Context-Dependent Effects of Nutrition and Dam Behavior on Neonatal Survival in a Long-Lived Herbivore

A Thesis Presented in Partial Fulfillment of the Requirements for the Degree of Master of Science with a Major in Natural Resources in the College of Graduate Studies University of Idaho

by

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Authorization to Submit Thesis

This thesis of Nicole M. Bilodeau, submitted for the degree of Master of Science with a Major in Natural Resources and titled "Context-Dependent Effects of Nutrition and Dam Behavior on Neonatal Survival in a Long-Lived Herbivore," has been reviewed in final form. Permission, as indicated by the signatures and dates below, is now granted to submit final copies to the College of Graduate Studies for approval.

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Abstract

Behavior represents one of the primary mechanisms by which animals overcome environmental constraints on survival and reproductive success. Females in particular often exhibit plastic behavioral strategies for coping with the different nutritional demands and degrees of susceptibility to predation imposed by gestation, parturition and lactation. Previous studies have demonstrated a link between space-use behavior and important correlates of fitness and have highlighted the value of mechanistic nutritional approaches for understanding the fitness consequences of behavior. However, the mechanisms by which individual responses to variation in the nutritional landscape scale up to influence population performance remain unclear. We quantified relationships among the nutritional landscape (i.e., spatiotemporal variation in forage biomass), dam behavior, and neonatal survival in bighorn sheep (Ovis canadensis). We conducted intensive vegetation sampling and used generalized additive modeling to map the nutritional landscapes available to sheep during summer (May–September) in three population ranges in Idaho: Owyhee River, East Fork of the Salmon River, and Lost River Range. We used GPS collars and lamb surveys to monitor ewe behavior and lamb survival in each study area, and used known-fate survival modeling to test for behaviorally mediated effects of nutrition on lamb survival. Relationships among the nutritional landscape, ewe behavior, and lamb survival were context dependent and varied among study sites. In the Lost River, where lamb survival was highest (83.9%), probability of lamb survival increased when ewes traded access to rugged terrain for access to higher forage biomass. We observed the opposite pattern in the East Fork (i.e., probability of lamb survival increased when ewes traded access to forage for access to rugged terrain), however, and in the Owyhee no metric of ewe behavior was significantly related to the probability of lamb survival. We also observed a strong, positive relationship between spring nutritional condition and probability of lamb survival across study sites. Our research helps to establish mechanistic links among habitat heterogeneity, individual space-use behavior, and reproductive success in bighorn sheep, and underscores the fundamental importance of nutrition as a driver of ungulate performance. Continuing to improve our understanding of such relationships will provide valuable insights for managers and conservationists, and will aid in accurately parameterizing models of population dynamics. Maximizing the usefulness of such models requires knowledge of the mechanisms that underpin variation in population

demographics, and nutritional-ecological approaches like those used in our study shed important light on those mechanisms.

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Table of Contents

Authorization to Submit Thesis	ii
Abstract	iii
Acknowledgments	v
Table of Contents	viii
List of Tables	. ix
List of Figures	xii
CONTEXT-DEPENDENT EFFECTS OF NUTRITION AND DAM BEHAVIOR ON	
NEONATAL SURVIVAL IN A LONG-LIVED HERBIVORE	1
Introduction	1
Methods	6
Results	17
Discussion	20
References	25
Tables	36
Figures	47
Appendices	53

List of Tables

Table 1. Number of female bighorn sheep captured and monitored from 2016–2019 in each of	f
three bighorn sheep population management units in Idaho, USA (Owyhee, East Fork, and	
Lost River). Number of sheep monitored included some ewes that were captured in previous	
years and monitored for multiple years	6

Table 6. Competing models ($\Delta AICc < 2$) for explaining variation in the probability of lamb survival during summer (May–September) as a function of available forage biomass (total or

Table 8. Competing models (Δ AICc < 2) for explaining variation in the probability of lamb survival during summer (May–September) as a function of available forage biomass (total or accepted; see Methods) and terrain ruggedness at random locations generated within the home ranges of GPS-collared bighorn sheep in three population ranges in Idaho, USA from 2016– 2019. The sheep-year (i.e., data from one GPS-collared sheep in one year) was considered the unit of replication for the analysis. Models were fit using the known-fate modeling framework in Program MARK and are ranked according to Akaike's Information Criterion corrected for small sample size (AICc); model weights (*wi*) are also shown. Variable definitions are as follows: Max_AB = Maximum accepted forage biomass; Mean_AB = Mean accepted forage biomass; CV_AB = Coefficient of variation for accepted forage biomass; CV_TB = Maximum total forage biomass; Mean_TB = Mean total forage biomass; CV_TB = Coefficient of variation for total forage biomass; Max_Ruggedness = Maximum terrain

ruggedness; Mean_Ruggedness	= Mean terrain ruggedness; CV_Ruggedness = Coefficient of
variation for terrain ruggedness	

List of Figures

Figure 1. Bigh	orn sheep popul	ation ranges in I	daho, USA, w	here we conducte	d our study
	••••••				47

Figure 4. Boxplots of predicted (based on generalized additive models of the nutritional landscape; see Table 2) total and accepted forage biomass at random locations versus locations used by GPS-collared female bighorn sheep in each of three population ranges in Idaho, USA. GPS location data were collected during summer (May–September), 2016–2019

Figure 6. Relationship between summer (May–September) lamb survival and spring body	
condition (quantified by % ingesta-free body fat) of female bighorn sheep across three	
bighorn sheep population ranges in Idaho, USA during 2016–2018	52

CONTEXT-DEPENDENT EFFECTS OF NUTRITION AND DAM BEHAVIOR ON NEONATAL SURVIVAL IN A LONG-LIVED HERBIVORE

Introduction

Understanding the complex mechanisms that drive variation in population abundance across space and time is a fundamental goal of population ecology and management (Krebs 2002). A multitude of factors, both extrinsic and intrinsic, can influence reproduction, recruitment, and survival in wildlife populations (Caughley and Krebs 1983, Clutton-Brock and Pemberton 2004), and quantifying those factors is therefore critical for understanding and predicting patterns of population performance. For example, demographic variation can be influenced by population density, intra- and interspecific competition, predation, habitat quality, resource availability, disease, or climatic variation (Gaillard et al. 1998). Many of these factors are challenging to quantify, however, and the mechanisms that underpin their effects on wildlife populations are not always intuitive.

Nutrition integrates a variety of biotic and abiotic factors that influence fitness (Humphries et al. 2004, Parker et al. 2009), and thus nutritional ecology can provide important, mechanistic insights into the drivers of population dynamics. Nutritional condition is defined as the state of body components that are controlled by nutrition and in turn influence future survival and reproduction (Harder and Kirkpatrick 1994, Saltz et al. 1995). Nutritional condition can have a wide range of impacts on the physiology and productivity of ungulates (Cook 2002), including effects on fecundity (Verme and Ullrey 1984; Cook et al. 2001, 2005; Tollefson et al. 2010; Morano et al. 2013), timing of parturition (Hass 1997, Cook et al. 2005), neonate birth mass and survival (Thorne et al. 1976, Côté and Festa-Bianchet 2001, Long et al. 2016), juvenile growth rate and mass (Cook et al. 1996), adult mass gain (Morgantini and Hudson 1989), and adult survival (Parker et al. 2009, Monteith et al. 2013). Nutrition can also modulate wildlife population dynamics via effects on life-history traits and patterns of behavior (Franzmann 1985, Harder and Kirkpatrick 1994). For example, Monteith et al. (2011) showed that individual traits such as age, reproductive status, and nutritional condition affected the timing of seasonal migration in mule deer.

Over the past several decades, researchers have developed a variety of methods for quantifying nutritional condition of ungulates based on measurements of body mass, fat reserves, protein reserves, or some combination of these (Stephenson et al. 1998, Cook et al.

2001a, Mysterud et al. 2001, Monteith et al. 2013). Cook et al. (2003) concluded that direct measurements of fat deposits are the most accurate and reliable method for estimating nutritional condition of ungulates. Accordingly, Cook et al. (2001a, 2010) developed and validated equations for predicting percent ingesta-free body fat (%IFBF) of ungulates from measurements of subcutaneous rump fat thickness (obtained using ultrasonography; Stephenson et al. 1998, 2002) and standardized palpation scores. Percent IFBF reflects the amount of stored energy reserves available to individual ungulates for allocation to growth, reproduction, immune function, etc., and is thus a useful metric for understanding the physiological causes and consequences of animal behavior (Parker et al. 2009).

Capital-breeding ungulates subsidize the cost of reproduction with energy stores accrued prior to the breeding season (Festa-Bianchet et al. 1998, Harrison et al. 2011). Allocation to reproduction, however, must be balanced against the need to retain adequate reserves for survival in stochastic environments, and is thus 'risk-sensitive' (Festa-Bianchet et al. 1998, Monteith et al. 2013). Risk-sensitive allocation is based on the premise that partitioning of endogenous energy reserves by an individual is state-dependent, and that there are seasonal thresholds of energy reserves that must be reached to facilitate investment in reproduction (Monteith et al. 2013). Accordingly, individuals entering the breeding season with greater reserves tend to have higher reproductive success, presumably because they have surplus energy to dedicate to breeding, gestation, and lactation (Cook et al. 2004, Stephens et al. 2009).

Nutritional condition reflects both the quality and quantity of available forage, and deficiencies in either of these traits on summer and autumn ranges can negatively affect population performance of ungulates. Nutrient deficiencies and concomitantly poor condition can inhibit ovulation and implantation (Bronson 1989, Frisch 1984, Gunn and Doney 1975, Kincaid 1988, Neville and Neathery 1974, Robbins 1983), and can potentially result in repeated cycling that may delay parturition. Several studies have shown that the probability of conceiving and carrying a fetus to term is strongly influenced by summer forage conditions and autumn body mass (Cameron et al. 1993, Gerhart et al. 1996, Pekins et al. 1998, Cook et al. 2004). Females in poor condition may also have lower milk production, resulting in slower neonatal growth (Oftedal 1985), which may predispose neonates to early death from a variety of sources (Rachlow and Bowyer 1991, Roffe 1993, Côté and Festa-Bianchet 2001, Long et

al. 2016). Even small differences in forage quality can have an outsized influence on animal performance via multiplier effects (White 1983). Cook et al. (2004) demonstrated such an effect in captive female elk and concluded that ruminants cannot compensate for low forage quality simply by eating more. Those authors' results revealed that even a moderate level of nutrition significantly delayed conception, and a low level of nutrition effectively precluded pregnancy of most females.

Behavior represents one of the primary mechanisms by which animals overcome environmental constraints on survival and reproductive success (Krebs and Davies 1997, Huey et al. 2003). Females in particular often exhibit plastic behavioral strategies for coping with the different nutritional demands and degrees of susceptibility to predation imposed by gestation, parturition and lactation (Long et al. 2009). Merems et al. (2020) revealed a significant relationship between use of the nutritional landscape and early-winter condition of female deer. They concluded that individuals who used, on average, areas that provided greater biomass of preferred forage plants during spring and summer entered winter in better nutritional condition. Another recent study demonstrated that selection for areas that provided relatively high-quality forage during autumn had a positive effect on the accumulation of fat reserves (i.e., nutritional condition) by lactating female elk (Long et al. 2016). Those authors' highlighted the positive relationship between maternal nutritional condition in late winter and the probability of neonate survival. These studies add to the growing body of literature linking space-use behavior to important correlates of fitness and highlight the value of nutritional approaches to understanding the fitness consequences of behavior (Parker et al. 2009, Cook et al. 2010, Monteith et al. 2013, Long et al. 2016).

Bighorn sheep (*Ovis canadensis*) are an iconic species of the Rocky Mountains and are an important resource both ecologically and economically (Gordon et al. 2004). They are the largest herbivore in much of the habitat they occupy, and they have important effects on habitat structure. They also serve as an important source of prey for large predators (Festa-Bianchet 1999). Economically, bighorn sheep are a major resource for local communities and government agencies. In Idaho, harvest tags directly contribute hundreds of thousands of dollars to big game conservation, research, and management. Furthermore, indirect income is generated from sheep hunting activities, which includes monies spent by hunters on travel, food, lodging, outfitters and guides, and taxidermists (IDFG 2010). In addition to their

economic value, bighorn sheep are historically and culturally significant to Native Americans for tools, subsistence, and ceremonies (Demarchi et al. 2000).

In recent decades populations of bighorn sheep have declined throughout North America largely due to infectious respiratory disease (Cassirer et al. 2017). The most prevalent disease among bighorn is pneumonia. Pneumonia is polymicrobial, but is typically initiated by the bacterium *Mycoplasma ovipneumoniae* (*M. ovi.*). *M. ovi.* is host-specific to Caprinae and is commonly carried by domestic sheep and goats without affecting their health (Foreyt and Jessup 1982). However, many bighorn sheep populations in the lower 48 states have experienced all-age die-offs after contracting the disease (Western Association of Fish and Wildlife Agencies Wild Sheep Working Group 2012). Once introduced, *M. ovi.* can persist in a bighorn sheep population for decades (Cassirer et al. 2017). Moreover, persistently infected populations have a high likelihood of prolonged periods of disease in lambs, which often reduces recruitment in subsequent years and limits population growth (Ryder et al. 1992; Enk et al. 2001; Smith et al. 2014, 2015).

All-age outbreaks are usually associated with significant population declines, but mortality rates from pneumonia vary widely, and factors influencing disease severity are not well understood (Hobbs and Miller 1992). Researchers continue to seek evidence of host genetic resistance, which might be expected in populations that are successful even in the long-term presence of pneumonia, but to date a genetic basis for resistance to pneumonia has not been identified (Gutierrez-Espeleta et al. 2001, Boyce et al. 2011, Cassirer et al. 2017). Wildlife management agencies have used a variety of strategies (e.g., population eradication, culling individuals observed with symptoms, translocations, etc.) to reduce the occurrence of respiratory disease outbreaks, but to date no specific strategy has successfully stopped an outbreak, and there is no evidence that intervention has consistently reduced morbidity, mortality, or spread of disease (Cassirer et al. 2017). Similarly, no vaccine or antibiotic treatment has successfully reduced infection or controlled the spread of disease in domestic or wild sheep (Cassirer et al. 2017). Inconsistencies in the frequency and severity of outbreaks, coupled with a lack of evidence for genetic resistance, suggest that there may be other underlying mechanisms (e.g., nutrition) contributing to the frequency and severity of die-off events from pneumonia.

One of the best-studied systems for shedding light on interactions among nutrition, disease, and immunity is the feral Soay sheep population in Scotland, which experiences periodic crashes of up to 50% due to the interactive effects of nutritional restriction during harsh winters (Clutton-Brock and Pemberton 2004) and gastrointestinal parasite loads (Coltman et al. 1999; 2001). Sheep that survive harsh winters may invest more in anti-parasite immunity, allowing them to slow the rate of decline in condition over winter relative to less-resistant conspecifics (Nussey et al. 2014). However, the benefits of increased immunity come at the expense of reduced reproductive performance following a harsh winter (Graham et al. 2010), highlighting the degree to which such tradeoffs can be mediated by nutrition.

