Do prey select for vacant hunting domains to minimize a multi-predator threat?

Abstract
Many ecosystems contain sympatric predator species that hunt in different places and times. We tested whether this provides vacant hunting domains, places and times where and when predators are least active, that prey use to minimize threats from multiple predators simultaneously. We measured how northern Yellowstone elk (Cervus elaphus) responded to wolves (Canis lupus) and cougars (Puma concolor), and found that elk selected for areas outside the high-risk domains of both predators consistent with the vacant domain hypothesis. This enabled elk to avoid one predator without necessarily increasing its exposure to the other. Our results demonstrate how the diel cycle can serve as a key axis of the predator hunting domain that prey exploit to manage predation risk from multiple sources. We argue that a multi-predator, spatiotemporal framework is vital to understand the causes and consequences of prey spatial response to predation risk in environments with more than one predator.

Keywords
Antipredator behaviour, cougar, diel activity, elk, habitat selection, predation risk, predator facilitation, risk enhancement, synergistic predation, wolf.

ECOLOGY LETTERS
We tested the extent that elk (*Cervus elaphus*) selected for vacant hunting domains to avoid predation from wolves (*Canis lupus*) and cougars (*Puma concolor*) in northern Yellowstone National Park (YNP). Wolves are cursorial predators that kill mainly in flat, open areas at morning and dusk (Kauffman et al. 2007; Kohl et al. 2018), whereas cougars are spot-and-stalk/ambush predators that kill mainly in topographically rugged, forested areas at night (Murphy et al. 1998; Ruth et al. 2019). We predicted that elk selected for flat, open areas at night (night-flat and night-open domains), and for rugged, forest areas during daylight (day-rugged and day-forest domains). We studied elk habitat selection in winter when wolves and cougars were the only major elk predators inside YNP (Evans et al. 2006; Barber-Meyer et al., 2010), and during a period (2001–2004) when densities of both predators were highest (Cubaynes et al. 2014; Ruth et al. 2019). Our results shed new light on how multiple predators can drive prey habitat selection in a predator-rich environment, and how prey can minimize these multiple threats simultaneously.

**MATERIALS AND METHODS**

**Study area**

Our study occurred within the northern Yellowstone elk winter range (hereafter, northern Yellowstone). This 1520-km² area consists of low-elevation (1500–2600 m) grasslands and shrub steppes that fan out from the Yellowstone River and its tributaries along the northern border of YNP and adjacent areas in Montana (Lemke, Mack & Houston 1998). Approximately 65% (995-km²) of the winter range is located within the Park (hereafter, northern YNP). Most of the data in our study were collected in northern YNP (Fig. 1, 2 in Appendix S1) because wolves were concentrated there (Stahler et al. 2016).

**Study population**

We analysed habitat selection behaviour of 27 adult (> 1 year-old) female elk from the northern Yellowstone elk population, which annually occupies the winter range from about 15 October to 31 May. This population ranged from approximately 10 700 to 17 600 individuals during the study (Tallian et al. 2017). Adult female elk were fitted with GPS radio-collars that collected locations at 4–6 h intervals. We limited our analysis to winter locations collected from 1 November to 30 April. Appendix S1: Fig. 1 illustrates the spatial distribution of these data.

Wolf abundance in northern YNP ranged between 70 and 98 individuals in 4–8 packs (mean individuals/pack = 8.36; SE = 0.90). Each winter, 20 to 30 wolves (35–40% of northern...
YNP wolf population), including 30–50% of pups born in the previous year were captured and fitted with VHF (n = 72 [11 unique packs], 2000–2004) or GPS radio-collars (n = 21 [12 unique packs], 2004–2013; Smith et al. 2004). Cougar abundance ranged between 26 and 42 individuals (Ruth et al. 2019). We monitored 54 (27 females, 27 males) radio-collared adult and independent, subadult cougars (68–93% of estimated adult cougar population; Ruth et al. 2011), including six GPS-collared individuals that were the focus of our activity analysis.

Predation sampling

We searched for elk killed by wolves (Fig. 2a in Appendix S1) and cougars (Fig. 2b in Appendix S1) from 1 November to 31 May by radio-tracking VHF- and GPS-collared wolves and cougars. We excluded adult male elk from our analysis because wolves killed male and female elk in different areas (Kohl et al. 2018), and because GPS data were unavailable to assess male habitat selection. We analysed wolf-killed elk (n = 400 adult females and calves) collected from 01 November 2000 to 31 May 2004 to correspond with the winters in which we monitored elk. We analysed cougar-killed elk (n = 257 adult females and calves) collected during winter over a longer period (April 1998 to May 2005) to maximize sample size (Appendix S2).

