long stretches of snow cover. One day snow events were also removed from the SNODAS algorithm to make both algorithms more consistent. For each water year from 2005 to 2014, we recorded the maximum total number of days where snow was present (to be referred to as the maximum snow duration).

To determine the relationship between elevation and snow seasonality, we took the average maximum snow duration across water years 2005-2014 and used elevation, and aspect as measured by a digital elevation model (DEM) obtained from the Shuttle Topography Mission resampled to the same resolution with bilinear sampling (Farr et al., 2007). To calculate northness, we used the equation:

$$Northness = cos(\frac{aspect * \pi}{180}) \tag{3}$$

We then categorized each MODIS pixel based on five 500 m elevation bins from a range of 1000 to >3000 m. Then, to remove bias based on the size of each bin, we used random sampling to make each bin contain the same number of points

as the least full bin (13548 points that were >3000 m). Then we combined each resampled bin into one dataset and created heatmaps to compare the elevation vs. the average maximum snow duration. We also use the same method to compare aspect to average maximum snow duration with aspect using eight 45 degree bins from a range of 0 to 360 degrees. We randomly sampled 195163 points from each bin (the size of the bin from 315 to 360 degrees). After resampling, we combined all the bins together and split them into three elevation categories: Low Elevation (Elevation <1500 m), Medium Elevation (1500 $\geq$ Elevation <2500), and High Elevation (Elevation $\geq$ 2500 m). Then, we resampled again to 82823 points per bin (the size of the High Elevation bin).

We used SNODAS data to differentiate the mechanisms that cause snow to become ephemeral. The four mechanisms were assigned if the net ablation (or rain) exceeded 50 % of the total winter precipitation (Fig. 2): 1) A mixture of rain and snow limiting snow accumulation (the rain-snow transition), 2) snowpack loss due to sublimation, 3) snowpack loss due to melt, and 4) snowpack loss due to wind scour. We determined the prevailing mechanism in each 1000 m SNODAS pixel in each year. We used Earth Engine to execute the modeled algorithm on each 1000 m SNODAS pixel in the Great Basin. We then chose six years (2009-2014) and created histograms of each mechanism by elevation for each year.

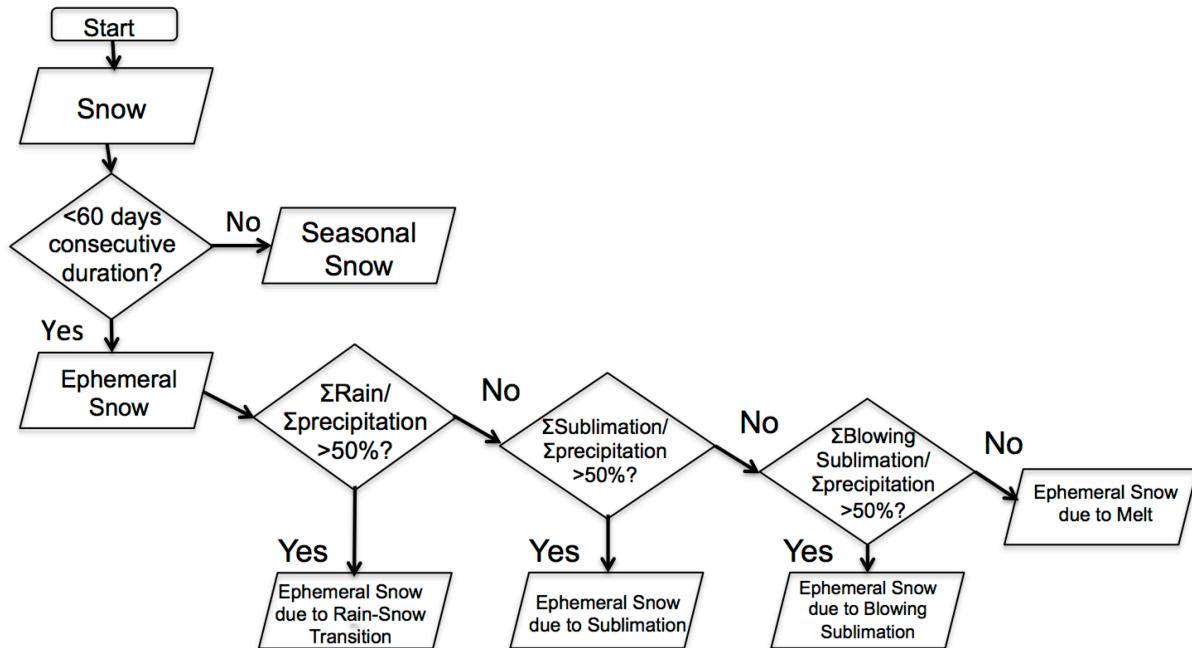



Figure 2. Diagram of the process for the ephemeral snow mechanism model. Seasonal snow outputs were rejected, all other outputs were categorized.

5 Results and Discussion

5.1 Ephemeral Snow and Soil Water Inputs

In order to quantify differing soil moisture responses between seasonal and ephemeral snowpacks that have important ecohydrological implications for the Great Basin, we use the five phases in the McNamara et al. (2005) framework for soil moisture response to seasonal snowmelt. First, we qualitatively compare two nearby sites with differing snow regimes. Second, we make quantitative analyses using all of the soil moisture records available in snow covered places of the Great Basin (Fig. 3).

We contrast soil moisture response at two adjacent SNOTEL stations that differ in elevation by >500 m (Fig. 1) to illustrate differences between ephemeral and seasonal snowmelt. Soil moisture at 5 and 50 cm depth was used to represent shallow and deep responses during a drought year (water year 2015) and a typical year (water year 2016). Porter Canyon had ephemeral snow (28 days maximum duration) in 2015 and seasonal snow (116 days) in 2016 (Fig. 3a). Big Creek had seasonal snowpack both years, although much shallower snowpack in 2015 (Fig. 3b). When seasonal snowpack is present at both sites in 2016, soil moisture follows the phases outlined by McNamara et al. (2005) for a semi-arid, snowmelt driven environment. Shallow and deep soil moisture was in a low-flux state during Dec-Feb (DJF) at Big Creek in 2016 (Fig. 3f). During Mar-May (MAM), soil moisture increased substantially and was in a high-flux state. Average shallow soil moisture in 2015 and 2016 was similar in the MAM period (24.4 % and 24.8 %, respectively) and DJF period (11.3 % and 19.8 %), suggesting that snow storage and melt negates differences in early season soil moisture between years with very different winter precipitation. Porter Canyon also showed a similar soil moisture increase in the MAM period after a stable low-flux pattern in the DJF period during water year 2016. Both sites also reach their near maximum annual soil moisture coincident with snow disappearance in 2016 (Harpold and Molotch, 2015), but Porter Canyon has snow disappearance in both years that preceded peak soil moisture by several months. The deeper 50 cm soil moisture had a smaller and shorter peak during 2015 at Porter Canyon as compared to 2016 and Big Creek response.

Using similar records to those illustrate at these two sites we use 328 site years (50 ephemeral and 278 seasonal site years) from all SNOTEL and SCAN sites in the Great Basin (Fig. 1) over water years 2014, 2015, and 2016 to illustrate the broader patterns of soil moisture response to ephemeral and seasonal snowmelt. We found that soil moisture following seasonal snowmelt reached a maximum 5 and 7 days prior to snow disappearance for shallow and deep soil moisture, respectively. This confirms previous findings that seasonal snowmelt drives coincident wetting and deeper water percolation (Harpold and Molotch, 2015; McNamara et al., 2005). In contrast, the median soil moisture peaked 79 and 48 days after of snow disappearance from ephemeral snowmelt for shallow and deep soil moisture, respectively (Fig. 4a). This is consistent with the peak shallow soil moisture occurring much earlier in the water year in shallow ephemeral snowmelt areas (Fig. 4b). The later deep soil moisture response in ephemeral areas reflects the lack of response, or low coefficient of variation (CV), as compared to seasonal snowmelt (Fig. 4c). The lower CV for deep ephemeral snowmelt (0.2) compared to deep seasonal snowmelt (0.4-0.5) is indicative of reduced deep percolation and less water becoming available to groundwater and streamflow.