The goal of our research was to evaluate the impact of (1) the nutritional landscape (i.e., spatiotemporal variation in the availability of forage plants), and (2) inter-individual variation in how bighorn sheep use the nutritional landscape, on lamb survival. We aimed to provide managers with a dynamic model of relationships among the nutritional landscape, ewe behavior, and lamb survival that could be combined with data on adult survival and demographics to aid in effective management of bighorn sheep populations. We hypothesized that both spring nutritional condition of ewes and patterns of movement and space use during summer would influence probability of lamb survival during the first four months of life. Accordingly, we predicted that lambs born to ewes in good condition in spring would have a higher probability of surviving their first four months of life. Additionally, we predicted that lambs born to ewes that consistently used the best parts of the nutritional landscape (i.e., areas that provided high forage biomass) available to them would have a higher probability of surviving their first four months of life.

Methods

Study Sites

The Idaho Department of Fish and Game manages big game species in distinct Game Management Units (GMU) that are combined into Population Management Units (PMU) based on population movements, similarity of habitats, and management objectives. We selected three bighorn sheep PMUs as study sites (Fig. 1) based on their distinct bighorn sheep populations and the wide range of habitats they encompassed.

The first PMU included in our study was the Owyhee River (Owyhee), which is located in the southwest corner of Idaho near the border with Oregon and Nevada. The majority of the Owyhee River PMU is designated as wilderness by the Bureau of Land Management (BLM), and the entirety of the Owyhee River is protected under the National Wild and Scenic Rivers Act. The Owyhee is part of the Columbia Plateau and is bisected by the narrow, 150-300 m deep Owyhee River canyon. Portions of the canyon that are not shearwalled support sagebrush (*Artemesia spp.*) and bunchgrasses (*Festuca spp, Pseudoroegneria spicata*), while the riparian areas support narrow bands of lush grasses, rushes (*Juncus spp.*), and sedges (*Carex spp.*). Habitat beyond the canyon rim consists of comparatively homogeneous sagebrush-steppe, which was expected to be of low nutritional value to bighorn sheep. Elevations range from 1,311 to 1,646 m. Our study site included ~604 km² of the 3,473-km² PMU and supported an estimated population of 158 bighorn sheep (IDFG 2017).

The East Fork of the Salmon River (East Fork) PMU is located in central Idaho between Stanley and Challis. The East Fork PMU encompasses predominately roadless land managed by the U.S. Forest Service (USFS) and includes the Wild and Scenic East Fork of the Salmon River corridor and three newly established (2015) wilderness areas (Boulder-White Cloud, Hemingway, and Jim McClure-Jerry Peak Wilderness). The East Fork is characterized by dry chalky cliffs along the river that rise to rocky peaks and spine ridges dotted with high-elevation mountain lakes. Habitat in the East Fork PMU varies from grasses (*Poa spp., Festuca spp., Achnatherum hyemoides*) and shrublands (*Artemesia spp., Purshia spp., Chrysothamnus spp.*) to high alpine forests (*Pinus spp.*) and meadows (*Antennaria spp., Lupinus spp., Phlox spp.*), and was expected to be of moderate to high nutritional value to bighorn sheep. Elevations range from 1,768 to 3,353 m. The East Fork PMU is roughly 1,994 km^2 (our study site included ~549 km^2) and has an estimated population of 102 bighorn sheep (IDFG 2017).

Our third site was the Lost River Range (Lost River) PMU located in Central Idaho, which extends from east of Challis to Arco. The Lost River PMU spans the entire Lost River mountain range, which includes the tallest peaks in the state and consists almost entirely of USFS, BLM, and state lands. The Lost River is characterized by large sweeping valleys that quickly ascend to sagebrush steppe foothills, timbered slopes, and barren ridges that connect rugged peaks. Habitat types in the Lost River PMU are similar to the East Fork PMU, but there are a greater proportion of high-elevation habitats; the Lost River PMU was expected to be of moderate nutritional value to bighorn sheep. Elevations range from 1,676 to 3,859 m. The Lost River PMU is about 4,662 km², of which ~3,186 km² were included in our study, and supports an estimated population of 256 bighorn sheep (IDFG 2017).

Bighorn Sheep Capture and Body Condition

During spring (Feb–March) of 2016–2018 we captured adult female bighorn sheep using a net gun fired from a helicopter (Krausman et al. 1985; Table 1). Typically, captured animals were suspended from a helicopter in sling bags and transported to a staging area for processing. We occasionally processed animals at the capture site if distance to the staging area was >5 km or if body temperature exceeded 41 °C. At the staging area we weighed and aged each sheep, collected biological samples (blood, fecal, and nasal swabs), obtained morphological measurements (horn length, horn basal circumference, neck circumference, chest girth, and hind foot length), and quantified nutritional condition and reproductive status using a combination of ultrasonography and manual palpation (Stephenson et al. 1998, 2002). We weighed ewes in sling bags using a tripod with either a digital or spring scale, and true weight was calculated by subtracting the weight of the sling bag after removing the ewe. We also fit each sheep with a GPS collar (Lotek: Iridium or Lifecycle Pro, Ontario, Canada; Vectronic Aerospace: Survey 1D, Berlin, Germany; Advanced Telemetry Systems, Minnesota, USA) programmed to record locations every 3 hours from the beginning of May to the end of August, and to remotely upload them 4 times per day. We affixed numbered, colored tags to the belting of each GPS collar to facilitate visual identification of ewes during field monitoring. We also ear-tagged each ewe with a small rototag inscribed with a unique

ID number. We estimated age based on patterns of tooth eruption and replacement up to 4 years old (Mahon 1975, Lawson and Johnson 1982). It is not reliable to age bighorn ewes >4 years old without extracting a tooth and counting cementum annuli (Turner 1977), so ewes estimated to be >4 years were aged as "4+".

We collected nasal and oral pharyngeal (OP) mucus swabs to test for disease, specifically *Mycoplasma ovipneumoniae*. Nasal swabs were inserted into both nostrils and gently rotated. A PVC tube was then inserted into the mouth to facilitate swabbing of the back of the throat while minimizing the likelihood of oral contamination. We also swabbed the ears of each ewe and visually inspected and scored them for scabies severity on a 0 to 4 scale, with 0 indicating no evidence of scabies and 4 indicating severe infestation. We collected blood samples using a 30-mL syringe with an 18 gauge \times 1" needle inserted into the jugular vein, and samples were partitioned into 5 test tubes for subsequent DNA, disease, and pregnancy analysis. We also collected fecal samples to test for parasites.

We assessed body condition using ultrasonography and palpation scoring (Stephenson et al. 1998, 2002). A portable ultrasound (E.I. Medical Imagine, Ibex Pro) was used to measure maximum subcutaneous fat thickness immediately posterior to the cranial process of the tuber ischium (Maxfat), and maximum thickness of the longissimus dorsi between the 12th and 13th ribs (Stephenson et al. 1998, 2002). If no measurable subcutaneous fat was detected, we calculated body fat using overall body condition scores (BCS). Overall body condition was estimated using a scoring system similar to that described by Cook et al. (2001a). We palpated several key locations on the body and scored each on a scale of 0.5 to 6, in intervals of 0.5, where 0.5 = emaciated and 6 = obese. We calculated percent ingesta-free body fat (%IFBF) from BCS or Maxfat measurements using the equations of Stephenson et al. (2020). Percent IFBF is a measure of body condition that is directly related to the amount of stored energy available for allocation to growth, reproduction, immune function, and survival, and is thus a useful metric for understanding the physiological causes and consequences of animal behavior (Parker et al. 2009, Stephenson et al. 2020). Following processing, study animals were either released at the staging area or returned to the capture site as time and circumstances allowed. Capture and handling procedures followed methods established by the American Society of Mammalogists (Sikes et al. 2016) and were approved by the University of Idaho Institutional Animal Care and Use Committee (IACUC-2017-69).

Lamb Survival Surveys

At the beginning of the lambing period (Owyhee: mid-April–May; East Fork & Lost River: mid-May–June) we conducted aerial surveys using a combination of visual observation and a high-definition infrared (IR) camera mounted to the plane to obtain an initial count of lambs born to collared ewes. An annual summary of the number of ewes monitored at each study site can be found in Table 1. In 2017 we conducted aerial surveys weekly during the first month of the lambing period in the East Fork and the Owyhee; the Lost River was only surveyed once in June and once in July due to weather constraints. During 2018 and 2019 we conducted aerial surveys weekly during the lambing period at all three study sites. After the initial survey period we attempted to locate and observe each collared ewe monthly from the ground through September. Each marked female was observed until it could be determined whether she was accompanied by a lamb. Confirmation required a direct observation of nursing or other behaviors indicative of close association (e.g., nuzzling, grooming, bedding together, etc.).

Forage Sampling

During May–August of 2018–2019 we conducted intensive vegetation surveys to quantify biomass of forage available to bighorn sheep within our three study sites. We used the 30-m USDA Landfire Biophysical Settings Potential Vegetation Type (PVT) layer to stratify each study site into similar vegetation associations for sampling purposes (LANDFIRE 2008; Table 2). Sampling locations in each PVT were then selected using the Generalized Random Tesselation Stratified (GRTS) sampling method (Stevens and Olsen 2004). Sampled PVTs and the proportion of each study site comprised by each PVT are shown in Table 2.

We attempted to sample forage biomass during the peak of plant diversity, the timing of which we determined from a combination of visual observation and data from long-term vegetation phenology plots at each study site. Biomass plots consisted of a 100-m transect with 1-m² quadrats placed every 20 m, beginning at 20 m and ending at 100 m. Within each quadrat we identified all plants to species and assigned each species to one of the following phenological stages: newly emergent, flowering, fruiting, mature, or cured. We then estimated

forage biomass in each quadrat using a double-sampling approach (Bonham 1989). We began by visually estimating percent horizontal cover (% cover) of each plant species within each 1m² quadrat to the nearest 1%. We then selected the 2 most species-rich quadrats along the transect and used standard clip-and-weigh methods to estimate dry biomass of each forage species within those quadrats (Bonham 1989, Butler and Wayne 2007, Proffitt et al. 2016). We clipped all plant species from 2 cm to 1.5 m in height (the approximate maximum foraging height of bighorn sheep), and we collected only leaves and current annual growth from shrubs and trees (all plant parts were collected for graminoids and shrubs). Clipped samples were placed in paper bags and dried in a forced-air convection oven at 100 °C for 24 hrs or until constant weight was achieved, typically within 3 days of collection. When samples could not be dried within this time frame they were stored in a dry, well-ventilated space until they could be processed. We weighed all samples using an Alaide jewelry scale (0.01g). Any samples weighing ≤0.01 g were assigned a trace value of 0.01 g. We sampled a total of 53 transects in the East Fork, 58 transects in the Lost River, and 19 transects in the Owyhee.

We used multiple linear regression (Neter et al. 1996) to fit species-specific predictive equations for estimating forage biomass in all unclipped quadrats as a function of % cover, Julian day, and the interaction between those variables. We did not include tree canopy cover as a covariate in biomass regressions because 70% of transects had 0% canopy cover, and average canopy cover was only 7.5% across transects that did include trees. We evaluated both linear and natural log-transformed terms for each predictor variable to test for potential non-linear relationships between covariates and forage biomass. We fit 8 competing models (see Appendix B) for each plant species for which we had obtained ≥ 10 paired biomass and % cover measurements. Many of our plant species (n = 86/353) had consistently low (i.e., <1%) % cover values, making it impossible to fit a legitimate regression model to the cover/biomass data. In those instances we (1) calculated mean biomass of the species in all clipped quadrats where % cover was $\leq 1\%$, (2) assigned that mean biomass value to all unclipped quadrats where % cover was $\leq 1\%$, and (3) upscaled (linearly) the average biomass value to unclipped quadrats where % cover was >1% (e.g., for a quadrat with 2% cover of the species we multiplied average biomass by 2 and assigned the resulting value to the quadrat). Species with <10 paired biomass and % cover measurements were assigned to a growth-form group

(evergreen shrubs, graminoids, tall forbs, short forbs, basal forbs, cacti, mid-tall shrubs, and evergreen trees) prior to fitting biomass regressions. Similarly, when the intercept-only model was the best-fitting model for an individual species with $n \ge 10$, or the adjusted R^2 of the best model for a species was <0.2, we assigned the species to the appropriate growth-form group prior to regression analysis. If inclusion of the species in the growth-form group either improved or did not significantly reduce ($\le 5\%$ reduction) the R^2 of the best model then the species was retained, and the growth-form model was used to predict biomass of that species in all unclipped quadrats. If inclusion of the species in the growth-form group significantly reduced ($\ge 5\%$ reduction) the R^2 of the best model then the species of all biomass regressions equal to 0 (i.e., regression through the origin). Once the respective species- or group-form-specific biomass regressions were applied to all unclipped quadrats, final biomass (kg/ha) of each plant species at each transect location was estimated as the average biomass of the species across all 5 quadrats along the transect.

We did not have detailed site-specific data on bighorn sheep diets and were therefore uncertain what the most biologically meaningful metric of the nutritional landscape for sheep would be. Accordingly, we used plant species-specific biomass data to generate two candidate response variables for quantifying the nutritional landscape available to sheep: total forage biomass and accepted forage biomass. Total forage biomass was estimated as the summed biomass of all sampled forage plants. Accepted forage biomass was estimated as the summed biomass of plants likely to be selected or used in proportion to their availability (i.e., not avoided) by sheep. We used published diet data (Smith 1954, Johnson 1980, Miller and Gaud 1989, Wagner and Peek 2006, Whitaker 2010) together with expert recommendations (L. A. Shipley, Washington State University; T. R. Stephenson, California Department of Fish and Wildlife, personal communication) to determine which forage species were likely to be accepted by bighorn sheep in each study area (Appendix A). We then used spatiotemporally explicit estimates of total and accepted forage biomass at each transect location as response variables in subsequent models of the nutritional landscape available to bighorn sheep at each study site.

Modeling the Nutritional Landscape

We modeled spatiotemporal variation in total and accepted forage biomass within our three study sites as a function of remotely sensed covariates known to influence vegetation dynamics at broad scales. Candidate predictor variables included the enhanced vegetation index (EVI, an index of vegetation greenness; U.S. Geological Survey, Earthdata), PVT, snowmelt date (Snow Data Assimilation System), monthly total precipitation (ppt; TerraClimate), monthly soil moisture (soil; TerraClimate), monthly average maximum temperature (tmax; TerraClimate), monthly Palmer drought severity index (PDSI; TerraClimate), elevation, slope, aspect, and Julian day (Table 3). The EVI layer was filtered to correct for cloud cover. We also used the Gapfill package in R to predict missing values in the EVI layer introduced by cloud cover, missing orbits, sensor geometry artifacts, or other anomalies (Gerber et al. 2018).

We used Generalized Additive Models (GAMs) to predict spatiotemporal variation in the nutritional landscape (i.e., total or accepted biomass of forage) available to bighorn sheep in each of our three study areas in Idaho. For each study site we modeled variation in total biomass and accepted biomass separately. We followed the approach of Merems et al. (2020) and conducted model selection in two stages to (1) assess which spatiotemporally dynamic variables to retain, and (2) select the functional form of each model for each study site. Our goal was to maximize predictive strength of the best model for each study site. Accordingly, in the first stage of model selection we fit models that contained different combinations of smoothing terms for spatial (x-y coordinates) and/or temporal (Julian day, average maximum temperature, PDSI, total precipitation, soil moisture, or snowmelt date) covariates, combined with fixed effects for all uncorrelated spatial covariates (PVT, aspect [transformed with sine and cosine functions to measure eastness and northness, respectively], slope, and elevation). We fit cubic regression splines for each candidate smoothing term and used cross-validation to determine the optimal amount of smoothing (Zuur et al. 2009). We then used Akaike's Information Criterion (AIC) to determine the most parsimonious combination of smoothing term(s) to carry forward to the second stage of the analysis (based on the model with the lowest AIC score). We limited the number of smoothing terms included in each model to a maximum of two (one spatial and one temporal smoother) to reduce model complexity and facilitate convergence, and because many of the time-dependent variables were highly correlated ($r \ge 0.6$) and could not be included together in the same model.