Spatial variation in predation risk

We used topographic roughness (Fig. 3a in Appendix S1) and vegetation openness (Fig. 3b in Appendix S1) as separate indices of spatial variation in predation risk from wolves (Creel et al. 2005; Fortin et al. 2005; Mao et al. 2005; Kohl et al. 2018) and cougars (Atwood et al. 2009; Bartnick et al. 2013). We did not consider human harvest as an important source of predation risk because only 1.8% of all elk locations and only one of 27 collared elk occurred outside YNP. We also did not consider the effects of other spatial covariates on elk habitat selection. Nevertheless, openness plausibly indicated the influence of forage availability, thermal cover and predation risk on elk habitat selection.

We developed and analysed single maps of openness and roughness because neither varied during the study period. Both spatial risk indices (30 × 30 m grid cell) were developed using the Focal tool within the raster package in R 3.2.3. Openness and roughness were not highly correlated (Pearson’s correlation coefficient, r = −0.18). We calculated openness (range 0 [dense forest] – 289 [open grassland]) as the sum of non-forested cells within a 500 × 500 m moving window centred on each grid cell following Boyce et al. (2003) using data from the LANDFIRE program in 2001 (Fig. 4 in Appendix S1). We calculated roughness (range 0 [flat]–1114 [cliff]) as the sum of the absolute value of the difference in elevation between each grid cell and surrounding eight neighbours (3 × 3 window) following criteria developed for Yellowstone cougars (Ruth et al. 2019).

To determine if openness and roughness were valid indices of predation risk, we separately modelled the relative probability of a wolf- or cougar-killed elk (adult females and calves only) as a function of openness or roughness using a resource selection function framework (RSF; Manly et al. 2002). We analysed wolf and cougar kill RSFs with a generalized additive model (GAM) to account for potential nonlinear effects of openness and roughness on the probability of a kill at a given location (Appendix S2).

Diel activity patterns

We used movement rate to index the diel activity patterns of wolves and cougars because speed of locomotion is a valid proxy for diel activity patterns in large mammals (Ensing et al. 2014; Vander Vennen et al. 2016). In addition, ungulates, including elk, are sensitive to temporal variation in predator locomotion (Fröhlich et al. 2012; Kohl et al. 2018). For wolves, we estimated movement rate at each hour of the day from the hourly winter positions of 21 GPS-collared animals during 2004–2013.

For cougars, we estimated movement rate at each hour of the day from winter locations collected at 3-h intervals from 6 GPS-collared individuals (2 females, 4 males) during 2001–2006. Movement rate equalled the Euclidean distance of the preceding 5-h time step for wolves and 6-h time step for cougars as these intervals most closely matched the 5-h interval between consecutive elk locations. We subsampled the wolf and cougar data to match the elk data by retaining every fifth wolf location or second cougar location. We used only consecutive 5- or 6-h locations to calculate movement rates.

We modelled the mean diel movement rate by applying a generalized additive mixed model (GAMM) to the 5-h wolf and 6-h cougar step-lengths. For wolves, we could not distinguish between individual and annual variation in diel activity due to small sample sizes within years (Kohl et al. 2018). For cougars, we included a fixed effect for sex to account for potential differences in movement rate between males and females (Wang et al. 2015; Appendix S2).