The differences in soil moisture response between seasonal and ephemeral snowpacks across the Great Basin could have important consequences for vegetation phenology and runoff generation. For example, the timing of soil moisture is a strong

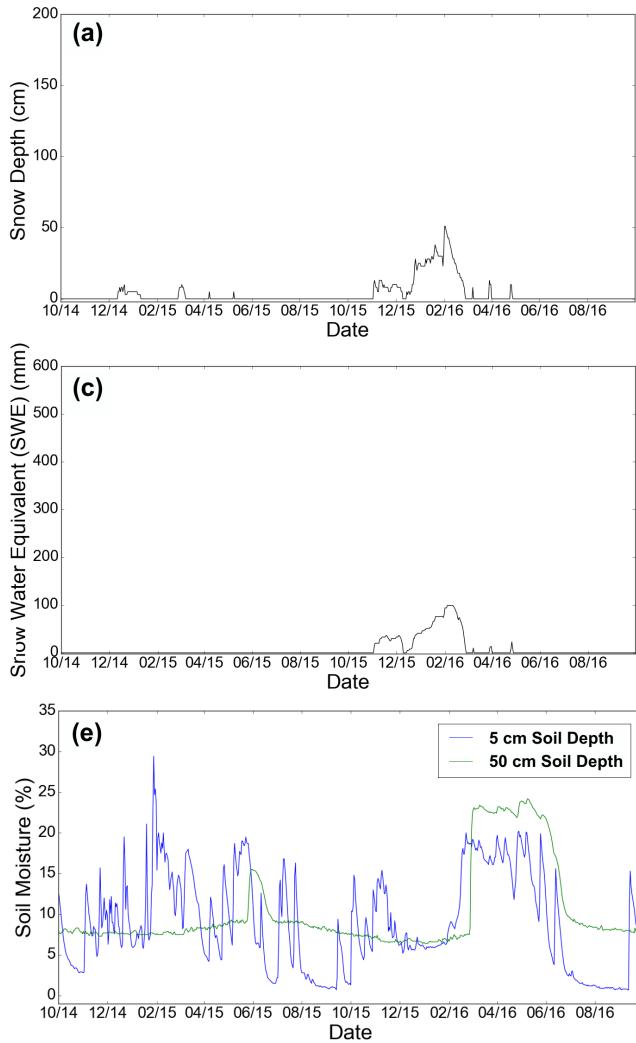
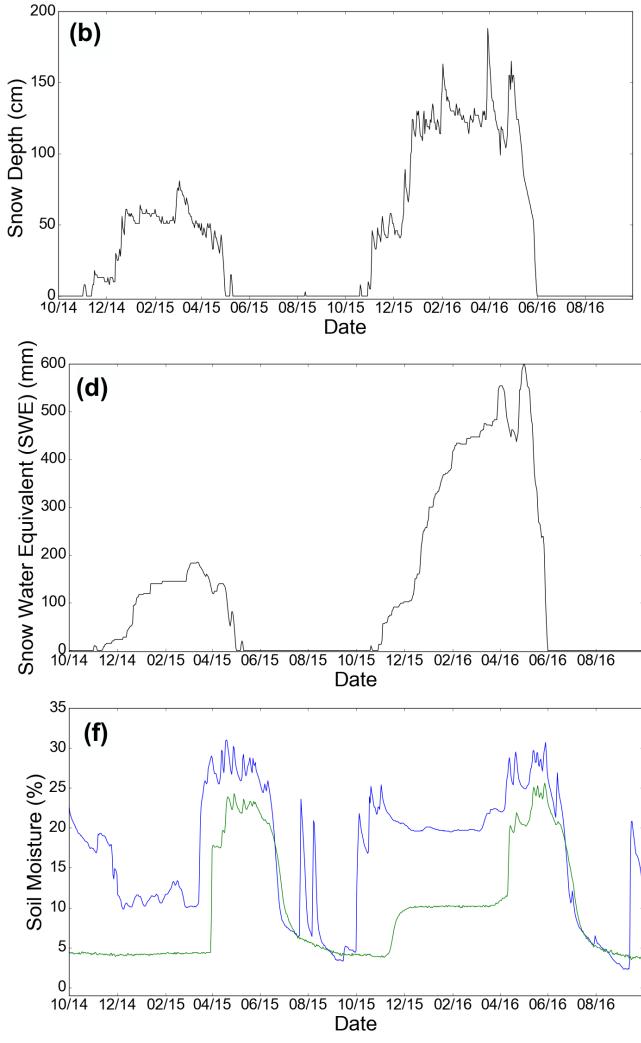



Figure 3. (a,b) Snow depth, (c,d) Snow Water Equivalent and (e,f) Soil Moisture measured at Porter Canyon and Big Creek Snow Telemetry (SNOTEL) stations for water years 2015-2016, which were a drought year and a typical year respectively.

control on the timing and amount of net ecosystem productivity (Inouye, 2008), with earlier snowmelt causing an earlier and longer growing season with reduced carbon uptake (Hu et al., 2010; Winchell et al., 2016). Harpold (2016) also showed that earlier snow disappearance generally led to more days of soil moisture below wilting point at SNOTEL sites. Our finding that soil moisture peaked earlier in ephemeral snowmelt than seasonal snowmelt is thus likely to be correlated with reduced vegetation productivity and increased late season water stress in many areas. In addition to stressing local vegetation, ephemeral snowmelt may reduce groundwater recharge and streamflow. For example, baseflow contributions to streamflow and overall

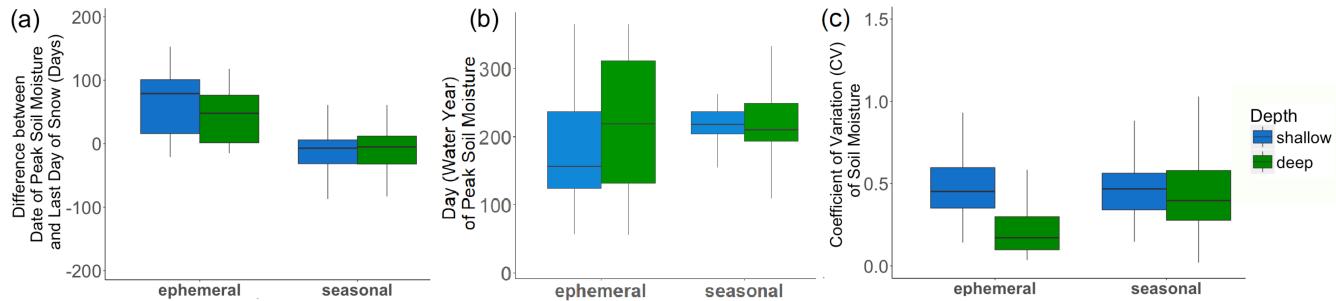



Figure 4. (a) The difference between date of peak soil moisture and last day of snow (Days) for shallow (5 cm) and deep (50 cm) soil moisture during water years 2014-2016 in Great Basin SNOTEL stations with ephemeral snow (50 site years) and seasonal snow (278 site years). (b) Day of peak soil moisture for SNOTEL and SCAN stations for shallow (5 cm) and deep (50 cm) soil moisture during water years 2014-2016 (c) The coefficient of variation (CV) for shallow (5 cm) and deep (50 cm) soil moisture during water years 2014-2016

water yield declined when snowmelt rates were smaller (Barnhart et al., 2016; Earman et al., 2006; Trujillo and Molotch, 2014), and overall water yields were lower in basins receiving more rain and less snow (Berghuijs et al., 2014). Changes in percolation patterns also affect the distribution of more shallow rooting plants versus deeper rooting plants that need long duration soil moisture pulses to grow and reproduce (Schwinning and Sala, 2004). These differences in how ephemeral versus seasonal snowmelt affects soil moisture provide a strong motivation to understand the distribution and causes of ephemeral snowpacks across the Great Basin.