In the second stage of the analysis we fit the following set of four models for each continuous covariate (all models also included the smoothing term/s brought forward from the first stage of analysis): (1) untransformed covariate; (2) covariate²; (3) ln(covariate); and (4) covariate removed. Each model thus specified a different functional form of the relationship between the covariate and forage biomass (except the last model, in which the covariate was absent). We used AIC to determine whether the covariate should be retained at all, and if so, whether a transformation was appropriate. Following selection of continuous covariates and their optimal functional form, we removed PVT, a categorical covariate with 6 levels in the Lost River and East Fork and 4 levels in the Owyhee (Table 2), from the model to determine whether PVT should be included in the final predictive model for each study site (based on whether inclusion of PVT resulted in a lower AIC score).

Following model selection we used the gam.check function in R to evaluate adherence of the final model to assumptions, and to identify outliers. Those diagnostics sometimes indicated that the assumption of homogeneity of variance was not satisfied, and that differences in variance among PVTs were the source of the problem (i.e., there was significantly more variation in forage biomass in some PVTs than others). When this occurred, we used the varIdent variance structure (Zuur et al. 2009) to calculate unique variance estimates for each PVT, and to fit a weighted GAM in which the inverse of the variance in forage biomass in each PVT was used as the weighting factor. We then repeated the second stage of the analysis with the weighted model. Lastly, we used the CVgam function in R to calculate cross-validation statistics for each of the six final models (see Table 3 for candidate predictor variables; see Table 4 for best models). Those models were then applied to the three study sites to calculate spatiotemporally explicit estimates of predicted total and accepted forage biomass available to bighorn sheep during the study from 2017-2019. We fit all models using the mgcv package in program R v4.0.2 (R Development Core Team 2019; Wood 2006). We then compared the mean accepted and total forage biomass in PVTs that comprised >0.2% of each study site (Fig. 2).

We extracted model-predicted values of total and accepted forage biomass to both used (i.e., locations obtained from GPS-collared bighorn ewes) and random (i.e., locations that represented habitat availability) locations in each respective study site and year. We restricted our subsequent analyses of used locations to those that occurred in sampled PVTs (Table 2; 99%, 87%, and 91% of locations obtained in the East Fork, Lost River, and Owyhee study sites, respectively). To quantify variation in the availability of total and accepted forage biomass we generated random locations equal to the number of used locations we obtained from collared sheep (n = 118) at each study site (n = 48,514 in the Lost River, 40,807 in the East Fork, and 9,575 locations in the Owyhee, respectively). Random locations were generated within 100% minimum convex polygons (MCPs) derived from the complete sample of used locations at each study site. We used extracted values of total and accepted forage biomass at random locations to estimate the proportion of each study site that fell into each of four quartiles of habitat quality (Fig. 3; High \geq 705 kg/ha, Mid-High = 423–704 kg/ha, Mid-Low = 246–422 kg/ha, Low \leq 245 kg/ha). We also compared the distribution of predicted total and accepted biomass values between random locations and locations that were used by sheep at each study site (Fig. 4).

Quantifying Individual Variation in Space Use

We used the adehabitatHR package in R to estimate 95% fixed-kernel utilization distributions (UD) and associated home-range isopleths for individual ewes during summer of each year (Silverman 1986, Worton 1989). We used the ad hoc method for bandwidth selection, which is designed to prevent under-smoothing in kernel home-range analysis (Kie 2013). We estimated ad hoc bandwidths by incrementally reducing the reference bandwidth by 10% in successive steps until the 95% contour fragmented into two or more polygons (Berger and Gese 2007, Jacques et al. 2009). We excluded one (of 117) ewe from these analyses because we obtained <50 GPS locations during the period when she was collared.

We quantified the nutritional quality of each ewe's home range by casting random points within individual home range polygons and extracting the underlying values of total and accepted biomass to each point. We determined the minimum number of random locations necessary to accurately represent available habitat in each home range using the methods of Long et al. (2014). Because access to rugged escape terrain is a well-known determinant of space use by bighorn sheep (Hansen 1980, Festa-Bianchet 1988, Smith et al. 1991, Rachlow and Bowyer 1998), we also extracted terrain ruggedness values to the random locations within each ewe's home range. We quantified terrain ruggedness using the methods outlined by Sappington et al. (2007). Briefly, this approach quantifies ruggedness by measuring the dispersion of vectors orthogonal to the terrain surface, which combines variation in slope and aspect into a single measure that provides a more accurate representation of terrain heterogeneity than indices based only on slope or elevation. The resulting values are low both in flat areas and in extremely steep areas, but values are high in areas that are both steep and variable. After total and accepted forage biomass and terrain ruggedness values had been extracted to all of the used and random locations within each home range, we calculated a series of descriptive statistics designed to represent (1) the overall 'quality' of each ewe's home range with respect to forage and escape terrain (descriptive statistics derived from random locations), and (2) variation in use of the nutritional landscape and escape terrain among individual ewes at each study site (descriptive statistics derived from used locations). For each variable (total forage biomass, accepted forage biomass, and terrain ruggedness) we calculated the mean, max, and coefficient of variation across random locations within each home range, and the mean and max values across used locations within each home range. Those metrics were then used as candidate predictor variables in subsequent models of lamb survival as a function of ewe behavior.

Modeling Lamb Survival

Our ultimate goal was to evaluate the influence of ewe behavior (i.e., use of the nutritional landscape and escape terrain) and nutritional condition on lamb survival during summer (May–September) at each of our three study sites. Accordingly, the unit of replication for our analyses was the ewe-year. Our GPS location dataset included 156 ewe-years from 2016–2019 derived from 97 unique individuals (some individuals were monitored for multiple years). We used the known-fate model in Program Mark (White and Burnham 1999) to model lamb survival at each study site as a function of the descriptive statistics representing use of total biomass, accepted biomass, and terrain ruggedness by each ewe. We fit all possible combinations of candidate covariates in separate model sets for each study site with two exceptions: (1) correlated ($|r| \ge 0.6$) pairs of variables were not included in the same model, and (2) descriptive statistics for total biomass and available biomass were never included in the same model. In addition, we fit separate model sets for evaluating the influence of landscape-scale behavioral choices (covariates derived from random locations within ewe home ranges) versus home-range scale choices (covariates derived from used locations) on

lamb survival at each site. Models were ranked by AICc and inferences were based on models with $\Delta AICc < 2$ (Burnham and Anderson 2002).

We were only able to obtain spring body condition measurements for a subset of our ewe-years (n = 77), and thus we did not have sufficient sample size to include spring condition as a covariate in site-specific models of lamb survival. Therefore, we conducted a separate known-fate survival analysis in which we combined data across sites and modeled lamb survival as a function of spring condition.

Results

Nutritional-Landscape Modeling

Generalized additive models (GAMs) for explaining spatiotemporal variation in accepted or total forage biomass generally performed well for each of our three study sites, with adjusted R^2 values ranging from 0.28 (East Fork total biomass model; Table 4) to 0.63 (Lost River accepted biomass model; Table 4). Top models for accepted biomass generally had higher predictive power (all Adj. $R^2 > 0.49$) than top models for total biomass across study sites. Smoothing terms and spatial covariates included in top models varied considerably across sites and between response variables (i.e., between models of accepted versus total biomass; Tables 4 and 5). All top models for the Lost River and Owyhee study areas included a temporal smoother for Julian day, whereas neither Julian day nor any other temporal smoothing term were retained in top models for the East Fork (Table 5). The enhanced vegetation index (EVI, an index of vegetation greenness) or its square were included in 5 of the 6 top models across study sites, and EVI was always positively related to forage biomass (Table 5). In contrast, the influence of topography and potential vegetation type on forage biomass was more variable (Table 5). Top models of accepted versus total forage biomass in the Owyhee were nearly identical because these two measurements were equivalent at most transect locations (i.e., all sampled plants along most transects were classified as being acceptable to bighorn sheep; Table 4).

Predicted forage biomass varied among PVTs at each site but was more variable in the Lost River and East Fork (range $\approx 375-1,000$ kg/ha across PVTs) than in the Owyhee (range $\approx 400-800$ kg/ha; Fig. 2). With the exception of the Owyhee, where measurements of total and accepted biomass were typically equivalent, the proportion of each site classified as high-or mid-high-quality (top two quartiles) foraging habitat was greater based on total forage biomass than when accepted biomass was used as the metric of habitat quality (Fig. 3). Surprisingly, the proportion of each study site falling into the top two quartiles of total forage biomass was roughly equal across all three sites (Fig. 3). However, the relative proportion of high-quality habitat (based on total forage biomass) was higher in the East Fork and Lost River than in the Owyhee (Fig. 3). Also surprising was our observation that mean values of both accepted and total forage biomass were consistently lower at locations used by female sheep than at random locations generated within the boundaries of each study site (Fig. 4).

This result suggests that sheep at all three sites consistently used lower-quality foraging habitat than was generally available. This trend was most pronounced in the Lost River and least pronounced in the Owyhee (Fig. 4).

Effects of Nutrition on Lamb Survival

Raw estimates of lamb survival probability were comparable to estimates derived from known-fate modeling (Fig. 5) and were highest in the Lost River ($83.9\% \pm 0.05 SE$), intermediate in the East Fork ($61.0\% \pm 0.06$), and lowest in the Owyhee ($51.5\% \pm 0.09$). At the home-range scale (i.e., locations used by GPS-collared sheep within their home ranges), use of the nutritional landscape by female sheep had a greater influence on lamb survival in the Lost River than at the other two sites, evidenced by the null model outperforming all other competing models of lamb survival in the East Fork and the Owyhee (Table 6). The top model for predicting summer lamb survival in the Lost River included a single nutritional covariate, max_TB (maximum total forage biomass at locations used by sheep; Table 6). Moreover, that covariate was included in three of the top four models (models with $\Delta AICc < 2$; Burnham and Anderson 2002; Table 6) for the Lost River, as was terrain ruggedness (either mean or maximum values of terrain ruggedness at locations used by female sheep; Table 6). Use of higher-quality foraging habitat (i.e., locations with higher maximum values of total forage biomass) in the Lost River was positively related to the probability of lamb survival, whereas use of more rugged terrain was negatively related to lamb survival (Table 7).

At the landscape scale, the overall quality of ewe home ranges (assessed based on random locations generated within each ewe's home range; see Methods) with respect to both the nutritional landscape (total forage biomass) and terrain ruggedness was related to the probability of lamb survival in the Lost River and East Fork, but the null model was once again the top model for the Owyhee (Table 8). The top model for predicting lamb survival as a function of home-range quality in the Lost River included mean_TB (mean total forage biomass at random locations within sheep home ranges) and CV_Ruggedness (the coefficient of variation for terrain ruggedness, considered to be a measure of terrain heterogeneity within sheep home ranges; Table 8). The top model in the East Fork included only mean_TB, but the next-best model (the only other model in the set with $\Delta AICc < 2$) included both mean_TB and CV_Ruggedness (Table 8). Interestingly, the signs of the coefficients for these two covariates

differed between the Lost River and the East Fork; mean_TB was positively related to lamb survival in the Lost River but negatively related to lamb survival in the East Fork, whereas CV_Ruggedness was negatively related to lamb survival in the Lost River and positively related to lamb survival in the East Fork (Table 9).

When we combined data across study sites for the subset of ewes for which we had estimates of spring nutritional condition, we found a strong, positive relationship between spring condition and probability of lamb survival (Fig. 6). Model results indicated that lambs born to ewes in relatively good condition (18% ingesta-free body fat) were roughly three times more likely to survive the summer months than lambs born to ewes in poor condition (3% ingesta-free body fat; Fig. 6).

Discussion

Relationships among the nutritional landscape, ewe behavior, and lamb survival were context dependent. Although female sheep at all three sites consistently used lower-quality foraging habitat than was generally available, this trend was most pronounced in the Lost River. Yet, lamb survival was highest in the Lost River population, and this was the only site where use of the nutritional landscape by ewes at both the home-range and landscape scales positively influenced lamb survival (i.e., when ewes consistently used locations with greater forage biomass their lambs had a higher probability of survival). This result suggests that habitat heterogeneity plays a fundamental role in contextualizing relationships between behavior and fitness. Spatial variation in total forage biomass was much higher in the Lost River than at the other two sites, which likely led to the correspondingly greater variation in ewe behavior we observed at that site (Fig. 4). It is challenging for animals to consistently optimize their behavior (Belovsky 1984, Kie 1999), and inter-individual variation in behavior often increases as habitat heterogeneity increases (Morales et al. 2005; van Beest and Milner 2013; Long et al. 2014, 2016). Our study, grounded in the principles and techniques of nutritional ecology, provides mechanistic support for the hypothesis that inter-individual variation in behavior can have important fitness consequences that could eventually scale up to influence population performance in heterogeneous landscapes (Stephenson et al. 2020).

Our results also suggest that sheep in alpine habitats exhibited context-dependent strategies for coping with tradeoffs between forage availability and vulnerability to predation. Probability of lamb survival in the Lost River increased when dams consistently used locations that provided high forage biomass but reduced access to escape terrain at both the home-range and landscape scales (although the specific metrics identified during model selection differed between scales; Tables 7, 9). In contrast, ewes in the East Fork that positioned their home ranges in areas that provided less forage biomass but greater access to heterogeneous escape terrain had higher lamb survival. Trade-offs between forage and predation risk are well-known in bighorn sheep (e.g., Festa-Bianchet 1988), but our results suggest that the nature and magnitude of such tradeoffs may be modulated by thresholds in the relevant risk factors (i.e., context-dependent variation in which factors are most limiting). Although escape terrain has consistently been identified as an essential component of habitat for mountain sheep (Geist 1971, Krausman and Leopold 1986, Bleich et al. 1997), low overall

risk of predation combined with a heterogeneous nutritional landscape may relax reliance of sheep on escape terrain while increasing the benefits of trading access to rugged terrain for access to forage. We had no data on predator densities at any of our study sites, but the high rate of lamb survival in the Lost River population suggests that that population is not currently limited by predation. Moreover, because the proportion of the nutritional landscape that fell into each quartile of total forage biomass was comparable between the Lost River and the East Fork (Fig. 3), differences between those sites in vulnerability of sheep to predation are more likely to explain our results than differences in forage availability. Empirical work designed to identify the conditions under which large herbivores like bighorn sheep begin to adjust their responses to tradeoffs between forage and perceived predation risk will be a fruitful avenue for future research.