Elk habitat selection

For each spatial risk index (openness and roughness), we evaluated models that tested whether elk selection for risky and safe places was (1) independent of wolf and cougar activity (‘space-only’ models), (2) dependent on the activity of wolves or cougars (‘single-predator space × activity’ models), or (3) dependent on the activity of wolves and cougars (‘multi-predator space × activity’ models). Space × activity models included terms for the interaction between spatial risk and mean diel movement rate of wolves and/or cougars. These models therefore evaluated how elk selection for risky and safe places at the end of a 5-h movement step was affected by the mean movement rate of predators during that step. Because male and female cougars exhibited different diel activity patterns (see Results), we specified separate models for the effects of male and female cougars. Our multi-predator models considered only wolves and male cougars because the mean activity patterns of wolves and female cougars were collinear (Appendix S2).
For each model, we tested different forms of the relationship between habitat selection and spatial risk to account for how elk tolerate low levels of spatial risk (Kohl et al. 2018). Specifically, we tested for a response threshold in elk habitat selection by comparing models with a linear effect for spatial risk to models with a threshold effect specified by two linear splines. We performed a grid search of candidate models to determine the presence and position of thresholds (Appendix S2) and identified the best-fitting model using the quasi-likelihood under independence criteria (QIC; Pan 2001). We performed 5-fold cross validation \((n = 1000)\) for a step-selection function design to evaluate the predictive accuracy of each best-fit model (Boyce et al. 2002). Average Spearman rank correlations \((r_s) > 0.70\) indicated satisfactory fit of models to data (Boyce et al. 2002).

Visualizing elk selection for vacant hunting domains
We used our models of diel predator movement rate and spatial kill probability to visualize cougar and wolf hunting domains. The cougar hunting domain was based on the activity of males and the kills of males and females. We excluded female activity because it was a poor predictor of elk habitat selection (see Results), and included kills by both sexes because their spatial distributions with respect to openness and roughness were similar (Appendix S3). Using predicted mean values of diel predator movement rate and spatial kill probability, we calculated percentile-specific ellipses to illustrate the hunting domain of each predator. The 90th percentile ellipse indicated the places and times where and when predation risk was greatest. We overlaid the ellipses on a contour plot of elk habitat selection with respect to spatial risk (openness and roughness) and time of day to reveal the extent that elk selected for vacant hunting domains (see Appendix S2). This provided a visualization of elk habitat selection relative to the hunting domains of wolves and male cougars that was comparable to the vacant domain hypothesis illustrated in Figure 2. Our visualization assumes that areas occurred within the study area where all possible values of openness (0–289) or roughness (0–150) were accessible to an elk within a 5-h movement step. We demonstrate the validity of this assumption in Appendix S2.

RESULTS
Hunting domains
Wolves and cougars hunted elk (adult females and calves) in different places at different times of the day. Cougars mainly killed elk in moderately forested, rugged areas while wolves killed elk in open, flat areas (Fig. 3). And whereas cougars, especially males, hunted mainly at night, wolves hunted mainly during morning and at dusk (Fig. 4; Appendix S4). The diel activity pattern of male cougars was unimodal with a single peak in activity at 0300-h (Fig. 4a). Female cougars exhibited a bimodal pattern, with comparatively lower activity peaks at about 0400-h and 2000-h (Fig. 4a). The activity pattern of wolves was also bimodal with activity peaks at 1000-h and 2000-h (Fig. 4b). Together, cougars hunting at night in rugged forests, and wolves hunting in daylight (morning/dusk) in flat open areas indicated four vacant hunting domains: ‘night-open’, ‘night-flat’, ‘day-rough’ and ‘day-forest’.

Elk habitat selection
The effect of spatial risk on elk habitat selection was nonlinear. For each spatial risk index (openness, roughness), the top space-only model included a linear spline for spatial risk (Appendix S5, S6), indicating a threshold at which the effect of spatial risk on habitat selection changed. Evidence against a model describing a simple linear relationship between spatial risk and habitat selection was strong for openness (AQIC = 37.3; Appendix S5) and roughness (AQIC = 89.2; Appendix S6).

Support for the top space-only models was substantially weaker compared to single-predator models that included space × activity interactions between mean diel movement rate of male cougars (or wolves) and linear splines for openness (AQIC = 342.4–360.6; Appendix S5) or roughness (AQIC = 34.5–111.3; Appendix S6). Similar models including
mean diel movement rate of female cougars differed little from the top space-only model (openness: ΔQIC = 2.6; roughness: ΔQIC = −0.7; Appendix S5, S6), indicating little or possibly no effect of female cougars on elk habitat selection. Support for the single-predator wolf models was also notably weaker compared to the single-predator male cougar models (openness: ΔQIC = 18.3; roughness: ΔQIC = 76.8; Appendix S5 and S6). This suggests that male cougars had a stronger singular effect on elk habitat selection than did wolves.