5.2 Topographic Controls on Snow Seasonality

In a typical year, much the Great Basin experiences ephemeral snow (Fig. 5) that can only be comprehensively observed with remote sensing platforms because of the lack of standard ground stations (Fig. 1). Using MODIS imagery, we employ two new metrics to estimate snow ephemerality with daily snow cover products: 1) The maximum consecutive snow duration and 2) The snow seasonality metric (SSM). The SSM describes both the consecutive snow season length and shoulder-season ephemerality. A SSM value <1 means an area experiences at least one ephemeral snow event. The average SSM was -0.4 (Fig. 5), suggesting that on average the Great Basin was dominated by ephemeral snow extent. Maximum consecutive snow duration can be compared to the Sturm et al. (1995) 60-day threshold for ephemeral snow, as done in this case, but it is flexible enough to include a threshold of any day length. The average maximum consecutive snow duration in the Great Basin from MODIS data was 42.1 days (Fig. 5). We found higher estimates of the average maximum consecutive snow duration measured using SNODAS of 62.9 days, but a similar average SSM of -0.4 (Fig. 5). While the maps of the two products tend to produce similar results (Fig. 5), the SNODAS spatial patterns often miss finer scale topographic controls (e.g. Wasatch mountains in the far eastern Great Basin) and over estimates snow durations in the colder, lower elevations (e.g. basins below the Ruby Mountains in the central Great Basin). In general, SNODAS over estimates snow duration in areas with the longest and shortest snow durations, i.e. highest and lowest elevations (Fig. 6). In these critical water supply areas >2500 m, where snow would persist for >150 days according to MODIS, the SNODAS estimate were often biased by >50 days (Fig. 6). We explore the challenges of coarse, physically based models, such as SNODAS, later in this paper.

We investigate elevation and aspect as proxies for snowpack mass and energy dynamics in order to expand our understanding of snow ephemerality. Elevation is a primary control on near surface air temperature due to the adiabatic lapse rate (Bishop et al., 2011; Greuell and Smeets, 2001; Nolin and Daly, 2006). Prior research has found that there is a strong elevation dependence on snowmelt timing, runoff generation, snow water equivalent (SWE), and snow season length (Hunsaker et al., 2012; Jefferson, 2011; Jost et al., 2007; Molotch and Meromy, 2014). Elevation effects likely due to a variety of factors, including temperature controls on the rain-snow transition, longwave radiation in cloudy areas, and sensible heat flux. Aspect is often a secondary control on snow distributions because it influences incoming shortwave radiation (Jost et al., 2007; Pomeroy et al., 2003) and wind patterns (Knowles et al., 2015; Leathers et al., 2004; Winstral et al., 2013). Shortwave radiation is the primary driver of ablation via melt and sublimation (Cline, 1997; Marks and Dozier, 1992).

Dividing the Great Basin into low elevations (<1500 m), mid elevations (1500-2500 m), and high elevations (>2500 m) illustrated elevation's dominant role on snow cover duration (Fig. 7). Across the Great Basin, 96.2 % of low elevation area and 75.2 % of mid elevation area had a maximum consecutive snow duration of <60 days. Conversely, only 10.5 % of high elevations had a maximum consecutive snow duration of <60 days (Fig. 7). The results suggest that mid and low elevations of the Great Basin are more likely to be ephemerally dominant. The heat maps also illustrate that elevation alone is not a strong predictor of maximum consecutive snow cover days (Fig. 7). We use three smaller mountain ecoregions (Fig. A1) to illustrate variability in elevation effects (Fig. 8). There were similar average maximum snow duration values in the Ruby Mountains (Fig. 8a), eastern Sierra Nevada (Fig. 8b), and western Wasatch/Uinta ecoregion (Fig. 8c) (107, 100, and 95 days, respectively).

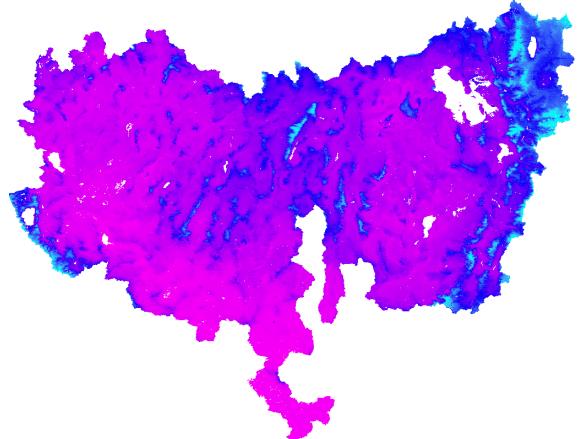
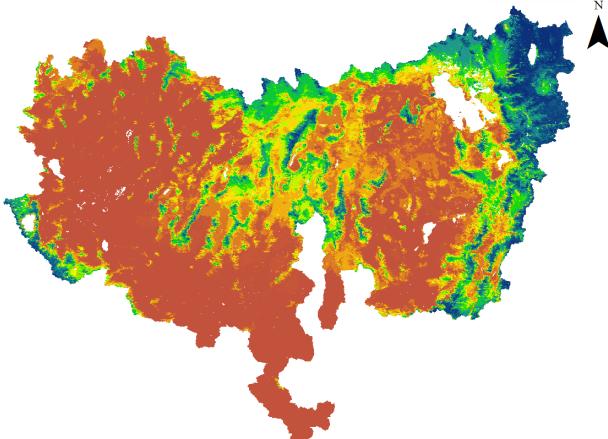
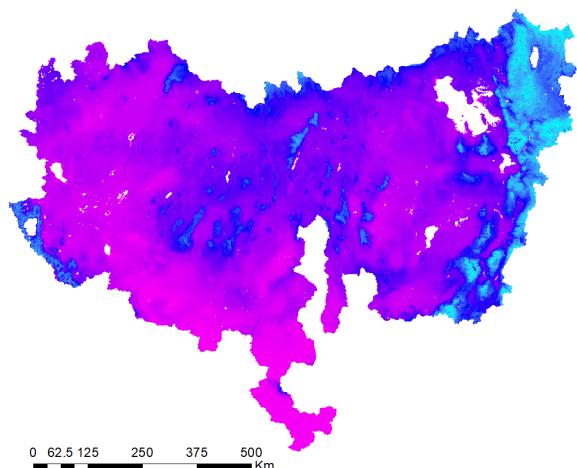
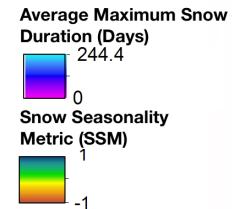
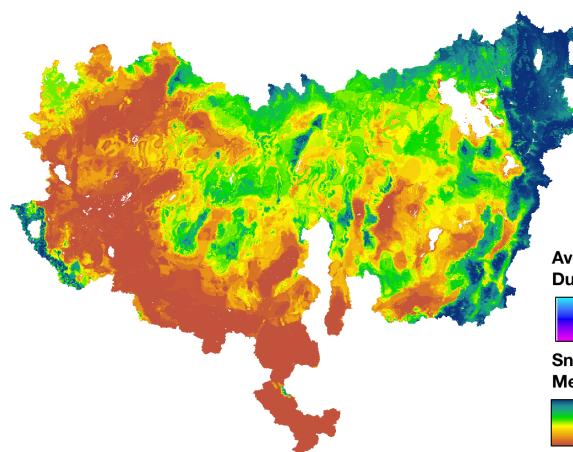



Figure 5. Average maximum consecutive snow duration (maximum snow duration) and snow seasonality metric (SSM) for the Great Basin measured using MODIS and snow data assimilation system (SNODAS) data in the Great Basin, USA for water years 2005-2014.

However, snow in the Ruby Mountains persisted longer than the Sierra Nevada and Wasatch/Uinta ecoregions. The Sierra Nevada ecoregion had a weaker relationship between snow persistence and elevation above 2500 m, while the Wasatch/Uinta ecoregion had a weaker relationship with elevation below 2500 m (Fig. 8). These differing relationships between maximum snow duration and elevation suggest other factors are affecting snow ephemerality.