We failed to detect any significant relationships among variation in the nutritional landscape, ewe behavior, and lamb survival in the Owyhee population at either the homerange or landscape scales. One simple explanation for this null result is our smaller sample size at that site (n = 54 ewe-years in the Owyhee, $\geq 40\%$ fewer than the East Fork and Lost River), which stemmed from a combination of inclement weather restricting capture operations and GPS-collar malfunctions. An alternative explanation, however, relates to the disease status of the Owyhee population and the potential for disease to have overridden more nuanced relationships among nutrition, behavior, and fitness. Lamb recruitment is severely affected by the presence of disease in bighorn sheep herds (Cassirer et al. 2017). In the Owyhee, an all-age pneumonia outbreak in 2015–16 led to a large die-off of sheep in the canyon (Dennehy 2017). For example, one population (the Lower Owyhee River Canyon population) declined from 384 in 2015 to 111 in 2016, and most recently to 92 in 2019 (ODFW 2020). Accordingly, biologists are concerned about lingering effects of the outbreak related to chronic infection in adults and decreased lamb survival (Dennehy 2017). Our test results from captured sheep in the Owyhee in 2016–18 resulted in two positive detections for Mycoplasma ovipneumoniae, although these results do not indicate whether or not a herd is currently infected. However, testing rates in the Owyhee were low (e.g., only 5 sheep were tested in the Owyhee in 2018 whereas 44 were tested in the Lost River that year), and thus additional testing would be necessary to more accurately determine disease status of the Owyhee population.

Although relationships among the nutritional landscape, ewe behavior, and lamb survival were context dependent and varied among sites, when we combined data across sites we found a strong, positive relationship between spring condition of ewes and probability of lamb survival over summer. Indeed, within the range of nutritional condition observed in our study, maximizing nutritional condition in spring led to a threefold increase in the probability of lamb survival over summer. These results support our hypothesis and are consistent with previous studies that have shown wide-ranging impacts of nutritional condition on the physiology and productivity of ungulates, including birth mass and neonatal survival (Thorne et al. 1976, Côté and Festa-Bianchet 2001, Monteith et al. 2014, Long et al. 2016). For example, in a recent study of bighorn sheep in particular, Stephenson et al. (2020) demonstrated that nutritional condition was directly related to the nutritional value of forage on occupied ranges and had pervasive effects on overwinter survival and reproductive success. Those authors also reported that nutritional condition of lactating ewes in autumn was positively associated with the finite rate of population increase (lambda), suggesting that individual-level response to the nutritional landscape and their associated effects on fitness components (e.g., nutritional condition and neonatal survival) do scale up to influence population performance.

One common criticism of studies of space-use behavior by animals is that they often fail to produce evidence that those behaviors have tangible fitness consequences (Morrison 2001). Our study was based on the premise that many potential fitness consequences of animal space-use and movement decisions should be energetically mediated, and that individuals that consume more forage should have more discretionary energy available to devote to reproduction (Monteith et al. 2013). Testing this hypothesis requires accurately quantifying spatiotemporal variation in the forage resources available to individuals, and our approach to accomplishing this goal combined intensive vegetation sampling with complex statistical modeling (following Merems et al. 2020). Although our models had relatively high predictive power, however, there were several limitations to our approach that are important to acknowledge. First, our approach focused on modeling variation in forage biomass rather than forage quality. It is possible that accounting for variation in energy and/or protein content of forage would have improved our ability to link patterns of space use to lamb survival. Forage biomass tends to be more variable across space than forage quality, however, and Tveraa et al. (2013) suggested that variation in quantity (i.e., biomass) of forage was more critical to female reproductive success and offspring body mass than variation in forage quality (a notion also supported by Merems et al. 2020). Second, defining what constitutes 'forage' requires knowing which plant species an individual will consume when encountered, and we did not have site-specific diet composition data for any of our study populations. We attempted to overcome this limitation using expert opinion and previously published data for other bighorn sheep populations. However, nutritional landscape maps based on all sampled plants were more useful for linking ewe behavior to lamb survival than maps based on the subset of plants assumed to be eaten when encountered by sheep. This suggests that our attempts to classify forage plants as 'accepted' or 'avoided' were largely unsuccessful, and that future efforts to quantify population-specific patterns of diet composition are warranted. Finally, our approach assumed that sheep that used areas with higher predicted forage biomass did, in fact, consume more forage. There are several possible reasons why this assumption may have been violated. For example, Berger (1978) concluded that bighorn sheep forage less efficiently and interrupt foraging more frequently when foraging in small groups (n < 5), and Rachlow and Bowyer (1998) reported similar results for Dall's sheep. We were not able to account for group size in our analyses. Nevertheless, this limitation should, if anything, add noise to our data and reduce our ability to detect relationships between ewe behavior and lamb survival. Accordingly, we suggest that are results are more likely to be conservatively biased (i.e., higher probability of a type II error) than the alternative.

Our research helps to establish mechanistic links among habitat heterogeneity, individual space-use behavior, and reproductive success in bighorn sheep, and underscores the fundamental importance of nutrition as a driver of ungulate performance. Continuing to improve our understanding of such relationships will provide valuable insights for managers and conservationists, and will aid in accurately parameterizing models of population dynamics. Maximizing the usefulness of such models requires knowledge of the mechanisms that underpin variation in population demographics, and nutritional-ecological approaches like those used in our study shed important light on those mechanisms. Future efforts to build on our work with bighorn sheep would benefit from (1) collaring lambs at birth to monitor survival and cause-specific mortality at finer time scales, (2) measuring forage quality in addition to forage biomass, (3) quantifying diet composition of sheep using DNA
metabarcoding, and (4) recapturing collared individuals at annual intervals to collect repeat measurements of nutritional condition.

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Tables

Table 1. Number of female bighorn sheep captured and monitored from 2016–2019 in each of three bighorn sheep population management units in Idaho, USA (Owyhee, East Fork, and Lost River). Number of sheep monitored included some ewes that were captured in previous years and monitored for multiple years.

	Ow	yhee	East	t Fork	Lost	River
Year	Captured	Monitored	Captured	Monitored	Captured	Monitored
2016	16	12	25	23	0	0
2017	13	23	6	23	14	12
2018	5	15	17	27	44	42
2019	0	4	1	17	0	38
Total	34	54	49	90	58	92

Table 2. Potential vegetation types (PVTs) present in each of three bighorn sheep population ranges in Idaho, USA, and the percentage of each range comprised by each PVT.

Range	Potential vegetation type	ID code	% of range
Lost River	Barren-Rock/Sand/Clay ^a	Barren	4.1
	Inter-Mountain Basins Big Sagebrush Shrubland	IMB_Shrub	12.6
	Inter-Mountain Basins Big Sagebrush Steppe	IMB_Steppe	17.0
	Inter-Mountain Basins Mixed Salt Desert Scrub	IMB_Desert	1.7
	Inter-Mountain Basins Montane Sagebrush Steppe	IMB_MtnSteppe	12.7
	Middle Rocky Mountain Montane Douglas-fir Forest and Woodland ^b	MRM_MDFW	9.5
	Northern Rocky Mountain Dry-Mesic Montane Mixed Conifer Forest ^b	NRM_Dry_MMCF	10.3
	Northern Rocky Mountain Subalpine Woodland and Parkland	NRM_SWP	6.8
	Perennial Ice/Snow ^a	Ice_Snow	0.4
	Rocky Mountain Subalpine Dry-Mesic Spruce-Fir Forest and Woodland	RMSM_Dry_SFW	11.2
	Rocky Mountain Subalpine Mesic-Wet Spruce-Fir Forest and Woodland ^b	RMSM_Wet_SFW	0.8
East Fork	Barren-Rock/Sand/Clay ^a	Barren	1.3
	Columbia Plateau Steppe and Grassland	CPS_Grassland	0.2
	Inter-Mountain Basins Big Sagebrush Shrubland	IMB_Shrub	2.6
	Inter-Mountain Basins Big Sagebrush Steppe	IMB_Steppe	2.4
	Inter-Mountain Basins Montane Sagebrush Steppe	IMB_MtnSteppe	12.6
	Middle Rocky Mountain Montane Douglas-fir Forest and Woodland ^b	MRM_MDFW	13.7
	Northern Rocky Mountain Dry-Mesic Montane Mixed Conifer Forest ^b	NRM_Dry_MMCF	15.1
	Northern Rocky Mountain Subalpine Woodland and Parkland	NRM_SWP	18.6
	Perennial Ice/Snow ^a	Ice_Snow	1.3
	Rocky Mountain Subalpine Dry-Mesic Spruce-Fir Forest and Woodland	RMSM_Dry_SFW	25.0
	Rocky Mountain Subalpine Mesic-Wet Spruce-Fir Forest and Woodland ^b	RMSM_Dry_SFW	1.6
Owyhee	Columbia Plateau Steppe and Grassland	CPS_Grassland	1.4

Inter-Mountain Basins Big Sagebrush Shrubland	IMB_Shrub	50.3	
Inter-Mountain Basins Big Sagebrush Steppe	IMB_Steppe	35.4	
Inter-Mountain Basins Mixed Salt Desert Scrub	IMB_Desert	2.3	

^aPVT lacks vegetation but was moderately used by bighorn sheep and so was included in analyses of sheep behavior. ^bPVT was not sampled but was combined with RMSM_Dry_SFW for modeling analyses based on similarity of habitat.

Predictor variable	Source
Palmer Drought Severity Index (PDSI); Precipitation (monthly total, ppt); Soil moisture (soil); Max temperature (°C, tmax)	Climatology Lab. 2019. TerraClimate. <http: terraclimate.html="" www.climatologylab.org="">. Accessed 20 July 2020.</http:>
Potential Vegetation Type (PVT)	LANDFIRE. 2008. Biophysical Settings Layer, LANDFIRE 1.1.0, U.S. Department of the Interior, Geological Survey. <http: landfire.cr.usgs.gov="" viewer=""></http:> . Accessed 11 Oct 2017.
Enhanced Vegetation Index (EVI)	NASA LP DAAC MOD13Q1 MODIS/Terra Vegetation Indices 16-DAY 13 Global 250m SIN Grid V005. NASA EOSDIS Land Processes DAAC, USGS Earth Resources Observation and Science (EROS) Center, Sioux Falls, South Dakota <https: lpdaac.usgs.gov="">. Accessed 6 Nov 2018.</https:>
Snowmelt date	National Operational Hydrologic Remote Sensing Center. 2004. Snow Data Assimilation System (SNODAS) Data Products at NSIDC, Version 1. Boulder, Colorado USA. NSIDC: National Snow and Ice Data Center. <https: 10.7265="" doi.org="" n5tb14tc="">. Accessed 27 Aug 2020.</https:>
Aspect (cosAspect, sinAspect); Elevation (m); Slope (degrees)	Inside Idaho, <http: insideidaho.org="" popular_data.html="">. Accessed 7 June 2020.</http:>

Table 3. Sources of candidate predictor variables for modeling forage biomass.

Table 4. Generalized additive models (GAMs) used to predict spatiotemporal variation in the nutritional landscape (i.e., biomass of forage) available to bighorn sheep in each of three population ranges in Idaho, USA. For each population range we fit separate models for predicting total biomass (i.e., total dry mass [kg/ha] of all plant species) versus accepted biomass (i.e., total dry mass [kg/ha] of plant species ostensibly consumed by sheep) of forage. Spatial and temporal smoothing terms were fit using cubic regression splines, and cross-validation was used to determine the optimal amount of smoothing for each term. Model selection procedures are described in detail in the Methods section.

Range	Best model	Adjusted R^2	Deviance explained	GCV ^a
Lost River	Total biomass ~ $s(JULIAN)^{b} + s(UTM_X) + PVT^{c} + sinAspect + Slope + Elevation + EVI^{2,d} + PDSI^{e}$	0.49	62.6%	125,240
	Accepted biomass ~ $s(JULIAN) + s(UTM_X, UTM_Y) + PVT + ln(Elevation) + EVI2 + PDSI2$	0.63	73.4%	90,971
East	Total biomass ~ sinAspect + $\ln(Slope) + \ln(ppt^{f})$	0.28	31.9%	284,990
Fork	Accepted biomass ~ $s(UTM_Y) + PVT + sinAspect + ln(Slope) + EVI2 + ln(ppt)$	0.51	60.2%	213,590
Owyhee	Total biomass ~ s(JULIAN) + cosAspect + EVI	0.49	57.9%	109,110
	Accepted biomass ~ s(JULIAN) + cosAspect + EVI	0.49	57.3%	110,460

^a Minimum generalized cross-validation score; ^b s() = smoothing term; ^c PVT = potential vegetation type; ^d EVI = enhanced vegetation index; ^e PDSI = Palmer drought severity index; ^f ppt = total precipitation; ^g soil = soil moisture.

Table 5. Regression coefficients and associated *SE*s and *P*-values for covariates included in the top generalized additive models (GAMs) for predicting spatiotemporal variation in the nutritional landscape (i.e., total or accepted biomass [kg/ha] of forage plants for bighorn sheep) in three population ranges in Idaho, USA, during summer (May-September), 2018–2019. Only *P*-values are shown for smoothing terms, which are denoted with an 's'. Potential vegetation type (PVT) was a categorical variable, and coefficients represent contrasts between each listed PVT and the reference PVT (IMB_Shrub). See Table 1 for PVT definitions. EVI = enhanced vegetation index.

Range	Model	Parameter	Estimate	SE	Р
Lost River	Total biomass	s(Julian)	-	-	0.041
		s(UTM_X,UTM_Y)	-	-	0.209
		PVT – IMB_Steppe	-183.052	158.739	0.255
		PVT – IMB_Desert	-247.195	197.390	0.217
		PVT – IMB_MtnSteppe	-220.035	243.104	0.371
		PVT – NRM_SWP	-648.605	314.320	0.045
		PVT-RMSM_Dry_SFW	-724.211	299.997	0.020
		Slope	-11.808	8.958	0.195
		Elevation	0.464	0.300	0.104
		EVI^2	8407.301	2467.238	0.001
	Accepted biomass	s(Julian)	-	-	0.036
		s(UTM_X,UTM_Y)	-	-	0.003
		Intercept	322.820	77.440	< 0.001
		EVI^2	8829.820	2155.80	< 0.001
East Fork	Total biomass	sinAspect	311.500	111.400	0.007
		ln(Slope)	-310.700	120.900	0.013
		ln(Total precipitation)	-339.500	113.300	0.004
	Accepted biomass	s(UTM_Y)	-	-	0.328
	-	PVT – CPS_Grassland	-77.850	249.010	0.756
		PVT – IMB_Steppe	32.200	261.240	0.903

		PVT – IMB_MtnSteppe	54.690	247.130	0.826
		$PVT - NRM_SWP$	-356.210	213.940	0.103
		PVT – RMSM_Dry_SFW	-583.340	222.720	0.012
		sinAspect	165.240	97.440	0.097
		ln(Slope)	-235.410	107.040	0.033
		EVI^2	9899.120	2907.180	0.001
		ln(Total precipitation)	-412.600	96.950	< 0.001
Owyhee	Total biomass	s(Julian)	-	-	0.648
		cosAspect	262.500	128.400	0.059
		EVI	7535.800	2393.100	0.007
	Accepted biomass	s(Julian)	-	-	0.613
		cosAspect	260.400	129.200	0.062
		EVI	7540.300	2407.900	0.007

Table 6. Competing models (Δ AICc < 2) for explaining variation in the probability of lamb survival during summer (May–September) as a function of available forage biomass (total or accepted; see Methods) and terrain ruggedness at locations used by GPS-collared bighorn sheep in three population ranges in Idaho, USA from 2016–2019. The sheep-year (i.e., data from one GPS-collared sheep in one year) was considered the unit of replication for the analysis. Models were fit using the known-fate modeling framework in Program MARK and are ranked according to Akaike's Information Criterion corrected for small sample size (AICc); model weights (w_i) are also shown. Variable definitions are as follows: Max_AB = Maximum accepted forage biomass; Mean_AB = Mean accepted forage biomass; Max_TB = Maximum total forage biomass; Mean_TB = Mean total forage biomass; Max_Ruggedness = Maximum terrain ruggedness; Mean_Ruggedness = Mean terrain ruggedness.