However, support for the single-predator male cougar model was substantially weaker compared to multi-predator models that also included the activity of wolves (openness: ΔQIC = 38.7; roughness: ΔQIC = 10.9; Appendix S5, S6). This suggests that elk responded to both predators simultaneously. Support for models that included the activity of male and female cougars was weaker compared to the best-fit models that included the activity of male cougars and wolves (openness: ΔQIC = 13.5; roughness: ΔQIC = 1.9; Appendix S5, S6). Fivefold cross validation revealed a strong correlation between observed and predicted values for these best-fit models (openness: r_s = 0.98; roughness: r_s = 0.95). Correlations of this magnitude indicate that these models were reliable.

Coefficients for the best-fit models (Appendix S7) indicate that when male cougar activity was high and wolf activity was low, elk selected for open areas (Fig. 5a, red line) and avoided rough areas (Fig. 5b, red line). Conversely, when wolf activity was high and male cougar activity was low, elk avoided the most open areas (Fig. 5a, blue line: descending limb) and selected for rougher areas (Fig. 5b, blue line: ascending limb). The subsequent shift to avoidance with increasing levels of roughness (Fig. 5b, blue line: descending limb) may reflect a latent response to female cougars. Females were more active than males when wolf activity was high (Fig. 4a), and the level of roughness where selection switched to avoidance approximates the point beyond which the probability of cougar predation exceeded that of wolf predation (Fig. 3b).

**Selection for vacant predator hunting domains**

Figure 6 illustrates the four vacant hunting domains: day-forest (< ~200 openness), night-open (> ~200 openness) (Fig. 6a), day-rough (> ~75 roughness) and night-flat (< ~75 roughness) (Fig. 6b). The best-fit model for openness (Appendix S7a) indicated that elk selected for the day-forest and night-open domains (Fig. 6a), and the best-fit model for roughness (Appendix S7b) indicated that elk selected for the night-flat domain but not the day-rough domain (Fig. 6b). During daylight hours, elk tended to select for rougher areas outside the riskiest wolf domain (90th percentile) and within the lower risk wolf domains (70–80th percentile).

**DISCUSSION**

Most prey species live in environments with more than one predator species, yet few empirical studies have tested the simultaneous effects of multiple predators on prey space use, especially in free-living vertebrate systems (Say-Sallaz et al. 2019). And whereas most predators exhibit predictable fluctuations in activity across the diel cycle (Kronfeld-Schor & Dayan 2003), previous multi-predator studies neither measured continuous variation in diel predator activity nor linked it to prey habitat selection as we did. These studies only compared prey space use between light and dark periods (e.g., Kotler et al. 1992, Clark et al. 2003, Lone et al., 2017). This approach would have obscured our results because wolf and cougar activity patterns were complex functions of time of day that did not neatly fit the conventional dichotomy of safe and dangerous periods. As far as we know, our study is the first to quantify how continuous variation in spatial risk (Fig. 3) and diel activity (Fig. 4) of more than one predator simultaneously affects the habitat selection of a shared prey. Our results advance understanding about the spatial ecology of predator–prey interactions in two important ways.

First, we demonstrate how it is perilous to assume that prey habitat selection in a multi-predator environment is sensitive to just one predator species. When wolves were reintroduced to Yellowstone National Park in 1995–1997 (Bangs & Fritts 1996), they joined a system that was already populated by other predators of elk, including growing numbers of grizzly...
bears (*Ursus arctos*, Kamath et al. 2015) and cougars (Ruth et al. 2019). Despite this predator diversity, subsequent research and commentary about elk space use in and around Yellowstone have assumed, implicitly or explicitly, that wolves are the only (or primary) predator that elk respond to (e.g., Ripple & Larsen 2000; Laundré et al. 2001; Creel et al. 2005; Fortin et al. 2005; Mao et al. 2005; Middleton et al., 2013a, 2013b; Kohl et al. 2018; Cusack et al. 2019). Our study is the first to test this long-held assumption, and our results suggest it is false. We found strong evidence that elk habitat selection was shaped by the risk of predation from wolves and cougars. Specifically, our top wolf-only models of elk habitat selection performed poorly compared to our best-fit models that included wolves and male cougars (openness: $\Delta$QIC = 57.0; roughness: $\Delta$QIC = 87.7; Appendix S5, S6).