5   Aspect is also an important control on snow seasonality in the Great Basin, but its importance is limited to mid and high elevations. We find that there are shorter maximum snow durations in south-facing aspects at elevations >1500 m (Fig. 9). At low elevations, the difference in average maximum snow duration between north and south facing slopes was 0.4 days, while for mid and high elevations, it was 2 and 5 days, respectively (Fig. 9). This is consistent with aspect strongly controlling

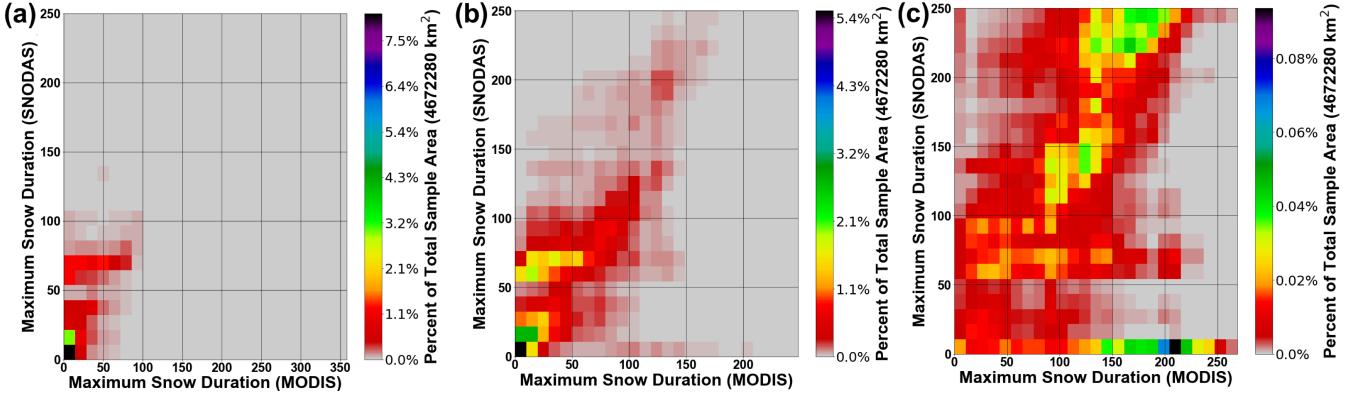



Figure 6. Maximum consecutive snow duration (maximum snow duration) measured using MODIS and snow data assimilation system (SNODAS) data and snow seasonality metric (SSM) for the Great Basin measured using MODIS.

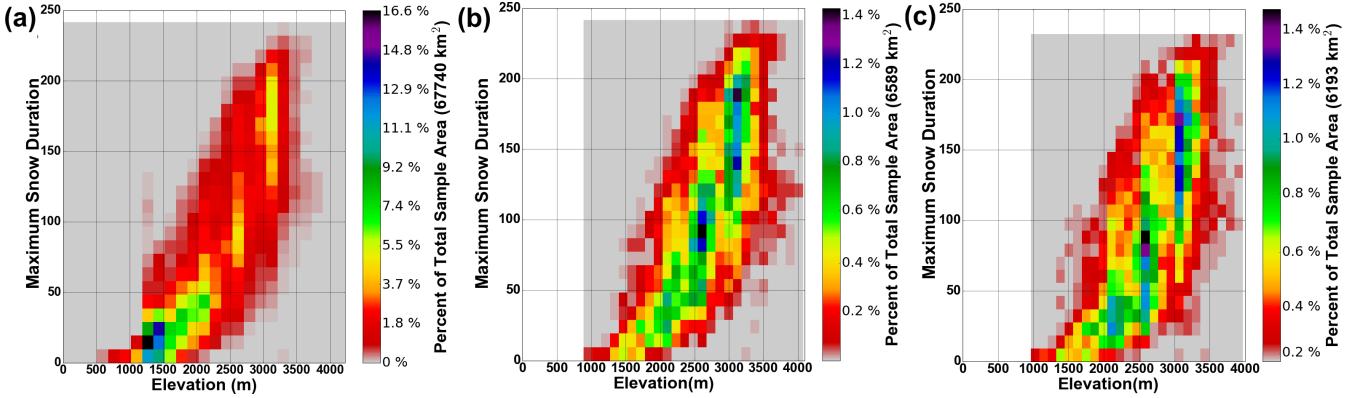



Figure 7. Heatmaps of the relationship between elevation and average maximum consecutive snow duration from MODIS at (a) all slopes,(b) north-facing slopes only, and (c) south facing slopes only in the Great Basin, USA. North facing was defined as Northness >0.25 and south facing was defined as Northness<-0.25. Color bar scale is different in panel (a) reflecting the much larger area at low elevation.

solar radiation, which is the main energy input to the snowpack. This suggests that deeper, high elevation snowpacks ablate in response to greater solar radiation and corresponding warmer temperature on south facing hillslopes (Hinckley et al., 2014; Kormos et al., 2014). In contrast, lower elevation areas appear to have maximum snow duration caused by factors other than aspect. This is consistent with the outsized importance of other energy fluxes and factors, like ground heat flux and rain-snow transition elevation, that are not captured by aspect and elevation (Fig. 7, 8 and 9).

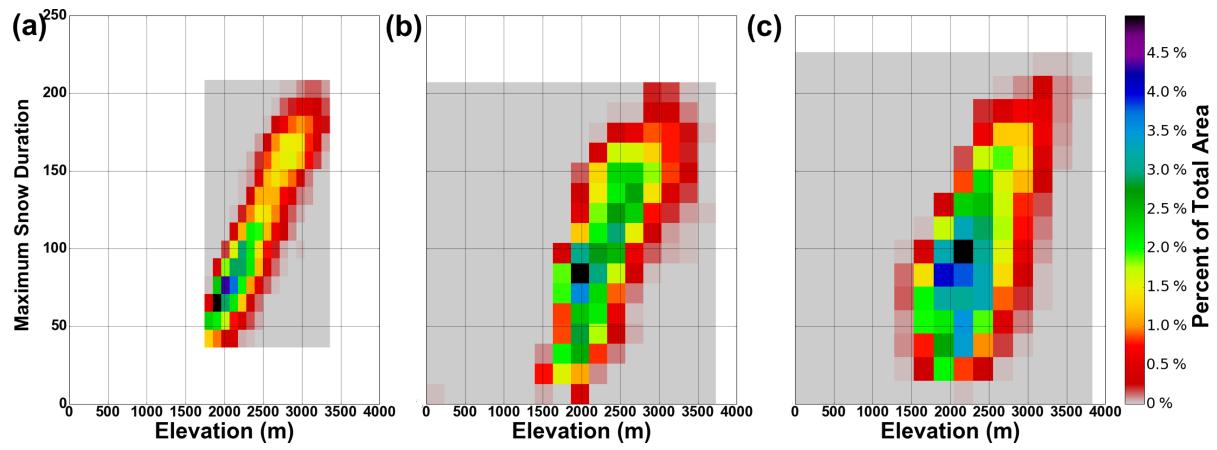



Figure 8. Heat maps showing the relationship between elevation and average maximum snow duration for three seasonally-dominant ecoregions in the Great Basin: (a) The Ruby mountains, (b) the Sierra Nevada mountains, and (c) the Wasatch/Uinta mountains.

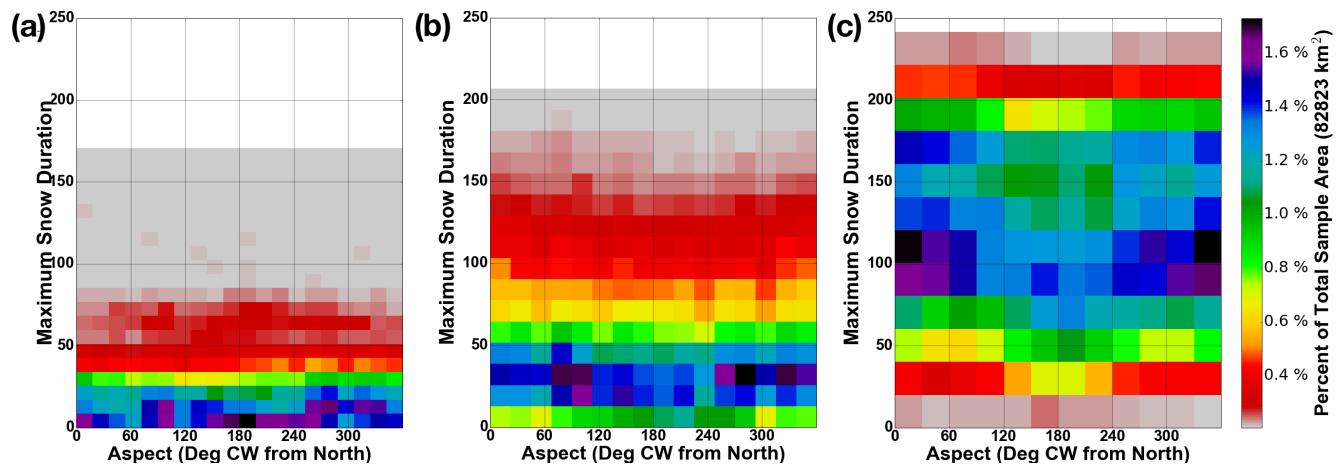



Figure 9. Heatmaps of the relationship between aspect and average maximum consecutive snow duration at (a) low elevations (0-1500 m), (b) medium elevations (1500-2500 m) and (c) high elevations (2500 m+).