Range	Model	AICc	ΔAICc	Wi
Lost River	Max_TB	64.0	0.00	0.21
	Max_TB + Max_Ruggedness	64.3	0.32	0.18
	Max_TB + Mean_Ruggedness	64.9	0.88	0.14
	Max_Ruggedness	65.8	1.80	0.09
East Fork	Null	81.0	0.00	0.20
	Mean_AB	81.3	0.33	0.17
	Max_Ruggedness	82.7	1.69	0.09
	Mean_TB	82.7	1.71	0.09
	Max_TB	82.8	1.84	0.08
	Max_AB	82.8	1.86	0.08
	Max_AB + Mean_AB	82.9	1.97	0.08
Owyhee	Null	47.8	0.00	0.23
-	Max_TB	49.3	1.45	0.11
	Mean_AB	49.3	1.50	0.11
	Mean_TB	49.4	1.51	0.11
	Max_AB	49.4	1.52	0.11

Table 7. Parameter estimates and associated *SEs* and 95% CIs for covariates included in one of four top models (see Table 5) for predicting bighorn sheep lamb survival as a function of forage availability and terrain ruggedness at locations used by sheep in the Lost River population range in Idaho, USA during summer (May-September), 2017–2019. Max_TB = Maximum total forage biomass; Max_Ruggedness = Maximum terrain ruggedness. The top used-location model for the East Fork and Owyhee population ranges was the null model, so no parameter estimates are presented for those ranges.

Range	Parameter	Estimate	SE	CI lower	CI upper
Lost River	Max_Ruggedness	-0.66	0.56	-1.76	0.45
	Max_TB	1.31	0.76	-0.18	2.80

Table 8. Competing models ($\Delta AICc < 2$) for explaining variation in the probability of lamb survival during summer (May–September) as a function of available forage biomass (total or accepted; see Methods) and terrain ruggedness at random locations generated within the home ranges of GPS-collared bighorn sheep in three population ranges in Idaho, USA from 2016– 2019. The sheep-year (i.e., data from one GPS-collared sheep in one year) was considered the unit of replication for the analysis. Models were fit using the known-fate modeling framework in Program MARK and are ranked according to Akaike's Information Criterion corrected for small sample size (AICc); model weights (w_i) are also shown. Variable definitions are as follows: Max_AB = Maximum accepted forage biomass; Mean_AB = Mean accepted forage biomass; CV_AB = Coefficient of variation for accepted forage biomass; Max_TB = Maximum total forage biomass; Mean_TB = Mean total forage biomass; CV_TB = Coefficient of variation for total forage biomass; Max_Ruggedness = Maximum terrain ruggedness; Mean_Ruggedness = Mean terrain ruggedness; CV_Ruggedness = Coefficient of variation for terrain ruggedness.

Range	Model	AICc	ΔAICc	Wi
Lost River	Mean_TB + CV_Ruggedness	56.3	0.00	0.65
East Fork	Mean_TB	76.6	0.00	0.25
	Mean_TB + CV_Ruggedness	77.1	0.54	0.19
Owyhee	Null	47.8	0.00	0.10
	CV_Ruggedness	48.3	0.41	0.08
	CV_AB	48.3	0.48	0.08
	CV_TB	48.4	0.52	0.08
	Mean_AB	49.1	1.23	0.05
	Mean_TB	49.1	1.24	0.05
	Max_TB	49.6	1.77	0.04
	Max_AB	49.6	1.78	0.04
	Mean_Ruggedness	49.7	1.88	0.04
	Max_Ruggedness	49.8	1.92	0.04

Table 9. Parameter estimates and associated *SEs* and 95% CIs for covariates included in top models (see Table 7) for predicting bighorn sheep lamb survival as a function of forage availability and terrain ruggedness at random locations within the home ranges of bighorn sheep in the Lost River and East Fork population ranges in Idaho, USA during summer (May–September), 2016–2019. Mean_TB = Mean total forage biomass; CV_Ruggedness = Coefficient of variation for terrain ruggedness. The top random-location model for the Owyhee population range was the null model, so no parameter estimates are presented for that range.

Range	Parameter	Estimate	SE	CI Lower	CI Upper
Lost River	Mean_TB	2.36	0.82	0.75	3.97
	CV_Ruggedness	-1.80	0.63	-3.02	-0.57
East Fork	Mean_TB	-0.87	0.32	-1.57	-0.18
	CV_Ruggedness	0.41	0.32	-0.23	1.04

Figures





Figure 2. Mean (±90% CI) accepted and total biomass of forage (kg/ha) in potential vegetation types (PVTs) that comprised >0.2% of three bighorn sheep population ranges in Idaho, USA during summer (May–September), 2018–2019. See Table 2 for PVT definitions.



Potential vegetation type (PVT)

Figure 3. Proportion of randomly sampled locations in each of three bighorn sheep population ranges in Idaho, USA during summer (May–September) that fell into each of four quartiles of predicted forage biomass (accepted and total biomass) based on generalized additive models of the nutritional landscape (see Table 2; High >= 705 kg/ha, Mid-High = 423–704 kg/ha, Mid-Low = 246–422 kg/ha, Low <= 245 kg/ha).



Figure 4. Boxplots of predicted (based on generalized additive models of the nutritional landscape; see Table 2) total and accepted forage biomass at random locations versus locations used by GPS-collared female bighorn sheep in each of three population ranges in Idaho, USA. GPS location data were collected during summer (May–September), 2016–2019.



Figure 5. Estimated survival of bighorn sheep lambs in each of three population ranges in Idaho, USA during summer (May– September), 2016–2019. Panel A shows raw results for each combination of population range and year where >10 lambs were monitored. Panel B shows two sets of range-specific survival estimates (±95% CI) derived from the known-fate model in Program MARK (see Methods): 1) estimates derived from the best model of lamb survival where individual covariates were extracted from locations used by GPS-collared sheep; and 2) estimates derived from the best model of lamb survival where individual covariates were extracted from random locations within the home range of each GPS-collared sheep.



Figure 6. Relationship between summer (May–September) lamb survival and spring body condition (quantified by % ingesta-free body fat) of female bighorn sheep across three bighorn sheep population ranges in Idaho, USA during 2016–2018.



Appendices

Appendix A. List of plant species sampled during our study (2017–2019) and the associated level of hypothesized selection by bighorn sheep. Level of selection by bighorn sheep was based on published and unpublished data provided by L. Shipley and T. Stephenson. Taxonomy: The PLANTS Database, USDA, NRCS, 2017 (http://plants.usda.gov, accessed 6/1/2017).

Plant code	Family	Scientific name	Common name	Level of
	j			selection
ABLA	Pinaceae	Abies lasiocarpa	Subalpine fir	Avoided
ACGL	Aceraceae	Acer glabrum	Rocky Mountain maple	Accepted
ACMI2	Asteraceae	Achillea millefolium	Common yarrow	Accepted
ACMIO	Asteraceae	Achillea millefolium var. occidentalis	Western yarrow	Accepted
ACHY	Poaceae	Achnatherum hymenoides	Indian ricegrass	Accepted
ACLEL	Poaceae	Achnatherum lemmonii var lemmonii	Lemmon's needlegrass	Accepted
ACLE9	Poaceae	Achnatherum lettermanii	Letterman's needlegrass	Accepted
ACNEN2	Poaceae	Achnatherum nelsonii	Columbia needlegrass	Accepted
ACTH7	Poaceae	Achnatherum thurberianum	Thurber's needlegrass	Accepted
AGAU2	Asteraceae	Agoseris aurantiaca	Orange agoseris	Avoided
AGGL	Asteraceae	Agoseris glauca	Pale agoseris	Avoided
AGGLL	Asteraceae	Agoseris glauca var. laciniata	False agoseris	Avoided
AGHE2	Asteraceae	Agoseris heterophylla	Annual agoseris	Avoided
AGCR	Poaceae	Agropyron cristatum	Crested wheatgrass	Accepted
AGHU	Poaceae	Agrostis humilis	Alpine bentgrass	Accepted
ALAC4	Amaryllidaceae	Allium acuminatum	Tapertip onion	Avoided
ALBR	Liliaceae	Allium brandegeei	Brandegee's onion	Avoided
ALBR2	Amaryllidaceae	Allium brevistylum	Shortstyle onion	Avoided
ALTO	Liliaceae	Allium tolmiei	Tolmie's onion	Avoided
ALIN2	Betulaceae	Alnus incana	Grey alder	Accepted
ALAL3	Brassicaceae	Alyssum alyssoides	Yellow alyssum	Accepted
ALDE	Brassicaceae	Alyssum desertorum	Desert madwort	Accepted

AMAL2	Rosaceae	Amelanchier alnifolia	Saskatoon serviceberry	Accepted
ANME	Boraginaceae	Amsinckia menziesii	Common fiddleneck	Accepted
ANSE4	Primulaceae	Androsace septentrionalis	Pygmy-flower rock-jasmine	Accepted
ANDR	Ranunculaceae	Anemone drummondii	Drummond's anemone	Avoided
ANMU	Ranunculaceae	Anemone multifida	Pacific anemone	Avoided
ANAL4	Asteraceae	Antennaria alpina	Alpine pussytoes	Avoided
ANDI2	Asteraceae	Antennaria dimorpha	Low pussytoes	Avoided
ANLA3	Asteraceae	Antennaria lanata	Woolly pussytoes	Avoided
ANME2	Asteraceae	Antennaria media	Rocky Mountain pussytoes	Avoided
ANMI3	Asteraceae	Antennaria microphylla	Tiny-leaved pussytoes	Avoided
ANRO2	Asteraceae	Antennaria rosea	Rosy pussytoes	Avoided
ANUM	Asteraceae	Antennaria umbrinella	Umber pussytoes	Avoided
AQCO	Ranunculaceae	Aquilegia coerulea	Colorado blue columbine	Accepted
AQFL	Ranunculaceae	Aquilegia flavescens	Yellow mountain columbine	Accepted
AQFO	Ranunculaceae	Aquilegia formosa	Western columbine	Accepted
ARCO	Brassicaceae	Arabis cobrensis	Sagebrush rockcress	Avoided
ARHI	Brassicaceae	Arabis hirsuta	Hairy rockcress	Avoided
ARHO2	Brassicaceae	Arabis holboellii	Holboell's rockcress	Avoided
ARLEL	Brassicaceae	Arabis lemmonii var. lemmonii	Lemmon's rockcress	Avoided
ARWIS	Brassicaceae	Arabis williamsii var. saximontana	William's rockcress	Avoided
ARDI2	Brassicaceae	Arabis xdivaricarpa	Spreadingpod rockcress	Avoided
ARUV	Ericaceae	Arctostaphylos uva-ursi	Kinnikinnick	Accepted
ARAC2	Caryophyllaceae	Arenaria aculeata	Prickly sandwort	Accepted
ARCO5	Caryophyllaceae	Arenaria congesta	Ballhead sandwort	Accepted
ARCO9	Asteraceae	Arnica cordifolia	Arnica cordifolia	Accepted
ARLA8	Asteraceae	Arnica latifolia	Broadleaf arnica	Accepted
ARLO6	Asteraceae	Arnica longifolia	Spearleaf arnica	Accepted
ARRY	Asteraceae	Arnica rydbergii	Rydberg's arnica	Accepted
ARSO2	Asteraceae	Arnica sororia	Twin arnica	Accepted
ARARA	Asteraceae	Artemisia arbuscula ssp. arbuscula	Low sage	Accepted

ARARL	Asteraceae	Artemisia arbuscula ssp. longiloba	Early sage	Accepted
ARART	Asteraceae	Artemisia arbuscula ssp. thermopola	Little sagebrush	Accepted
ARCA12	Asteraceae	Artemisia campestris	Field sagewort	Avoided
ARFR4	Asteraceae	Artemisia frigida	Prairie sagewort	Accepted
ARLU	Asteraceae	Artemisia ludoviciana	White sagebrush	Accepted
ARMI4	Asteraceae	Artemisia michauxiana	Lemon sagewort	Avoided
ARMI9	Asteraceae	Artemisia minima	Spreading sneezeweed	Avoided
ARTRT	Asteraceae	Artemisia tridentata ssp. tridentata	Basin big sagebrush	Accepted
ARTRV	Asteraceae	Artemisia tridentata ssp. vaseyana	Mountain big sagebrush	Accepted
ARTRW8	Asteraceae	Artemisia tridentata ssp. wyomingensis	Wyoming big sagebrush	Accepted
ARTR4	Asteraceae	Artemisia tripartita	Threetip sagebrush	Accepted
ASAL7	Fabaceae	Astragalus alpinus	Alpine milkvetch	Accepted
ASAT2	Fabaceae	Astragalus atropubescens	Hangingpod milkvetch	Accepted
ASAU4	Fabaceae	Astragalus australis	Indian milkvetch	Accepted
ASBE3	Fabaceae	Astragalus beckwithii	Beckwith's milkvetch	Accepted
ASCO11	Fabaceae	Astragalus conjunctus	Idaho milkvetch	Accepted
ASCU4	Fabaceae	Astragalus curvicarpus	Curvepod milkvetch	Accepted
ASDO	Fabaceae	Astragalus douglasii	Douglas' milkvetch	Accepted
ASFI	Fabaceae	Astragalus filipes	Basalt milkvetch	Accepted
ASKET	Fabaceae	Astragalus kentrophyta var. tegetarius	Mat milkvetch	Accepted
ASLE8	Fabaceae	Astragalus lentiginosus	Freckled milkvetch	Accepted
ASOB4	Fabaceae	Astragalus obscurus	Arcane milkvetch	Accepted
ASPU9	Fabaceae	Astragalus purshii	Woollypod milkvetch	Accepted
ASWHW	Fabaceae	Astragalus whitneyi var. whitneyi	Balloonpod milkvetch	Accepted
ATCO	Chenopodiaceae	Atriplex confertifolia	Shadscale saltbush	Accepted
BAHO	Asteraceae	Balsamorhiza hookeri	Hooker's balsamroot	Accepted
BASA3	Asteraceae	Balsamorhiza sagittata	Arrowleaf balsamroot	Accepted
BEPA	Betulaceae	Betula papyrifera	Paper birch	Accepted
BLSC	Asteraceae	Blepharipappus scaber	Rough eyelashweed	Accepted