We also found that male cougars, not wolves, exerted the most pressure on elk habitat selection. The evidence for this is twofold. First, there was substantially less support for our top wolf models compared to our top male cougar models (Appendix S5, S6). Had we ignored the simultaneous effects of wolves and male cougars, we would have concluded that wolves had no effect on elk habitat selection. Second, when we accounted for these simultaneous effects, we found that the magnitude of the male cougar effect was 25–80% greater than that of the wolf effect (Appendix S7). These findings suggest that a fixation on wolves in systems with cougars could lead to misleading conclusions about the causes and consequences of elk response to predation risk. This problem may apply to other multi-predator systems where one predator species is assumed primary to another.

We are uncertain why the strong cougar effect was linked to the activity of males and not females. Poor support for the female cougar models relative to the space-only models that excluded all predator activity (Appendix S5, S6) suggests that elk were insensitive to female cougars. Yet models that included male and female cougars ranked second to our best-fit models that included male cougars and wolves (Appendix S5, S6). Near equal support for these models with respect to roughness ($\Delta$QIC = 1.9; Appendix S6) indicates that we cannot rule out an influence of female cougars. We expected such an influence because adult females were more than three times as abundant as resident adult males, females with kittens had a higher kill rate than males (Ruth et al. 2019), and at least one female in our sample had kittens. On the other hand, the tendency for females, which are smaller than males, to focus their predation on elk calves (Murphy et al. 1998; Ruth et al. 2019), together with the possibility that few radio-collared elk had a calf at heel due to a scarcity of calves at the time (Proffitt et al. 2014), might have diminished the pressure of female cougars on elk habitat selection.

Our second important advance is that we show how the diel activity patterns of predators provide crucial insight to how prey species manage threats from multiple predators. These patterns clarified how it is possible for prey to be safe from multiple predator species simultaneously. Had we ignored diel predator activity, we would have concluded, incorrectly, that avoiding one predator necessarily increased exposure to the other. Recognizing that wolves and cougars hunted in different places and at different times revealed how elk could simultaneously minimize threats from both predators. Movement out of the forest and into the open to avoid cougars, for example, did not result in greater risk from wolves and vice versa because these predators were active at different times of day. Cougars were most active at night, whereas wolves were most active in the morning and at dusk (Fig. 4), and these temporal differences allowed elk to safely use open areas at night when cougars were a threat in the forest and wolves were little threat in the open (Fig 6a). Thus, two apparently conflicting predator-specific prey defenses (select open vegetation to avoid cougars vs. select forest cover to avoid wolves; Atwood et al. 2009) were actually compatible after accounting for diel predator activity.

Strong elk habitat selection for the night-open domain (Fig. 6a) and night-flat domain (Fig. 6b) provided the clearest evidence of this activity pattern.
support for our vacant hunting domain hypothesis. At night (c. 2300–0500 h), the most open and flat areas of the landscape were outside the high-risk (90th percentile) and low-risk (50th percentile) predator hunting domains. As night turned to day, elk selected for more forested and rugged areas outside the high-risk wolf domains. Although this included selection for the vacant day-forest domain, selection for vegetation openness was highest within the low-risk wolf domains. Similarly, selection for rough areas was concentrated at the edge of the high-risk wolf domain. Elk may have tolerated these areas because they corresponded to places where the probability of wolf predation was too low (< 0.05; Fig. 3) to warrant selection for denser forests and rougher topography. These patterns suggest that in some cases elk selected for hunting domains that were effectively vacant rather than totally vacant. This is consistent with an animal that seeks to minimize predation risk without sacrificing other critical activities, such as feeding (Lima 1999b).

Our findings are consistent with the ‘gap’ hypothesis (Fig. 1b) insofar as elk habitat selection was concentrated outside the high-risk predator domains. The ‘gaps’ in our system, however, were a function of temporal and spatial variation in predation risk, and we would have overlooked them had we ignored diel variation in predator activity. After accounting for this variation, we found no uniform support for the ‘lethality’ or ‘hunting mode’ hypotheses (Fig. 1c–d). Assuming wolves were more lethal and less predictable than cougars, elk selection for lightly forested areas within the cougar domain (Fig. 6a) was consistent with ‘lethality’ but not ‘hunting mode’, whereas elk selection for flatter areas within the wolf domain (Fig. 6b) was consistent with ‘hunting mode’ but not ‘lethality’.