5.3 Proximate Mechanisms Controlling Snow Ephemerality

We propose a three-mechanism classification scheme to help frame our understanding of snow ephemerality: 1) rain-snow transitions limit snow accumulation, 2) snowpack ablation from melt and sublimation, and 3) wind scour or redistribution. Probably the most explored and observed mechanism is the potential for rising rain-snow transition elevations to limit snow accumulation and duration (Bales et al., 2006; Klos et al., 2014; Knowles and Cayan, 2004; Mote, 2006). Reduction in snow duration can also be caused by the melt of snowpack (Mote, 2006) and losses from sublimation (Harpold et al., 2012; Hood et al., 1999); however, much less is known about the role and distribution of these processes outside of the seasonal snowpack zone. Finally, wind scour can reduce snowpacks by redistributing it to other areas or by increasing blowing wind sublimation (Knowles et al., 2015; Leathers et al., 2004).

We chose six years to evaluate the dominant mechanisms causing snowpack ephemerality using a new classification system (Fig. 2) based on SNODAS data that compared favorably to estimates from MODIS (Fig. 5 and 6). In that six year period, the year with the lowest average winter (Dec 1st to Apr 1st) temperature using GRIDMET estimates was 2013 at -0.9° C while the year with the highest average winter temperature was 2014 at 1.0° C (Abatzoglou 2012 Table 1). In water year 2013 and water year 2010, the two coldest years, seasonal snowpacks were dominant in most of the Great Basin and Western United States (Fig. 10-11). In the coldest years of 2010 and 2013, the rain-snow transition and melt caused ephemerality to shift lower in elevation (Fig. 11). In the warmest year of 2014, seasonal snowpack was lowest at lower elevations throughout the Western US mountain ranges (Fig. 10), including the Great Basin where the increase in ephemeral snowpacks at higher elevations was due primarily to a rain-snow mechanism (Fig. 10 and 11). Melt caused snow ephemerality also increased in the warm 2014, but ephemeral snow remained sparse above 2500 m in all years. Overall, our findings are consistent with the importance of variability in rain-snow transition elevations limiting snow accumulation and duration (Bales et al., 2006; Klos et al., 2014; Knowles and Cayan, 2004; Mote, 2006). Sublimation was only present as a limiting mechanism in 2010 and only for a small area (Fig. 10). Blowing snow sublimation was not the dominant cause of snow ephemerality in the Great Basin for any year, but its known that SNODAS struggles to represent wind redistribution of snow (Clow et al., 2012; Hedrick et al., 2015). Our approach to classify proximate causes of snow ephemerality has some limitations. Namely, it assigns only a single mechanism to each grid cell when there could be multiple mechanisms. Moreover, the method cannot consider changes in the mechanisms with time (e.g. melt tends to occur more in spring) because we applied annualized estimates of snow cover duration and concerns about the fidelity of the SNODAS model at short time scales.

The mechanisms causing snow ephemerality that can be inferred from the SNODAS model have important implications for water availability in the Great Basin, but we lack confidence in the model fidelity in these shallow snowpacks given their differences with the MODIS observations (Fig. 6). These limitations are present in all snowpack energy models because the models were developed for deeper snowpacks where terms like ground heat flux and albedo depth relationships can be ignored or are insensitive (Cline, 1997; Harstveit, 1984; Liang et al., 1994; Tyler et al., 2008; Slater et al., 2017). In shallow snowpacks, these terms are more critical (Hawkins and Ellis, 2007; Şensoy et al., 2006; Slater et al., 2017; Tyler et al., 2008), and the lack of SWE means the internal energy state of the snowpack (i.e. cold content) is more easily varied by short term climate forcing

Table 1. Average winter (Dec 1st-Apr 1st) temperature (°C) and average elevation (m) for both dominant mechanisms of snow ephemerality and seasonal snow from 2009-2014 in the Great Basin.

Water Year	Average Winter Temp (deg C)	Mean Elev for Rain Snow Transition (m)	Mean Elev for Melt (m)	Mean Elev for Seasonal Snow (m)
2009	0.1	1806.3	1750.8	1728.4
2010	-0.6	1811.3	1747.1	1761.3
2011	-0.2	1803.7	1765.6	1699.6
2012	0.4	1803.7	1745.2	1709.8
2013	-0.9	1815.6	1709.8	1754.1
2014	1.0	1789.9	1748.9	1731.5

(e.g. warm, sunny days) (Liston, 1995). Ephemeral snowpacks also exist at lower elevations with warmer soils and increased ground heat flux (Slater et al., 2017; Tyler et al., 2008). Uncertainty in the rain-snow transition principally arises from predicting climate forcing and in particular temperature and humidity in places like the Great Basin (Harpold et al., 2017a). However, the underlying phase prediction method and related model decisions and climate forcing data can also be important for the quality

5   of precipitation phase prediction (Harpold et al., 2017b). Further complicating rain-snow transition mechanisms is storage or drainage of liquid water on existing snowpacks (Lundquist et al., 2008; Marks et al., 2001). Although SNODAS assimilates MODIS imagery into the model, it does not appear to capture the finer elevation patterns we found using the MOD10A product (Fig. 5 and 6), and in particular, seemed to overestimate consecutive days of snow cover. Part of the challenges at higher elevations is modeling blowing snow patterns over 1-km grid cells, which gives consistent lower accuracy of SNODAS above

10 tree line and in more windy areas (Clow et al., 2012; Hedrick et al., 2015). The Great Basin shows tremendous variability in snow ephemerality caused by interactions of topography, elevation, and prevailing wind (Fig. 10-11) and thus, represents an area where improvements in the physically-based modeling will be critical to predicting snow water resources under a variable and changing climate.

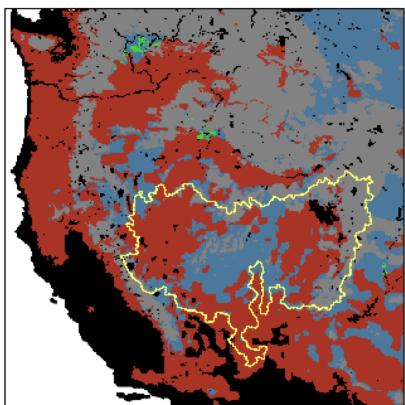
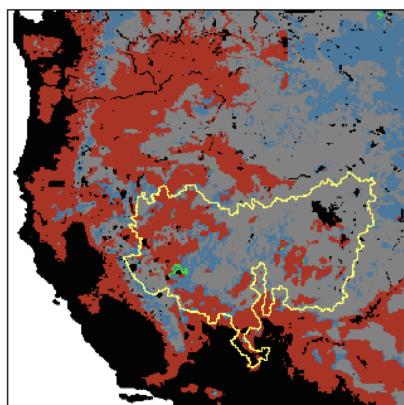
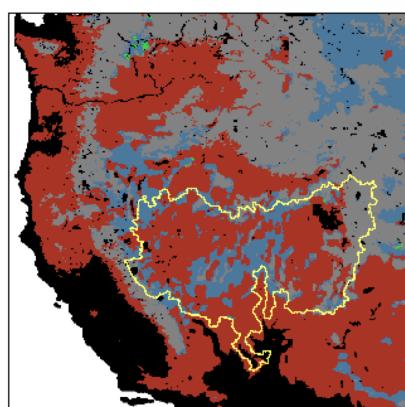
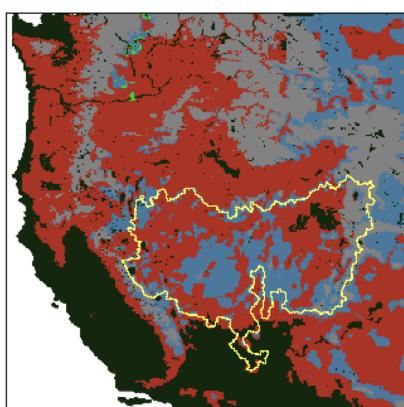
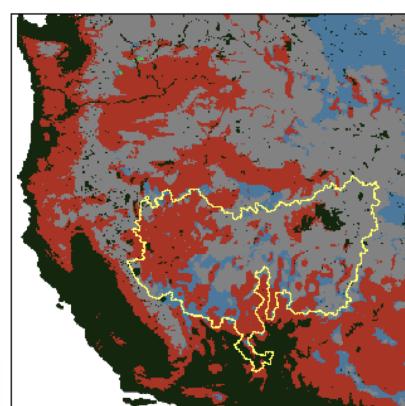
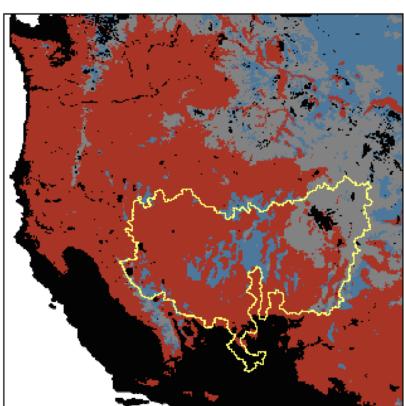



Figure 10. Dominant mechanisms for snow ephemerality from water years 2009-2014 in the Western United States. Areas with seasonal snow (grey), no snow (black), and water bodies (black) are also depicted. The Great Basin region is outlined in yellow.