BROBO	Asteraceae	Brickellia oblongifolia	Narrowleaf brickellbush	Accepted
BRAR5	Poaceae	Bromus arvensis	Field brome	Accepted
BRER3	Poaceae	Bromus erectus	Meadow brome	Accepted
BRTE	Poaceae	Bromus tectorum	Cheatgrass	Accepted
BUAM2	Apiaceae	Bupleurum americanum	American thorowax	Accepted
CAPU	Poaceae	Calamagrostis purpurascens	Purple reedgrass	Accepted
CARU	Poaceae	Calamagrostis rubescens	Pinegrass	Accepted
CANU3	Liliaceae	Calochortus nuttallii	Sego lily	Accepted
CALE4	Ranunculaceae	Caltha leptosepala	White marsh marigold	Accepted
CAMI2	Brassicaceae	Camelina microcarpa	Little false flax	Accepted
CASA2	Brassicaceae	Camelina sativa	False flax	Accepted
CAAB2	Cyperaceae	Carex abrupta	Abrupt-beaked sedge	Accepted
CADO2	Cyperaceae	Carex douglasii	Douglas' sedge	Accepted
CAEL3	Cyperaceae	Carex elynoides	Blackroot sedge	Accepted
CAFI	Cyperaceae	Carex filifolia	Threadleaf sedge	Accepted
CAGE2	Cyperaceae	Carex geyeri	Geyer's sedge	Accepted
CAHO5	Cyperaceae	Carex hoodii	Hood's sedge	Accepted
CALE9	Cyperaceae	Carex leporinella	Hare sedge	Accepted
CAMI7	Cyperaceae	Carex microptera	Small-wing sedge	Accepted
CAMU6	Cyperaceae	Carex multicostata	Manyrib sedge	Accepted
CAPA31	Cyperaceae	Carex paysonis	Payson's sedge	Accepted
CAPE42	Cyperaceae	Carex pellita	Woolly sedge	Accepted
CAPH2	Cyperaceae	Carex phaeocephala	Dunhead sedge	Accepted
CARO5	Cyperaceae	Carex rossii	Ross' sedge	Accepted
CASI2	Cyperaceae	Carex simulata	Analogue sedge	Accepted
CAVA3	Cyperaceae	Carex vallicola	Valley sedge	Accepted
CAME7	Ericaceae	Cassiope mertensiana	Western moss heather	Accepted
CAAN7	Scrophulariaceae	Castilleja angustifolia	Northwestern Indian paintbrush	Accepted
CAAP4	Scrophulariaceae	Castilleja applegatei	Wavyleaf Indian paintbrush	Accepted
CACO36	Scrophulariaceae	Castilleja covilleana	Coville's Indian paintbrush	Accepted

CAFL7	Scrophulariaceae	Castilleja flava	Yellow Indian paintbrush	Accepted
CAMIM6	Scrophulariaceae	Castilleja minor	Lesser Indian paintbrush	Accepted
CAOC4	Scrophulariaceae	Castilleja occidentalis	Western Indian paintbrush	Accepted
CAPA25	Scrophulariaceae	Castilleja pallescens	Pale Indian paintbrush	Accepted
CAPI3	Scrophulariaceae	Castilleja pilosa	Parrothead Indian paintbrush	Accepted
CAPIL	Scrophulariaceae	Castilleja pilosa ssp. longispica	Longspike Indian paintbrush	Accepted
CEST8	Asteraceae	Centaurea stoebe	Spotted knapweed	Accepted
CEAR4	Caryophyllaceae	Cerastium arvense	Field chickweed	Accepted
CENU2	Caryophyllaceae	Cerastium nutans	Nodding chickweed	Accepted
CELE3	Rosaceae	Cercocarpus ledifolius	Curl-leaf mountain mahogany	Accepted
CHDO	Asteraceae	Chaenactis douglasii	Douglas' dustymaiden	Accepted
CHDOA	Asteraceae	Chaenactis douglasii var. achilleifolia	Douglas' dustymaiden	Accepted
CHAN9	Onagraceae	Chamerion angustifolium	Fireweed	Accepted
CHLA13	Onagraceae	Chamerion latifolium	Broadleaf fireweed	Accepted
CHLE4	Chenopodiaceae	Chenopodium leptophyllum	Narrowleaved goosefoot	Accepted
CHUM	Pyrolaceae	Chimaphila umbellata	Prince's pine	Accepted
CHTW	Scrophulariaceae	Chionophila tweedyi	Tweedy snowlover	Accepted
CHTE2	Brassicaceae	Chorispora tenella	Purple mustard	Accepted
CHHU2	Asteraceae	Chrysothamnus humilis	Truckee rabbitbrush	Accepted
CHVI8	Asteraceae	Chrysothamnus viscidiflorus	Yellow rabbitbrush	Accepted
CHVIV2	Asteraceae	Chrysothamnus viscidiflorus ssp. viscidiflorus	Sticky leaved rabbitbrush	Accepted
CIAR4	Asteraceae	Cirsium arvense	Field thistle	Accepted
CICA6	Asteraceae	Cirsium canovirens	Graygreen thistle	Avoided
CICY	Asteraceae	Cirsium cymosum	Peregrine thistle	Avoided
CIFO	Asteraceae	Cirsium foliosum	Elk thistle	Accepted
CISC2	Asteraceae	Cirsium scariosum	Meadow thistle	Avoided
CIUN	Asteraceae	Cirsium undulatum	Wavyleaf thistle	Accepted
CIVU	Asteraceae	Cirsium vulgare	Common thistle	Accepted
CIUMU	Portulacaceae	Cistanthe umbellata var. umbellata	Mt. Hood pussypaws	Accepted
COPA3	Scrophulariaceae	Collinsia parviflora	Small-flowered blue-eyed Mary	Accepted

Polemoniaceae	Collomia grandiflora	Grand collomia	Accepted
Polemoniaceae	Collomia linearis	Tiny trumpet	Accepted
Polemoniaceae	Collomia tinctoria	Staining collomia	Accepted
Cornaceae	Cornus sericea	Red twig dogwood	Accepted
Asteraceae	Crepis acuminata	Tapertip hawksbeard	Accepted
Asteraceae	Crepis atribarba	Slender hawksbeard	Accepted
Asteraceae	Crepis modocensis	Modoc hawksbeard	Accepted
Asteraceae	Crepis occidentalis	Largeflower hawksbeard	Accepted
Boraginaceae	Cryptantha interrupta	Elko cryptantha	Accepted
Boraginaceae	Cryptantha sobolifera	Waterton Lakes cryptantha	Accepted
Boraginaceae	Cryptantha torreyana	Torrey's cryptantha	Accepted
Apiaceae	Cymopterus corrugatus	Wrinklewing spring-parsley	Avoided
Apiaceae	Cymopterus nivalis	Snow spring-parsley	Avoided
Boraginaceae	Cynoglossum officinale	Gypsy flower	Avoided
Poaceae	Danthonia intermedia	Mountain wild-oat grass	Accepted
Poaceae	Danthonia unispicata	Onespike danthonia	Accepted
Rosaceae	Dasiphora fruticosa	Shrubby cinquefoil	Accepted
Ranunculaceae	Delphinium bicolor	Little larkspur	Avoided
Ranunculaceae	Delphinium depauperatum	Slim larkspur	Avoided
Poaceae	Deschampsia cespitosa	Tuffed hair grass	Accepted
Brassicaceae	Descurainia incisa	Mountain tansy mustard	Avoided
Brassicaceae	Descurainia pinnata	Western tansy mustard	Accepted
Brassicaceae	Descurainia pinnata ssp. nelsonii	Nelson's tansy mustard	Accepted
Brassicaceae	Descurainia sophia	Flixweed	Accepted
Fumariaceae	Dicentra uniflora	Steer's head	Avoided
Poaceae	Distichlis spicata	Saltgrass	Accepted
Primulaceae	Dodecatheon jeffreyi	Sierra shooting star	Accepted
Primulaceae	Dodecatheon pulchellum	Darkthroat shooting star	Accepted
Primulaceae	Douglasia idahoensis	Idaho dwarf primrose	Accepted
Brassicaceae	Draba crassifolia	Snowbed draba	Accepted
	Polemoniaceae Polemoniaceae Polemoniaceae Cornaceae Asteraceae Asteraceae Asteraceae Boraginaceae Boraginaceae Boraginaceae Boraginaceae Apiaceae Boraginaceae Poaceae Poaceae Ranunculaceae Ranunculaceae Brassicaceae Brassicaceae Brassicaceae Brassicaceae Brassicaceae Poaceae Poaceae Poaceae Poaceae Poaceae Poaceae Poaceae Brassicaceae Brassicaceae Primulaceae Primulaceae Primulaceae	PolemoniaceaeCollomia grandifloraPolemoniaceaeCollomia linearisPolemoniaceaeCollomia tinctoriaCornaceaeCornus sericeaAsteraceaeCrepis acuminataAsteraceaeCrepis atribarbaAsteraceaeCrepis modocensisAsteraceaeCrepis occidentalisBoraginaceaeCryptantha interruptaBoraginaceaeCryptantha soboliferaBoraginaceaeCryptantha torreyanaApiaceaeCymopterus nivalisBoraginaceaeCynoglossum officinalePoaceaeDanthonia intermediaPoaceaeDanthonia unispicataRosaceaeDelphinium bicolorRanunculaceaeDelphinium depauperatumPoaceaeDescurainia pinnataBrassicaceaeDescurainia sophiaBrassicaceaeDescurainia pinnataBrassicaceaeDicentra unifloraPoaceaeDicentra unifloraPoaceaeDescurainia pinnataBrassicaceaeDescurainia sophiaFrimulaceaeDicentra unifloraPoaceaeDistichlis spicataBrassicaceaeDodecatheon pulchellumPrimulaceaeDodecatheon pulchellum	PolemoniaceaeCollomia grandifloraGrand collomiaPolemoniaceaeCollomia linearisTiny trumpetPolemoniaceaeCollomia tinctoriaStaining collomiaCornaceaeCornus sericeaRed twig dogwoodAsteraceaeCrepis acuminataTapertip hawksbeardAsteraceaeCrepis acuminataSlender hawksbeardAsteraceaeCrepis occidentalisLargeflower hawksbeardBoraginaceaeCryptantha interruptaElko cryptanthaBoraginaceaeCryptantha interruptaElko cryptanthaBoraginaceaeCryptantha soboliferaWaterton Lakes cryptanthaBoraginaceaeCryptantha torreyanaTorrey's cryptanthaApiaceaeCympterus corrugatusWrinklewing spring-parsleyApiaceaeCymopterus nivalisSnow spring-parsleyBoraginaceaeDanthonia intermediaMountain wild-oat grassPoaceaeDanthonia unispicataOnespike danthoniaRosaceaeDesiphora fruticosaShrubby cinquefoilRanunculaceaeDelphinium bicolorLittle larkspurPoaceaeDescurainia pinnataWestern tansy mustardBrassicaceaeDescurainia pinnata ssp. nelsoniiNelson's tansy mustardBrassicaceaeDiscuraina spinataSter's headPoaceaeDistichlis spicataSter's headPoaceaeDescurainia pinnata ssp. nelsoniiNelson's tansy mustardBrassicaceaeDescurainia spinataSter's headPoaceaeDistichlis spicataSter's headPoaceaeDistichl

DRLO	Brassicaceae	Draba lonchocarpa	Lacepod draba	Accepted	
DROL	Brassicaceae	Draba oligosperma	Fewseed draba	Accepted	
DRSP2	Brassicaceae	Draba sphaerocarpa	Globe-fruit whitlow grass	Accepted	
DRTR3	Brassicaceae	Draba trichocarpa	Stanley creek draba	Accepted	
DRVE2	Brassicaceae	Draba verna	Common whitlow grass	Accepted	
ELPA5	Cyperaceae	Eleocharis parvula	Common hairgrass	Accepted	
ELAL5	Poaceae	Elymus alaskanus	Alaska wild rye	Accepted	
ELEL5	Poaceae	Elymus elymoides	Squirreltail	Accepted	
EPAN4	Onagraceae	Epilobium anagallidifolium	Pimpernel willowherb	Accepted	
EPBR3	Onagraceae	Epilobium brachycarpum	Tall annual fireweed	Accepted	
EPLA3	Onagraceae	Epilobium lactiflorum	Milkflower willowherb	Accepted	
EQAR	Equisetaceae	Equisetum arvense	Field horsetail	Accepted	
EQLA	Equisetaceae	Equisetum laevigatum	Smooth horsetail	Accepted	
ERSP3	Polemoniaceae	Eriastrum sparsiflorum	Great basin woollystar	Accepted	
ERNA7	Asteraceae	Ericameria nana	Dwarf goldenbush	Accepted	
ERNA10	Asteraceae	Ericameria nauseosa	Rubber rabbitbrush	Accepted	
ERSU13	Asteraceae	Ericameria suffruticosa	Singlehead goldenbush	Accepted	
ERAP	Asteraceae	Erigeron aphanactis	Rayless shaggy fleabane	Accepted	
ERAS	Asteraceae	Erigeron asperugineus	Idaho fleabane	Accepted	
ERBL	Asteraceae	Erigeron bloomeri	Scabland fleabane	Accepted	
ERCO4	Asteraceae	Erigeron compositus	Cutleaf daisy	Accepted	
ERCO5	Asteraceae	Erigeron corymbosus	Longleaf fleabane	Accepted	
ERFI2	Asteraceae	Erigeron filifolius	Threadleaf fleabane	Accepted	
ERLA14	Asteraceae	Erigeron latus	Broad fleabane	Accepted	
ERLE6	Asteraceae	Erigeron leiomerus	Rockslide fleabane	Accepted	
ERLI	Asteraceae	Erigeron linearis	Desert yellow fleabane	Accepted	
ERPE3	Asteraceae	Erigeron peregrinus	Subalpine fleabane	Accepted	
ERPU2	Asteraceae	Erigeron pumilus	Shaggy fleabane	Accepted	
ERCA8	Polygonaceae	Eriogonum caespitosum	Matted buckwheat	Accepted	
ERFL4	Polygonaceae	Eriogonum flavum	Alpine golden buckwheat	Accepted	
ERHE2	Polygonaceae	Eriogonum heracleoides	Parsnip-flower buckwheat	Accepted	
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ERME6	Polygonaceae	Eriogonum meledonum	Bridle buckwheat	Accepted	
ERMI4	Polygonaceae	Eriogonum microthecum	Great basin buckwheat	Accepted	
EROV	Polygonaceae	Eriogonum ovalifolium	Cushion buckwheat	Accepted	
ERST4	Polygonaceae	Eriogonum strictum	Blue mountain buckwheat	Accepted	
ERUM	Polygonaceae	Eriogonum umbellatum	Sulphur-flower buckwheat	Accepted	
ERLA6	Asteraceae	Eriophyllum lanatum	Common woolly sunflower	Accepted	
ERNA	Boraginaceae	Eritrichium nanum	Arctic alpine forget-me-not	Accepted	
ERCA14	Brassicaceae	Erysimum capitatum	Sand-dune wallflower	Accepted	
ERIN7	Brassicaceae	Erysimum inconspicuum	Shy wallflower	Accepted	
FEBRB	Poaceae	Festuca brachyphylla ssp. brachyphylla	Alpine fescue	Accepted	
FECA4	Poaceae	Festuca campestris	Rough fescue	Avoided	
FEID	Poaceae	Festuca idahoensis	Idaho fescue	Accepted	
FEOC	Poaceae	Festuca occidentalis	Western fescue	Avoided	
FRVI	Rosaceae	Fragaria virginiana	Wild strawberry	Accepted	
FRSP	Gentianaceae	Frasera speciosa	Elkweed	Accepted	
FRAT	Liliaceae	Fritillaria atropurpurea	Spotted fritillary	Accepted	
GAMU2	Rubiaceae	Galium multiflorum	Shrubby bedstraw	Accepted	
GASE2	Rubiaceae	Galium serpenticum	Northern bedstraw	Accepted	
GAHU	Ericaceae	Gaultheria humifusa	Alpine spicy wintergreen	Accepted	
GADI2	Onagraceae	Gayophytum diffusum	Spreading groundsmoke	Accepted	
GARA	Onagraceae	Gayophytum racemosum	Blackfooted groundsmoke	Accepted	
GARA2	Onagraceae	Gayophytum ramosissimum	Pinyon groundsmoke	Accepted	
GEAF	Gentianaceae	Gentiana affinis	Pleated gentian	Avoided	
GECA	Gentianaceae	Gentiana calycosa	Mountain bog gentian	Avoided	
GEMA4	Rosaceae	Geum macrophyllum	Large-leaf avens	Accepted	
GERO2	Rosaceae	Geum rossii	Alpine avens	Accepted	
GETR	Rosaceae	Geum triflorum	Prairie smoke	Accepted	
GIIN2	Polemoniaceae	Gilia inconspicua	Shy gilia	Accepted	
GLST	Poaceae	Glyceria striata	Fowl mannagrass	Accepted	60