Wolves were plausibly more lethal than cougars during our study given evidence from a concurrent study that radio-col lared female elk, including some from our study, were more likely killed by wolves than by cougars (Evans et al. 2006). On the other hand, snow-tracking and direct observation suggest that cougars may be more successful hunters than are wolves (Hornocker 1970; MacNulty et al. 2012). Ambiguous support for the hunting mode hypothesis may reflect how the cursorial mode of wolves was no less predictable than the spot-and-stalk/ambush mode of cougars in space (Fig. 3) or time (Fig. 4). Nevertheless, the overall stronger response of elk to cougars accords with the general notion that predator identity affects the strength of antipredator responses due to differences in functional traits of the predators (Schmitz et al. 2004; Preisser et al. 2007). Our results suggest that the effect of predator identity can arise from functional differences among predators unrelated to their spatiotemporal predictability.

Our findings are broadly relevant to ecology and conservation because they reveal a novel mechanism that may dampen the ecological effects of predators in at least three ways. First, a prey species that avoids one predator without increasing its risk to another reduces the potential for synergistic effects, where combined predators kill more prey than expected by their individual effects (also termed predator facilitation or risk enhancement; Sih et al. 1998; McCoy et al. 2012). Although conflicting prey defenses are not the sole cause of synergistic predation (Sih et al. 1998), our findings may help explain why synergistic predation is rare in natural systems (Schmitz 2007).

Second, prey that can simultaneously forage and be safe from multiple predators are less reliant on energetically costly antipredator defenses (e.g., vigilance, habitat shifts) that reduce survival and/or reproduction (termed non-consumptive or risk effects). In the elk we studied, high-body fat levels and correspondingly high-pregnancy rates (Cook et al. 2004; White et al. 2011; MacNulty et al. 2016; Kohl et al. 2018) were consistent with the hypothesis that selection for the vacant night-open and night-flat domains (Fig. 6a), which
included preferred grassland foraging habitats, was sufficient to offset the effects of wolves and cougars on over-winter fat loss and pregnancy rate. This may explain how elk in other populations maintained access to grasslands, and why they too maintained high levels of over-winter nutrition and/or pregnancy rate despite the presence of wolves and cougars (Hamlin et al. 2009; White et al. 2009; Middleton et al. 2013a).

Third, selection for vacant domains, which permits prey to forage in risky places at safe times, limits the capacity for predators to trigger behaviourally mediated trophic cascades. In our system, numerous studies suggest that the growth increase of some palatable woody deciduous plants (e.g., aspen, *Populus tremuloides*; willow, *Salix spp.*) is, at least partly, attributable to reduced herbivory from elk avoiding risky places where they might be killed by wolves (reviewed by Peterson et al. 2014; Beschta & Ripple 2016). Our results suggest that this scenario is unlikely because wolves were not the main predator affecting elk habitat selection, and neither wolves nor cougars excluded elk from risky places. On the contrary, elk maintained regular access to risky places during daily lulls in wolf and cougar activity. This may help explain why many aspen and willow have not escaped herbivory (Kauffman et al. 2010; Marshall et al. 2014).

The overarching insight from our study is that prey habitat selection in nature, where almost all prey species are exposed to many predator species distributed unevenly in space and time, is unlikely affected by only one predator in one dimension. The classic single-predator, space-only paradigm of spatial predator–prey interactions (Fig. 1a) has provided tremendous insights. But given its limits for making sense of a wild, speciose system like northern Yellowstone National Park, we encourage ecologists to pursue a multi-predator, spatiotemporal understanding of spatial predator–prey interactions (e.g., Fig. 2). Such an approach may offer new clarity about the causes and consequences of prey spatial response to predation risk.

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AUTHORSHIP

M.T.K. and D.R.M. conducted the analyses and wrote the manuscript. T.K.R., M.C.M., D.R.S., P.J.W., and D.W.S assisted with data collection and field logistics. All authors contributed substantially to revisions.

DATA AVAILABILITY STATEMENT

Data available from the Dryad Digital Repository: http://doi.org/10.5061/dryad.28d5v76.

REFERENCES


**SUPPORTING INFORMATION**

Additional supporting information may be found online in the Supporting Information section at the end of the article.