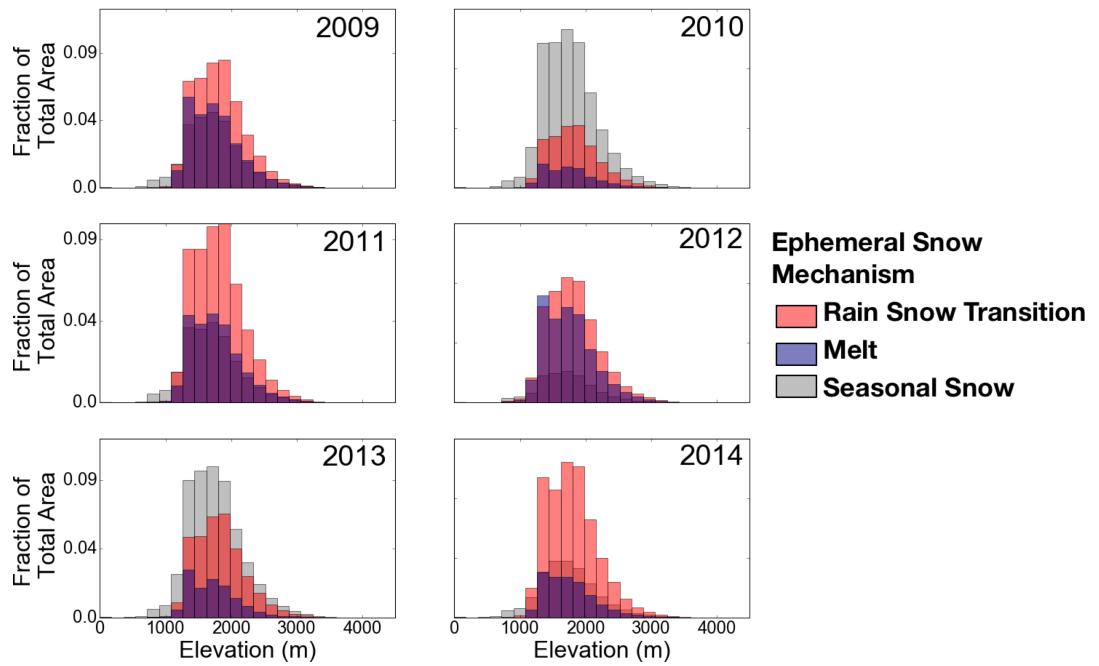



Figure 11. Histograms of the relationship between elevation and the dominant mechanisms for snow ephemerality in the Great Basin from water years 2009-2014.

6 Conclusions

Mapping, measuring, and modeling ephemeral snow is challenging with current techniques, but it is vital for understanding future water resources and vegetation water use. Ephemeral snowpacks do not have distinct accumulation and ablation periods, which means the timing of soil moisture input varies and is more challenging to predict than seasonal snowmelt (e.g. McNamara et al. 2005). Consequently, as snowpacks shift from seasonal to ephemeral, there are potential ecohydrological consequences such as changes to vegetation response, vegetation distribution, lateral water flow, and solute transport. Our work shows that while topography and climate variability have strong controls on the distribution of ephemeral snowpacks (Fig. 7 and 10), those factors will not be sufficient for predicting snow ephemerality under varying climate. Instead, we will need physics-based models capable of capturing the three broad mechanisms identified by this study: 1) rain-snow transitions limit snow accumulation, 2) snowpack ablation from melt and sublimation, and 3) wind scour and redistribution. These classifications could help better identify local and regional sensitivity to increased snow ephemerality (Fig. 10 and 11). This work has also highlighted major weaknesses in the observational infrastructure, data analysis, and modeling techniques needed to support the growing importance of ephemeral snowpacks in the Great Basin. In light of these diverse needs, we conclude with a short summary of recommendations meant to guide future research directions:

• Improving and standardizing snow ephemerality metrics: Our research suggests there is a snow duration threshold where snowpack and soil moisture patterns begin to resemble seasonal instead of ephemeral snowmelt, and perhaps a second threshold when they begin to resemble rain (Fig. 3). Yet evidence that this threshold is near the 60 days used in the Sturm et al. (1995), or consistent across space, is lacking. Instead of using this arbitrary 60 day threshold, we recommend that future research use the snow properties and soil moisture response of ephemeral snowpacks combined with a sensitivity analysis to create a snow duration threshold capable of differentiating seasonal and ephemeral soil moisture response (e.g. McNamara et al. 2005).

• Increasing snow and soil moisture observations in ephemeral areas: In the Great Basin, only two snow telemetry (SNOTEL) stations and 26 soil climate analysis network (SCAN) stations observe ephemeral snowpacks (Fig. 1). The lack of observations makes it more difficult to leverage develop relationship betweeen snow melt soil moisture. To help develop better criteria for categorizing snowpack as ephemeral, we need more snow and soil moisture observations in ephemeral areas. Also, we show that observing both shallow and deep soil moisture can add significant hydrological inferences. We can then also use these observations to verify results derived from remote sensing and physically-based models.

• Improved remote sensing algorithms: There is currently no consistent standard for defining the length of snow covered period. It is still common for papers to define the length of a snow covered period by the first and last days of snow cover. This approach does not account for short-term snow disappearance between those days. Approaches that report the total number of snow covered days miss information contained during show snow-free periods. Additionally, there is no consistent algorithm for accounting for cloud cover and that may make these types of methods infeasible for some regions. More widespread use of the object-oriented techniques, like the one used in this study, is needed to evaluate their efficacy and accuracy across differing regions.

• Improved spatial resolution and fidelity of snow and climate data: The MOD10A data product has a spatial resolution of 500 m. The coarse resolution made it difficult to verify our ephemeral snow results with SNOTEL observations that use 3 m wide snow pillows. Topographic complexity leads to variations in climate on much finer resolutions than the 4000 m gridded meteorology data used for this analysis. Gridded snow and climate data should have a spatial resolution more consistent with the variability in snowpacks on the order of 10-100 meters. While very fine resolution climate datasets are beginning to be produced, there is a large need to merge existing remote sensing snow observations into a data product that maximizes the current space and time resolutions across different platforms (e.g. spatial resolution of Sentinel 2 but the temporal resolution of MODIS).

• Improved physics-based modeling: Identifying weaknesses in physically-based models was not the objective of this study; however, it is clear this is a need for better prediction of snow ephemerality. Improving model parameterization of ground heat flux and ensuring the temporal model resolution is sufficient to capture rapid changes in cold content are two ways to improve these models. These improvements are contingent on new and better observations of mass and energy fluxes to support greater model fidelity in ephemeral snow.

Data availability. Petersky, R., and Harpold, A.: Now You See It Now You Don't: A Case Study of Ephemeral Snowpacks in the Great Basin U.S.A. ScholarWorks, https://scholarworks.unr.edu/handle/11714/2952 2018.