GRSP	Chenopodiaceae	Grayia spinosa	Spiny hopsage	Accepted
HADE	Boraginaceae	Hackelia deflexa	Nodding stickseed	Avoided
HAMI	Boraginaceae	Hackelia micrantha	Jessica stickseed	Avoided
HAPA	Boraginaceae	Hackelia patens	Spotted stickseed	Avoided
HAGL	Chenopodiaceae	Halogeton glomeratus	Saltlover	Avoided
HECO26	Poaceae	Hesperostipa comata	Needle and thread grass	Accepted
HECOI	Poaceae	Hesperostipa comata ssp. intermedia	Intermediate needle and thread grass	Accepted
HECY2	Saxifragaceae	Heuchera cylindrica	Roundleaf alumroot	Accepted
HEPA11	Saxifragaceae	Heuchera parvifolia	Little-leaf alumroot	Accepted
HICY	Asteraceae	Hieracium cynoglossoides	Hound's tongue hawksweed	Accepted
HOJU	Poaceae	Hordeum jubatum	Foxtail barley	Accepted
HYCA4	Hydrophyllaceae	Hydrophyllum capitatum	Ballhead waterleaf	Accepted
IOAL	Asteraceae	Ionactis alpina	Lava aster	Accepted
IPAG	Polemoniaceae	Ipomopsis aggregata	Scarlet gilia	Accepted
IPCO5	Polemoniaceae	Ipomopsis congesta	Ballhead ipomopsis	Accepted
IPCOP	Polemoniaceae	Ipomopsis congesta ssp. palmifrons	Ballhead gilia	Accepted
IPSPO2	Polemoniaceae	Ipomopsis spicata ssp orchidaceae	Orchid ipomopsis	Accepted
IVGO	Rosaceae	Ivesia gordonii	Gordon's ivesia	Accepted
JUAR2	Juncaceae	Juncus arcticus	Arctic rush	Avoided
JUBA	Juncaceae	Juncus balticus	Baltic rush	Accepted
JUDR	Juncaceae	Juncus drummondii	Drummond's rush	Accepted
JUEN	Juncaceae	Juncus ensifolius	Swordleaf rush	Avoided
JUME3	Juncaceae	Juncus mertensianus	Mertens' rush	Avoided
JUNE	Juncaceae	Juncus nevadensis	Sierra rush	Avoided
JUCO6	Cupressaceae	Juniperus communis	Common juniper	Avoided
JUOC	Cupressaceae	Juniperus occidentalis	Western juniper	Avoided
JUSC2	Cupressaceae	Juniperus scopulorum	Rocky mountain juniper	Avoided
KOMA	Poaceae	Koeleria macrantha	Prairie junegrass	Accepted
KRLA2	Chenopodiaceae	Krascheninnikovia lanata	Winterfat	Accepted
LASE	Asteraceae	Lactuca serriola	Prickly lettuce	Accepted

LARA	Asteraceae	Lagophylla ramosissima	Common hareleaf	Accepted
LAOC3	Boraginaceae	Lappula occidentalis	Flatspine stickseed	Accepted
LAGL5	Asteraceae	Layia glandulosa	White daisy tidytips	Accepted
LEGL	Ericaceae	Ledum glandulosum	Western Labrador tea	Avoided
LEPE2	Brassicaceae	Lepidium perfoliatum	Clasping pepperweed	Accepted
LEVI3	Brassicaceae	Lepidium virginicum	Least pepperwort	Accepted
LENU8	Polemoniaceae	Leptosiphon nuttallii	Nuttall's linanthus	Accepted
LESE17	Polemoniaceae	Leptosiphon septentrionalis	Northern linanthus	Accepted
LEKI2	Poaceae	Leucopoa kingii	Spike fescue	Accepted
LEPY2	Portulacaceae	Lewisia pygmaea	Pygmy bitterroot	Accepted
LERE7	Portulacaceae	Lewisia rediviva	Bitterroot	Accepted
LECI4	Poaceae	Leymus cinereus	Basin wildrye	Accepted
LIGR	Apiaceae	Ligusticum grayi	Gray's licorice-root	Accepted
LIPU11	Polemoniaceae	Linanthus pungens	Granite prickly phlox	Accepted
LILE3	Linaceae	Linum lewisii	Wild blue flax	Accepted
LIPA5	Saxifragaceae	Lithophragma parviflorum	Smallflower woodland star	Accepted
LIRU4	Boraginaceae	Lithospermum ruderale	Western stoneseed	Accepted
LOCO4	Apiaceae	Lomatium cous	Cous biscuitroot	Accepted
LOGR	Apiaceae	Lomatium grayi	Gray's biscuitroot	Accepted
LOMA3	Apiaceae	Lomatium macrocarpum	Bigseed biscuitroot	Accepted
LONU2	Apiaceae	Lomatium nudicaule	Bare-stem biscuitroot	Accepted
LOTR2	Apiaceae	Lomatium triternatum	Nineleaf biscuitroot	Accepted
LOIN5	Caprifoliaceae	Lonicera involucrata	Twinberry honeysuckle	Accepted
LUAR3	Fabaceae	Lupinus argenteus	Silvery lupine	Accepted
LULE2	Fabaceae	Lupinus lepidus	Prairie lupine	Accepted
LUSE2	Fabaceae	Lupinus sellulus	Donner lake lupine	Accepted
LUSE4	Fabaceae	Lupinus sericeus	Silky lupine	Accepted
LUWY	Fabaceae	Lupinus wyethii	Wyeth's lupine	Accepted
LUAR5	Juncaceae	Luzula arcuata	Curved wood rush	Avoided
LUSP4	Juncaceae	Luzula spicata	Spiked wood rush	Avoided

MACA2	Asteraceae	Machaeranthera canescens
MAGR3	Asteraceae	Madia gracilis
MAST4	Liliaceae	Maianthemum stellatum
MADI6	Asteraceae	Matricaria discoidea
MESP	Poaceae	Melica spectabilis
MEOF	Fabaceae	Melilotus officinalis
MEAR4	Lamiaceae	Mentha arvensis
MEAL6	Loasaceae	Mentzelia albicaulis
MEDI	Loasaceae	Mentzelia dispersa
MECA6	Boraginaceae	Mertensia campanulata
MEOB	Boraginaceae	Mertensia oblongifolia
MIGR	Polemoniaceae	Microsteris gracilis
MINU4	Caryophyllaceae	Minuartia nuttallii
MIOB2	Caryophyllaceae	Minuartia obtusiloba
MYST2	Boraginaceae	Myosotis stricta
NABR	Polemoniaceae	Navarretia breweri
NALE	Polemoniaceae	Navarretia leucocephala
NEST5	Asteraceae	Nestotus stenophyllus
OECA10	Onagraceae	Oenothera caespitosa
OPPO	Cactaceae	Opuntia polyacantha
ORAL4	Asteraceae	Oreostemma alpigenum
ORFA	Orobanchaceae	Orobanche fasciculata
PIEX4	Poaceae	Oryzopsis exigua
OSCH	Apiaceae	Osmorhiza chilensis
OXSE	Fabaceae	Oxytropis sericea
PACA15	Asteraceae	Packera cana
PADI11	Asteraceae	Packera dimorphophylla
PAST10	Asteraceae	Packera streptanthifolia
PASU40	Asteraceae	Packera subnuda
PASM	Poaceae	Pascopyrum smithii

Hoary tansyaster	Accepted
Grassy tarweed	Avoided
Starry false lily of the valley	Accepted
Wild chamomile	Avoided
Purple oniongrass	Avoided
Sweet clover	Accepted
Wild mint	Accepted
Whitestem blazingstar	Accepted
Bushy blazingstar	Accepted
Idaho bluebells	Avoided
Oblongleaf bluebells	Avoided
Slender phlox	Avoided
Nattall's sandwort	Accepted
Alpine stitchwort	Accepted
Blue forget-me-not	Accepted
Yellow pincushion	Accepted
White-flowered pincushion	Accepted
Narrowleaf goldenweed	Accepted
Tufted evening primrose	Accepted
Plains prickly pear	Accepted
Alpine aster	Accepted
Clustered broomrape	Accepted
Little ricegrass	Accepted
Mountain sweet cicely	Accepted
White locoweed	Accepted
Woolly groundsel	Accepted
Splitleaf groundsel	Accepted
Rocky Mountain groundsel	Accepted
Buek's groundsel	Accepted
Western wheatgrass	Accepted

PEGR2	Scrophulariaceae	Pedicularis groenlandica	Elephanthead lousewort	Accepted
PESI	Cactaceae	Pediocactus simpsonii	Mountain ball cactus	Accepted
PEAT3	Scrophulariaceae	Penstemon attenuatus	Sulphur penstemon	Accepted
PECY3	Scrophulariaceae	Penstemon cyaneus	Blue penstemon	Accepted
PEDE4	Scrophulariaceae	Penstemon deustus	Scabland penstemon	Accepted
PEER	Scrophulariaceae	Penstemon eriantherus	Fuzzy-tongue penstemon	Accepted
PEFR3	Scrophulariaceae	Penstemon fruticosus	Bush penstemon	Accepted
PEMOI2	Scrophulariaceae	Penstemon montanus var. idahoensis	Cordroot beardtongue	Accepted
PEPE12	Scrophulariaceae	Penstemon perpulcher	Minidoka beardtongue	Accepted
PEPR2	Scrophulariaceae	Penstemon procerus	Alpine beardtongue	Accepted
PEPU12	Scrophulariaceae	Penstemon pumilus	Salmon river beardtongue	Accepted
PERY	Scrophulariaceae	Penstemon rydbergii	Rydberg's penstemon	Accepted
PEWI	Scrophulariaceae	Penstemon wilcoxii	Wilcox's penstemon	Accepted
PHGL2	Hydrophyllaceae	Phacelia glandulifera	Sticky phacelia	Accepted
PHHA	Hydrophyllaceae	Phacelia hastata	Silverleaf phacelia	Accepted
PHLI	Hydrophyllaceae	Phacelia linearis	Threadleaf phacelia	Accepted
PHSE	Hydrophyllaceae	Phacelia sericea	Silky phacelia	Accepted
PHAL2	Poaceae	Phleum alpinum	Alpine timothy	Accepted
PHPR3	Poaceae	Phleum pratense	Timothy grass	Accepted
PHAU3	Polemoniaceae	Phlox austromontana	Mountain phlox	Accepted
PHDI3	Polemoniaceae	Phlox diffusa	Spreading phlox	Accepted
PHHO	Polemoniaceae	Phlox hoodii	Spiny phlox	Accepted
PHHOM	Polemoniaceae	Phlox hoodii ssp. Muscoides	Musk phlox	Accepted
PHLO2	Polemoniaceae	Phlox longifolia	Longleaf phlox	Accepted
PHMU3	Polemoniaceae	Phlox multiflora	Rocky mountain phlox	Accepted
PHPU5	Polemoniaceae	Phlox pulvinata	Cushion phlox	Accepted
PHCH	Brassicaceae	Phoenicaulis cheiranthoides	Dagger pod	Accepted
PHEM	Ericaceae	Phyllodoce empetriformis	Pink mountain heath	Accepted
PHDI6	Brassicaceae	Physaria didymocarpa	Common twinpod	Accepted
PIEN	Pinaceae	Picea engelmannii	Engelmann spruce	Avoided

PIDE4	Asteraceae	Picrothamnus desertorum	Bud sagebrush	Accepted	
PIAL	Pinaceae	Pinus albicaulis	Whitebark pine	Avoided	
PICO	Pinaceae	Pinus contorta	Lodgepole pine	Avoided	
PIFL2	Pinaceae	Pinus flexilis	Limber pine	Avoided	
PIPO	Pinaceae	Pinus ponderosa	Ponderosa pine	Avoided	
PLMA2	Plantaginaceae	Plantago major	Broadleaf plantain	Avoided	
PLAQ2	Orchidaceae	Platanthera aquilonis	Northern green orchid	Accepted	
POAR2	Poaceae	Poa arctica	Artic bluegrass	Accepted	
POBU	Poaceae	Poa bulbosa	Bulbous bluegrass	Accepted	
POCO	Poaceae	Poa compressa	Canada bluegrass	Accepted	
POCU3	Poaceae	Poa cusickii	Cusick's bluegrass	Accepted	
POFE	Poaceae	Poa fendleriana	Muttongrass	Accepted	
PONEI2	Poaceae	Poa nemoralis ssp. interior	Inland bluegrass	Accepted	
POPA2	Poaceae	Poa palustris	Fowl bluegrass	Accepted	
POPR	Poaceae	Poa pratensis	Kentucky bluegrass	Accepted	
POSE	Poaceae	Poa secunda	Sandberg bluegrass	Accepted	
POWH2	Poaceae	Poa wheeleri	Wheeler's bluegrass	Accepted	
POPU3	Polemoniaceae	Polemonium pulcherrimum	Jacob's ladder	Accepted	
POVI	Polemoniaceae	Polemonium viscosum	Sticky polemonium	Accepted	
POBI6	Polygonaceae	Polygonum bistortoides	American bistort	Accepted	
PODOJ2	Polygonaceae	Polygonum douglasii ssp johnstonii	Johnston's knotweed	Accepted	
POPO4	Polygonaceae	Polygonum polygaloides	Milkwort knotweed	Accepted	
POVI3	Polygonaceae	Polygonum viviparum	Alpine bistort	Accepted	
POBA2	Salicaceae	Populus balsamifera	Balsam poplar	Accepted	
POTR5	Salicaceae	Populus tremuloides	Quaking aspen	Accepted	
POBR5	Rosaceae	Potentilla brevifolia	Sparseleaf cinquefoil	Accepted	
PODI2	Rosaceae	Potentilla diversifolia	Mountain meadow cinquefoil	Accepted	
POGL9	Rosaceae	Potentilla glandulosa	Sticky cinquefoil	Accepted	
POOV2	Rosaceae	Potentilla ovina	Sheep cinquefoil	Accepted	
PSSP6	Poaceae	Pseudoroegneria spicata	Bluebunch wheatgrass	Accepted	