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Appendix S1. Figure 1. Winter (01 November – 30 April) locations of GPS-collared northern Yellowstone adult female elk \((n = 27)\) from 09 February 2001 to 31 March 2004. These data were used to analyze elk habitat selection in response to predation risk.
Appendix S1. Figure 2. Winter (01 November – 31 May) locations of northern Yellowstone adult female elk and calves killed by wolves during 2000-2004, n = 400 (a), and cougars during 1998-2006, n = 257 (b). We used these data to assess if vegetation openness and topographic roughness were valid indices of spatial variation in elk predation risk from wolves and cougars.
Appendix S1. Figure 3. Spatial distribution of topographic roughness (a) and vegetation openness (b) in northern Yellowstone, 2000-2004. Topographic roughness and vegetation openness were not highly correlated (Pearson’s correlation coefficient, $r = -0.18$). We used these data to index spatial variation in elk predation risk from wolves and cougars.
Appendix S1. Figure 4. Spatial distribution of forest canopy cover in northern Yellowstone during 2001. Data are from the LANDFIRE program and were used to estimate vegetation openness (see Appendix S1. Fig 3b).
Appendix S2

Supplementary information pertaining to study methodology and data collection.

Collar monitoring protocols

We evaluated elk habitat selection during winter (defined here as 01 November to 30 April; 6 months) to minimize potential bias associated with seasonal migration. If individuals arrived on the winter range after 1 November, we censored data to the individual’s arrival date (1-22 November). Elk GPS location data ( \( \bar{X} \pm SD \) location error = 6.15 ± 5.24 m; Forester et al. 2007) were censored to include only high-quality locations following guidelines developed by Forester et al. (2009). GPS collars collected data between 4 and 6 hour intervals (5 hour intervals: \( n = 23 \); alternating between 4 and 6 hour intervals: \( n = 4 \)).

For wolves, we maintained at least two radio-collars in all northern YNP wolf packs and recorded their locations approximately daily during two 30-day periods in early (mid-November to mid-December) and late (March) winter, when wolf packs were intensively monitored from the ground and fixed-wing aircraft, and approximately weekly during the rest of the year. GPS collars recorded locations every hour during the 30-day periods and at variable intervals outside these periods. For cougars, we located or monitored VHF- and GPS-collared cougars for mortality signals every 1-5 days from the ground and 7-14 days from the air. GPS collars recorded locations at either 3-hour fix intervals or recorded 8 locations per day skewed toward night and crepuscular hours (Ruth et al. 2019). Both VHF and GPS data was used to assess kill locations, whereas only GPS data were used to assess diel activity patterns (see below).

Study period

Our analysis of elk habitat selection used locations of GPS-collared elk collected during winter (as defined above) from 09 February 2001 to 31 March 2004 (\( n = 4 \) winters). Our analyses of predation risk from wolves and cougars also used data collected during winter. To satisfy the sample size requirements of our risk analyses, we extended the data collection window for predators to include a 7-month winter (1 November to 31 May) across 4-7 winters (Table S1). Analyses of the spatial probability of predation used data on wolf-killed elk from 4 winters (01 November 2000 to 31 May 2004), and data on cougar-killed elk from 7 winters (April 1998 to May 2005). Analyses of diel predator activity used cougar GPS data from 6 winters (February 2001 to November 2006) and wolf GPS data from 10 winters (2004-2013). The wolf GPS data do not overlap the winters when elk GPS data were collected because GPS collars were not fitted to wolves until 2004. Prior to 2004, only VHF (Very High Frequency) radio-collars were used to track Yellowstone wolves.

Animal handling

We captured and handled elk, wolves, and cougars following protocols in accord with applicable guidelines from the American Society of Mammalogists (Sikes & Gannon 2011) and approved by the Hornocker Wildlife Institute/Wildlife Conservation Society or National Park Service Institutional Animal Care and Use Committees. Elk were captured in February (2001-2003) via helicopter net-gunning (Hawkins and Powers, Greybull, Wyoming, USA; Leading Edge
Aviation, Lewiston, Idaho, USA) and fitted with Telonics (Telonics, Mesa, Arizona, USA) or Advanced Telemetry Systems Inc. (Isanti, Minnesota, USA) GPS radio-collars. Wolves were fitted with VHF (Telonics Inc., Mesa, AZ, USA) or GPS (Televilt, Lindesberg, Sweden; Lotek, Newmarket, ON, Canada) radio-collars. Cougars were fitted with VHF (Telonics Inc., Mesa, AZ, USA) or GPS (Telonics, Inc. and Televilt, Inc., Lindesberg, Sweden) radio-collars.