Appendix A

Contents of this file: Figures A1-A4

5   **Introduction:**

The following figures provide additional information about the ephemeral snow algorithm. Figure A1 is an elevation map of the Great Basin, USA showing key ecoregions and major cities. Figure A2 is a map of average winter (Dec 1-Apr 1) temperature, precipitation, and radiation across water years 2001-2015. Figure A3 shows how the measured number of ephemeral and seasonal snow events at SNOTEL sites corresponded to the number derived from the ephemeral snow algorithm. Figure

10   A4 shows how the 30 % snow fraction was chosen using a sensitivity analysis.

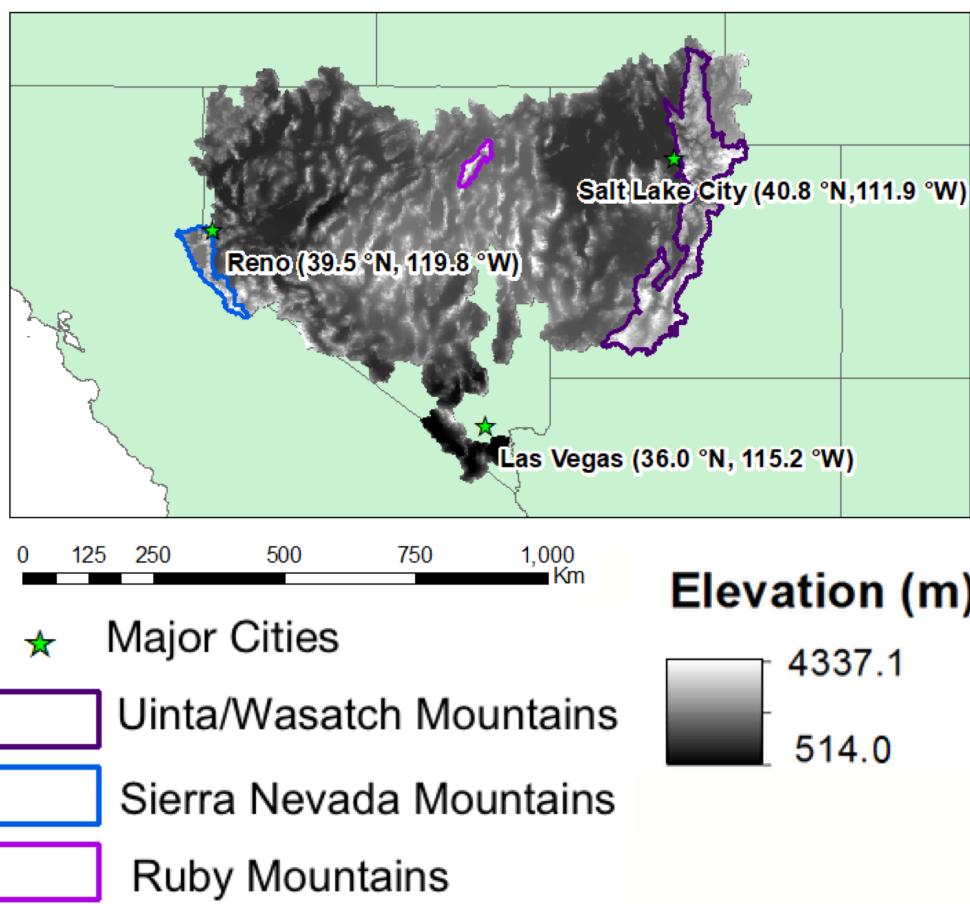



Figure A1. Map of the Great Basin region, USA as defined by the USGS HUC Region 16 along with major cities and mountain ranges. The Sierra Nevada, Ruby, and Wasatch/Uinta mountain ranges are highlighted.

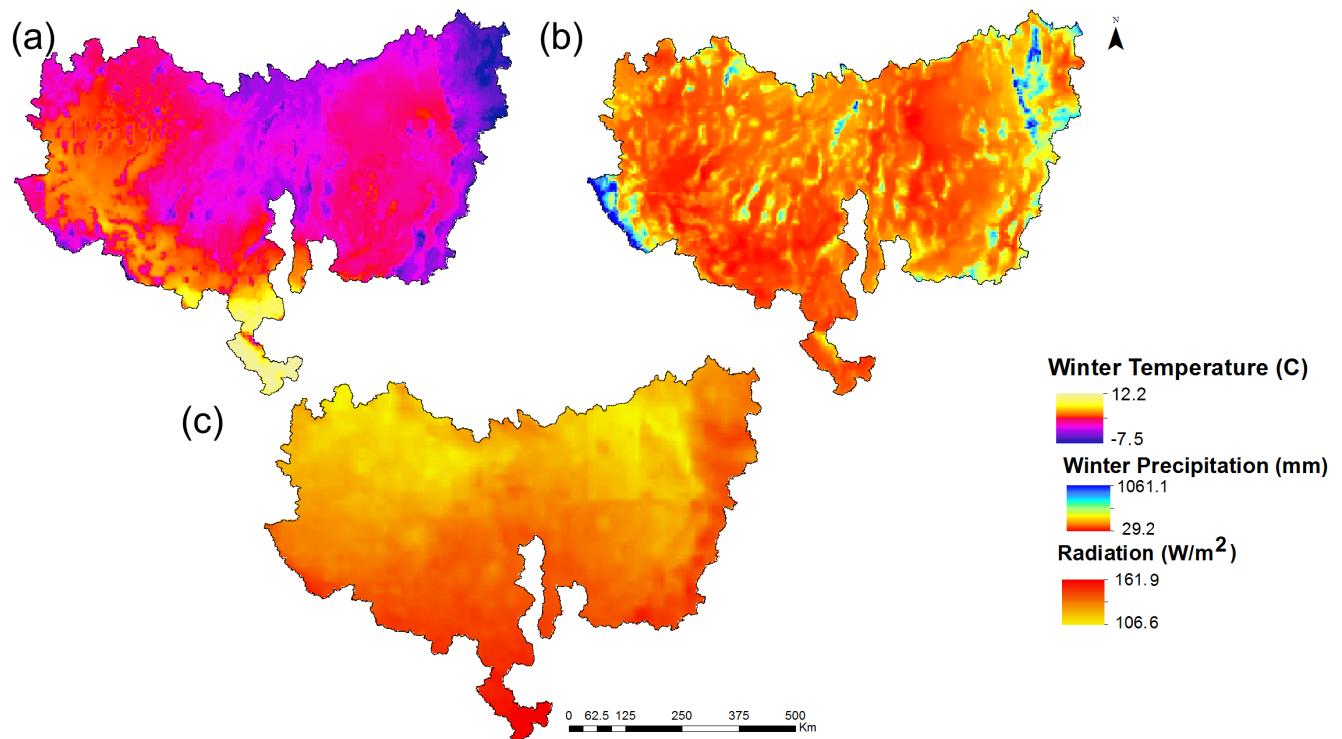



Figure A2. (a) Average winter temperature, (b) average winter precipitation, and (c) average winter radiation across water years 2001-2015 in the Great Basin.