PSME	Pinaceae	Pseudotsuga menziesii	Douglas-fir	Avoided
PUTR2	Rosaceae	Purshia tridentata	Antelope bitterbrush	Accepted
PYEL	Pyrolaceae	Pyrola elliptica	Wax-flower shinleaf	Avoided
PYMI	Pyrolaceae	Pyrola minor	Snowline wintergreen	Avoided
RAAC3	Ranunculaceae	Ranunculus acris	Tall buttercup	Accepted
RAAN	Ranunculaceae	Ranunculus andersonii	Anderson's buttercup	Accepted
RAES	Ranunculaceae	Ranunculus eschscholtzii	Eschscholtz's buttercup	Accepted
RAGL	Ranunculaceae	Ranunculus glaberrimus	Sagebrush buttercup	Accepted
RATE	Ranunculaceae	Ranunculus testiculatus	Bur buttercup	Accepted
RIAU	Grossulariaceae	Ribes aureum	Golden currant	Accepted
RICE	Grossulariaceae	Ribes cereum	Wax currant	Accepted
RIHU	Grossulariaceae	Ribes hudsonianum	Northern black currant	Accepted
RIMO2	Grossulariaceae	Ribes montigenum	Gooseberry currant	Accepted
RILE2	Asteraceae	Rigiopappus leptocladus	Wireweed	Accepted
ROWO	Rosaceae	Rosa woodsii	Western wild rose	Accepted
RUAR9	Rosaceae	Rubus armeniacus	Himalayan blackberry	Accepted
RUUR	Rosaceae	Rubus ursinus	Pacific blackberry	Accepted
RUCR	Polygonaceae	Rumex crispus	Curly dock	Accepted
RUSA	Polygonaceae	Rumex salicifolius	Willow dock	Accepted
SAAR27	Salicaceae	Salix arctica	Artic willow	Accepted
SABE2	Salicaceae	Salix bebbiana	Bebb's willow	Accepted
SABO2	Salicaceae	Salix boothii	Booth's willow	Accepted
SAGE2	Salicaceae	Salix geyeriana	Geyer willow	Accepted
SALU	Salicaceae	Salix lucida	Shining willow	Accepted
SALU2	Salicaceae	Salix lutea	Yellow willow	Accepted
SAME2	Salicaceae	Salix melanopsis	Dusky willow	Accepted
SANI8	Salicaceae	Salix nivalis	Snow willow	Accepted
SAPL2	Salicaceae	Salix planifolia	Diamond leaf willow	Accepted
SAWO	Salicaceae	Salix wolfii	Wolf's willow	Accepted
SAVE4	Chenopodiaceae	Sarcobatus vermiculatus	Greasewood	Accepted

SAOC4	Saxifragaceae	Saxifraga occidentalis	Western saxifrage	Accepted	
SARH2	Saxifragaceae	Saxifraga rhomboidea	Diamond-leaf saxifrage	Accepted	
SCAR7	Poaceae	Schedonorus arundinaceus	Tall rye grass	Accepted	
SEDE	Crassulaceae	Sedum debile	Orpine stonecrop	Accepted	
SELA	Crassulaceae	Sedum lanceolatum	Spearleaf stonecrop	Accepted	
SEHY	Asteraceae	Senecio hydrophiloides	Tall groundsel	Accepted	
SEIN2	Asteraceae	Senecio integerrimus	Lambstongue ragwort	Accepted	
SEME	Asteraceae	Senecio megacephalus	Large-headed ragwort	Accepted	
SESE2	Asteraceae	Senecio serra	Tall ragwort	Accepted	
SESP4	Asteraceae	Senecio sphaerocephalus	Ballhead ragwort	Accepted	
SHCA	Elaeagnaceae	Shepherdia canadensis	Russet buffaloberry	Accepted	
SIPR	Rosaceae	Sibbaldia procumbens	Creeping sibbaldia	Accepted	
SIDO	Caryophyllaceae	Silene douglasii	Douglas's catchfly	Accepted	
SIRE3	Caryophyllaceae	Silene repens	Creeping silene	Accepted	
SIAL2	Brassicaceae	Sisymbrium altissimum	Tall tumble mustard	Accepted	
SIID	Iridaceae	Sisyrinchium idahoense	Idaho blue eyed grass	Accepted	
SMCA	Brassicaceae	Smelowskia calycina	Alpine smelowskia	Accepted	
SOMI2	Asteraceae	Solidago missouriensis	Missouri goldenrod	Accepted	
SOMU	Asteraceae	Solidago multiradiata	Rocky mountain goldenrod	Accepted	
STCR	Caryophyllaceae	Stellaria crassifolia	Fleshy starwort	Accepted	
STLO2	Caryophyllaceae	Stellaria longipes	Longstalk starwort	Accepted	
STAC	Asteraceae	Stenotus acaulis	Stemless mock goldenweed	Accepted	
STLA7	Asteraceae	Stenotus lanuginosus	Wooley mock goldenweed	Accepted	
STTE2	Asteraceae	Stephanomeria tenuifolia	Narrow leaved wire lettuce	Avoided	
SWPE	Gentianaceae	Swertia perennis	Star swertia	Accepted	
SYOR2	Caprifoliaceae	Symphoricarpos oreophilus	Mountain snowberry	Accepted	
SYAS3	Asteraceae	Symphyotrichum ascendens	Western aster	Accepted	
SYEA2	Asteraceae	Symphyotrichum eatonii	Eaton's aster	Accepted	
SYFOF	Asteraceae	Symphyotrichum foliaceum	Alpine leafy-bract aster	Accepted	
SYPIC	Scrophulariaceae	Synthyris pinnatifida var. canescens	Cut-leaf kittentail	Accepted	67

TACA8	Poaceae	Taeniatherum caput-medusae	Medusahead	Avoided
TAOF	Asteraceae	Taraxacum officinale	Common dandelion	Accepted
TECA2	Asteraceae	Tetradymia canescens	Spineless horsebrush	Accepted
TEGR3	Asteraceae	Tetraneuris grandiflora	Old man of the mountain	Accepted
THOC	Ranunculaceae	Thalictrum occidentale	Western meadow rue	Avoided
THPL	Cupressaceae	Thuja plicata	Western red cedar	Avoided
TOAL	Asteraceae	Townsendia alpigena	Wyoming townsend daisy	Accepted
TRDU	Asteraceae	Tragopogon dubius	Western goats beard	Accepted
TRGY	Fabaceae	Trifolium gymnocarpon	Holly-leaf clover	Accepted
TRGYG	Fabaceae	Trifolium gymnocarpon ssp. gymnocarpon	Plummer's clover	Accepted
TRPL2	Fabaceae	Trifolium plumosum	Plumed clover	Accepted
TRPR2	Fabaceae	Trifolium pratense	Red clover	Accepted
TRRE3	Fabaceae	Trifolium repens	White clover	Accepted
TRSP2	Poaceae	Trisetum spicatum	Narrow false oat	Accepted
TRGR7	Liliaceae	Triteleia grandiflora	Wild hyacinth	Accepted
TRLA14	Ranunculaceae	Trollius laxus	American globeflower	Accepted
VASC	Ericaceae	Vaccinium scoparium	Grouse whortleberry	Avoided
VAAC	Valerianaceae	Valeriana acutiloba	Sharpleaf valerian	Accepted
VAACP	Valerianaceae	Valeriana acutiloba var. pubicarpa	Mountain valerian	Accepted
VETH	Scrophulariaceae	Verbascum thapsus	Common mullein	Avoided
VEWO2	Scrophulariaceae	Veronica wormskjoldii	American alpine speedwell	Avoided
VIAD	Violaceae	Viola adunca	Western dog violet	Accepted
VIPU4	Violaceae	Viola purpurea	Goosefoot violet	Accepted
VIPUV2	Violaceae	Viola purpurea ssp. venosa	Purple-marked yellow violet	Accepted
VIVA	Violaceae	Viola vallicola	Sagebrush violet	Accepted
WOOR	Dryopteridaceae	Woodsia oregana	Oregon woodsia	Accepted
ZIEL2	Liliaceae	Zigadenus elegans	Mountain death camas	Avoided
ZIVE	Liliaceae	Zigadenus venenosus	Meadow death camas	Avoided
ZIAQA2	Poaceae	Zizania aquatica var. aquatica	Annual wildrice	Accepted

Appendix B. Species-specific linear regressions of p1ant biomass against plant cover (%), sample date (i.e., Julian date), log transformations of plant cover and sample date, and interactions between plant cover and sample date with/without log transformations. Coefficients are shown for variables included in the best model for each species or growth-form group, along with the adjusted R^2 value of the model. Species for which it was not appropriate to fit a regression model were assigned a species-specific mean biomass value (see Methods) in the species cover column. We used these models to estimate forage biomass in all unclipped quadrats in the Lost River, East Fork, and Owyhee bighorn sheep population ranges in Idaho, USA.

Plant Code	# of samples	Adj. R^2	Intercept	Plant cover	log(Plant cover)	Julian	log(Julian)	Plant cover: Julian	log(Plant cover): log(Julian)
ACMIO	18		0.00	1.14					
AGGL	24	0.98	-0.79	0.06		-0.01		0.03	
AGHE2	7		0.00	1.13					
ALAC4	21		0.00	0.33					
ALAL3	14		0.00	2.13					
ALBR	2		0.00	0.56					
ALDE	10		0.00	2.19					
ANDI2	16	0.40	0.68	1.69					
ANLA3	6		0.00	0.74					
ANMI3	24	0.36	1.28	2.70					
ARAC2	51	0.67	41.10	-35.87		-0.22		0.22	
ARARA	13	0.89	6.20		5.56		-0.97		-0.91
ARARL	14	0.59	90.16	-32.35		-0.51		0.23	
ARCO	9		0.00	0.57					
ARCO9	2		0.00	0.07					
ARHI	5		0.00	0.35					
ARHO2	32		0.00	0.40					
ARTRT	13	0.80	123.97	-10.92		-0.67		0.09	
ARTRV	20	0.89	-15.64	11.17		0.11		-0.04	

ARTRW8	67	0.78	22.76	-4.95	
ARWIS	2		0.00	0.87	
ASAU4	2		0.00	1.10	
ASLE8	2		0.00	1.02	
ASOB4	10		0.00	1.99	
ASPU9	15	0.97	-2.35	5.81	
BLSC	6		0.00	0.61	
BRTE	71	0.42	32.78	-40.13	
CAAN7	28		0.00	1.25	
CACO36	6		0.00	0.46	
CADO2	4		0.00	2.84	
CAFL7	3		0.00	1.63	
CAGE2	31	0.59	1.34		1.52
CAMI2	2		0.00	0.66	
CAMI7	4		0.00	1.46	
CANU3	8		0.00	1.41	
CAPU	2		0.00	7.56	
CARU	3		0.00	4.42	
CHDOA	2		0.00	0.78	
CHVI8	14	0.54	1.61		0.69
CIFO	4		0.00	1.61	
CIVU	4		0.00	2.84	
COPA3	43		0.00	0.61	
CRAC2	24		0.00	1.28	
CROC	20	0.50	-77.31	155.07	
CRSO3	3		0.00	1.29	
CYNI3	10		0.00	0.76	
DEBI	5		0.00	0.84	
DEIN5	12		0.00	0.12	
DRCR2	4		0.00	2.36	

-0.10	0.06
-0.21	0.32

0.50

-1.00

DROL	3		0.00	3.95				
DRTR3	2		0.00	3.07				
DRVE2	14		0.00	0.34				
ELEL5	52	0.75	6.63	4.86		-0.04		
ERAS	16		0.00	1.16				
ERBL	8		0.00	0.30				
ERCO4	16	0.79	13.15	-19.46		-0.07		0.12
ERCO5	3		0.00	1.20				
ERLA14	2		0.00	0.54				
ERLE6	3		0.00	1.50				
ERNA10	16	0.90	54.73	-83.23		-0.37		0.61
EROV	23	0.93	-0.16	6.94		0.00		-0.02
ERSU13	7		0.00	1.35				
ERUM	10	0.47	1.06		1.47			
FEID	47	0.77	8.47		0.84		-1.32	
FRAT	3		0.00	0.12				
GADI2	16		0.00	0.19				
GARA2	3		0.00	0.43				
GERO2	4		0.00	2.27				
GETR	12	0.81	-0.29	4.39				
GLST	12	0.71	-86.18	171.78		0.42		-0.81
HAGL	2		0.00	1.28				
HAMI	2		0.00	4.75				
HECOI	20	0.84	1.31	4.96				
HEPA11	3		0.00	0.56				
HOJU	4		0.00	3.06				
IOAL	19	0.80	1.10	-17.08		-0.01		0.14
IPCO5	2		0.00	0.33				
LAOC3	7		0.00	0.37				
LASE	3		0.00	0.13				

	0.12
	0.61 -0.02
-1.32	

LENU8	2		0.00	2.54					
LERE7	11		0.00	0.22					
LESE17	4		0.00	2.59					
LILE3	13		0.00	1.02					
LIPU11	13	0.57	-3.41	10.46					
LIRU4	4		0.00	1.28					
LOMA3	5		0.00	1.15					
LOTR2	5		0.00	0.76					
LUSE2	3		0.00	1.95					
LUSE4	30	0.83	-0.64	5.71					
MACA2	2		0.00	0.76					
MEDI	3		0.00	0.19					
MEOB	6		0.00	0.95					
MIGR	11		0.00	0.43					
OPPO	4		0.00	13.42					
PACA15	10	0.93	-4.66	9.42					
PASM	16	0.78	-0.95	8.60					
PEDE4	3		0.00	4.15					
PEER	17	0.84	-27.29	49.13		0.13		-0.23	
PEMOI2	3		0.00	0.27					
PEPR2	33	0.54	-3.42	4.53		0.02		-0.02	
PESI	3		0.00	9.24					
РННА	9		0.00	1.16					
РННОМ	41	0.47	-12.74		1.64		2.81		
PHLI	3		0.00	0.46					
PHLO2	42	0.29	-1.40	4.76					
PHMU3	13	0.73	-68.03		95.26		13.12		-17.84
PODI2	6		0.00	1.75					
POPR	14	0.79	-62.61	139.47		0.30		-0.66	
POSE	136	0.48	8.22	2.53		-0.04			

PSSP6	130	0.72	30.84	-9.01		-0.15		0.09	
PYMI	2		0.00	0.16					
RAGL	13	0.96	1.48	2.78		-0.01			
RATE	3		0.00	0.41					
SARH2	5		0.00	1.98					
SEDE	2		0.00	2.46					
SELA	34	0.58	-2.81	7.46					
SESP4	4		0.00	0.79					
SIAL2	8		0.00	0.65					
SOMU	3		0.00	7.08					
STAC	16	0.52	9.63		-31.50		-1.60		6.36
SYOR2	11	0.59	0.12	2.44					
TAOF	9		0.00	0.49					
TOAL	3		0.00	1.15					
VIPU4	6		0.00	0.33					
VIVA	2		0.00	0.73					
ZIVE	3		0.00	1.89					
Basal forbs	208	0.58	7.99	0.19		-0.03		0.03	
Short forbs	514	0.52	1.39	-3.46		0.00		0.04	
Tall forbs	574	0.61	1.83	-1.37		0.00		0.03	
Graminoids	639	0.59	15.59	-2.83		-0.07		0.05	
Mid-tall shrubs	42	0.43	1.68		0.54				
Evergreen shrubs	209	0.68	2.04		0.73				
Evergreen-trees	17	0.88	2.34		1.01				