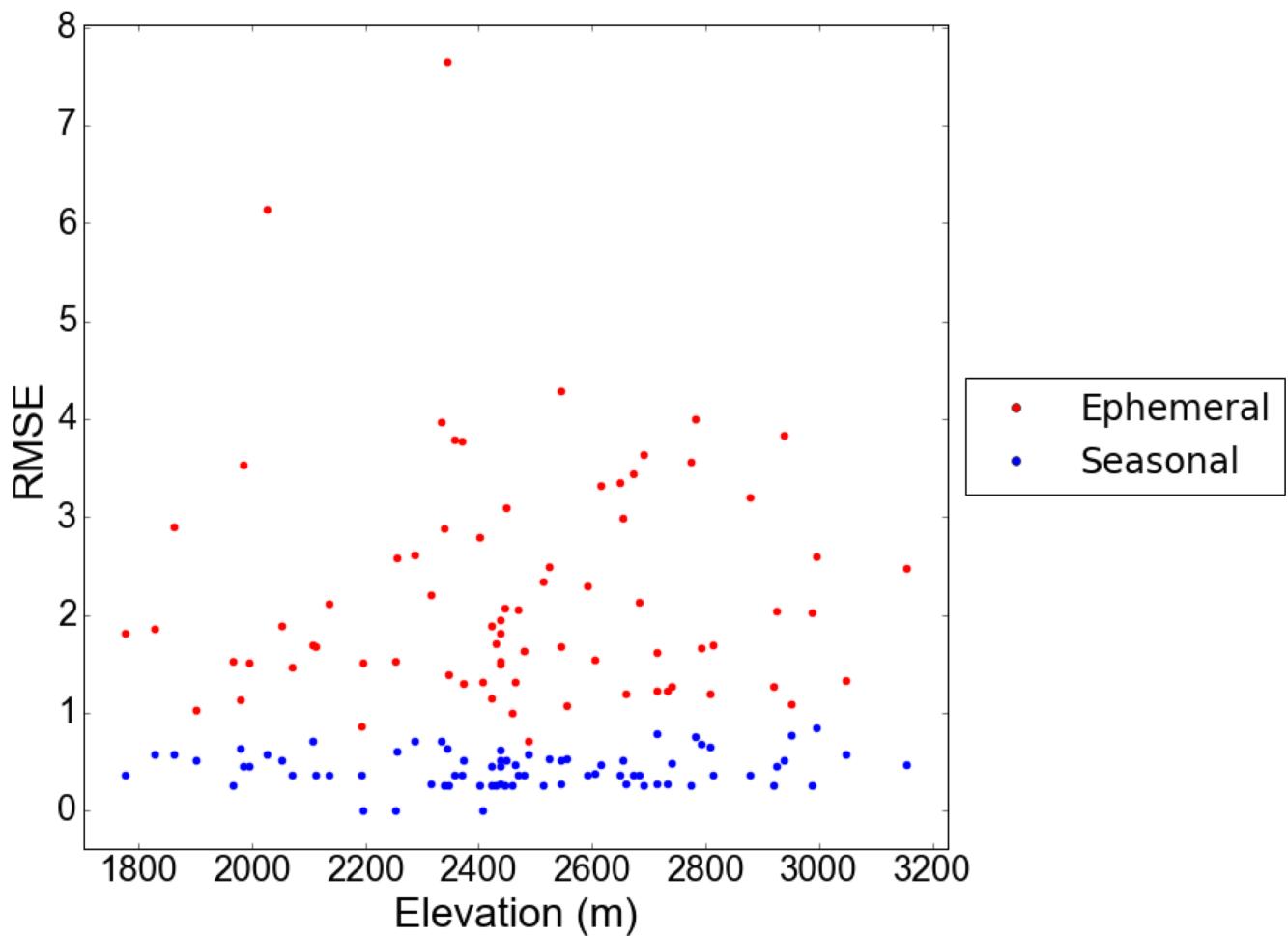
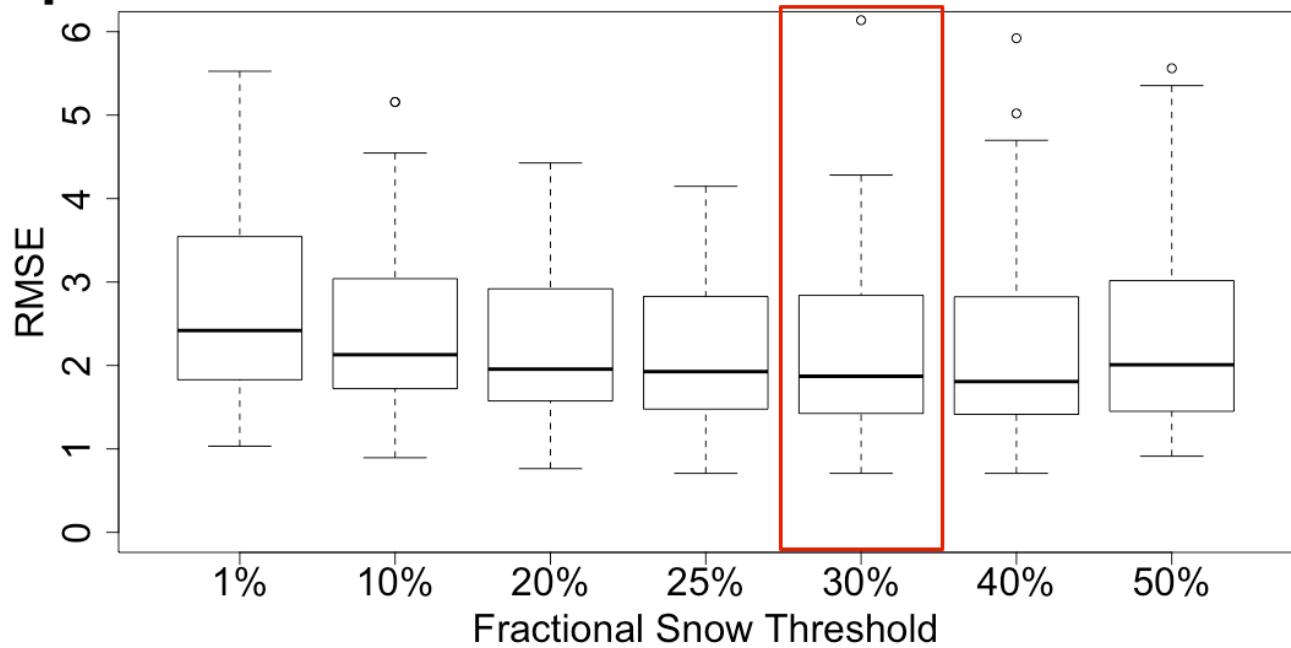



Figure A3. Root Mean Square Errors between the number of observed ephemeral and seasonal snow events at Snow Telemetry (SNOTEL) stations and the number of ephemeral and seasonal snow events derived from the algorithm in Google Earth Engine in each 500 m MODIS pixel corresponding to that station. Measured SWE (Snow Water Equivalent) of 0.3 cm. or greater was used to determine snow presence for SNOTEL sites.

Ephemeral Events

Seasonal Events

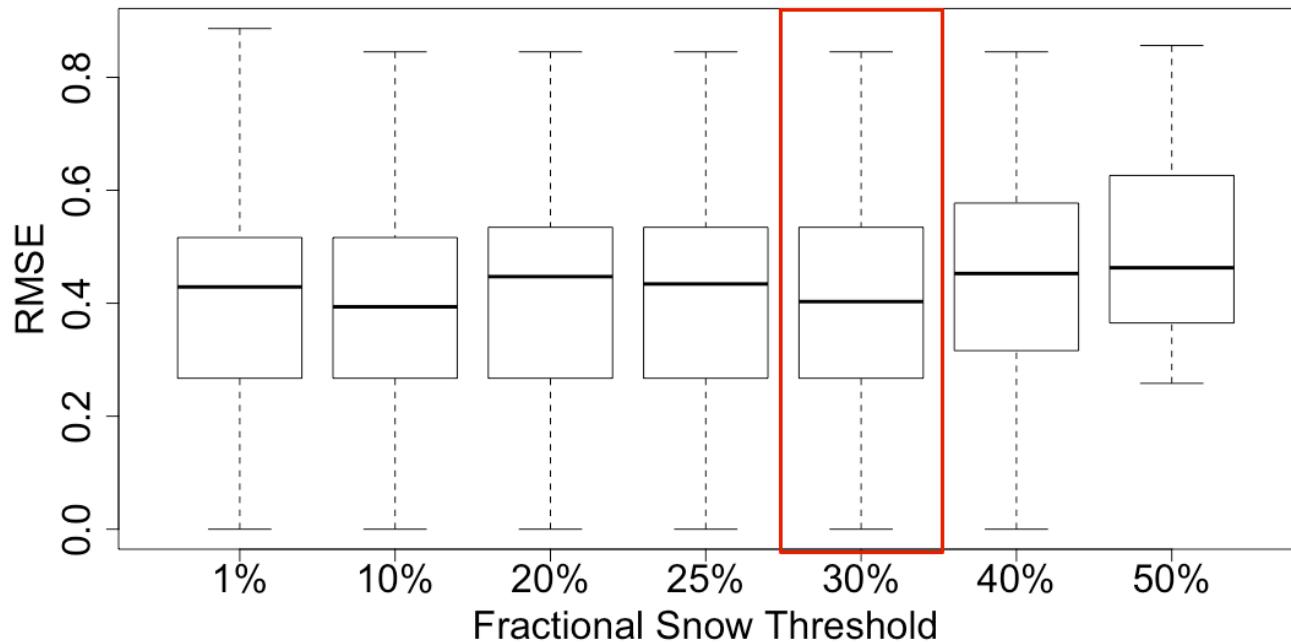



Figure A4. Boxplots depicting the Root Mean Square Errors between the number of observed ephemeral and seasonal snow events at Snow Telemetry (SNOTEL) stations and the number of ephemeral and seasonal snow events derived from the algorithm in Google Earth Engine in each 500 m MODIS pixel corresponding to that station at snow fractions of 1-50 %. 30 % (highlighted in red) was the chosen snow fraction.

Competing interests. The authors declare that they have no conflict of interest.

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