**Predation sampling**

**Wolf sampling:**
Wolf-killed elk (n = 400) were documented systematically during two 30-day periods of daily monitoring in early (mid-November to mid-December) and late (March) winter, and opportunistically outside these periods. Field personnel working on the ground and observers in fixed-wing aircraft documented kills by monitoring both VHF and GPS collared individuals. Poor weather occasionally precluded aerial observations.

Ground-based and aerial observers independently recorded the locations of wolf-killed elk using GPS or 1:24,000 scale topographic maps. Ground-based observers recorded locations when necropsying carcasses, whereas aerial observers recorded locations when flying directly over a kill. If more than one spatial location was available for a given kill, our criteria for inclusion in the analysis followed: ground GPS location > air GPS location > ground location from topographic maps. Our sample of wolf-killed elk were classified as “definitely” or “probably” killed by wolves (Mech et al. 2001).

**Cougar sampling:**
Cougar-killed elk (n =257) were documented opportunistically during surveys of predetermined sampling routes (55%) or systematically during a ‘predation sampling sequence’ of a randomly selected VHF or GPS radio-collared cougar (45%). All sampling sequences were searched for cougar-killed ungulates until at least 2 ungulate prey (elk, deer, or bighorn sheep) were identified. A VHF predation sampling sequence entailed triangulation on the radio-collared individual 1-3 times per day. A GPS predation sampling sequence involved analysis of GPS location clusters using ArcGIS 8.0 to identify potential kills. Location clusters consisting of at least 2 locations and located less than 200 m apart were selected as candidates for investigation (Ruth et al. 2010). Our sample of cougar-killed elk were classified as “definitely” or “probably” killed by cougars (Kunkel et al. 1999).

**Sampling protocol used to assign availability for spatial predation risk**

We analyzed wolf and cougar kill resource selection functions (RSF) using a generalized additive model (GAM) with a binomial distribution (used vs. available) and a logit-link function implemented in the mgcv package (version 1.8.0) of R. We used the default settings such that cross-validation determined the optimal amount of smoothing (Wood 2006). Used locations were the positions of kills and available locations were the positions of 10,000 randomly selected points from within northern Yellowstone. Each used and available location was assigned separate values of roughness and openness based on where they intersected these estimated spatial layers (Appendix S1. Fig 3). We estimated four separate kill RSFs, corresponding to one RSF for each combination of predator (wolf, cougar) and spatial metric (openness, roughness).
The fitted values of each model correspond to the predicted probability of a cougar (or wolf) killed elk as a function of openness (or roughness). Predicted kill probabilities for wolves and cougars were not directly comparable due to a large difference in the number of used locations for each predator (400 wolf kills vs. 257 cougar kills). To make them comparable, we rescaled the cougar kill RSF to account for how for the number of used cougar-kill locations represented 64.25% of the 400 used wolf-kill locations. Specifically, we estimated a cougar kill RSF that compared the 257 used cougar-kill locations to 6,425 locations randomly selected from the 10,000 available locations. This adjusted the magnitude of the predicted cougar kill probabilities to approximate that of the predicted wolf kill probabilities. Sample mean, standard deviation, and median were similar between the full and reduced cougar kill datasets, which indicated that our proportional sampling did not bias model coefficients.

**Modeling details for diel predator activity**

We separately fitted a Generalized Additive Mixed Model (GAMM) to step-lengths of wolves and cougars using the negative binomial family and a cyclic cubic regression spline so that the first and last hour of the day matched in accordance with the diel cycle (Kohl et al. 2018). We included a random intercept for individual identity to account for repeated measurements of individuals across the study. We used the mgcv package (version 1.8.0) in R to estimate each GAMM. Each wolf was considered an independent measure of movement rate because it was solitary, the only GPS radio-collared wolf in a pack, or rarely associated with other GPS-collar pack members (Kohl et al. 2018). Our estimated wolf diel pattern was most likely representative of the wolf diel pattern prior to 2004 because it was (1) correlated with the time of day that we directly observed wolves encountering \( r = 0.79 \) and killing \( r = 0.87 \) elk prior to 2004; (2) consistent across the years in which it was measured; and (3) similar to diel patterns described for other wolf populations (Kohl et al. 2018). Each cougar was also considered an independent measure of movement because they were mostly solitary (Ruth et al. 2019). We included a fixed effect of cougar sex and a sex-specific smoother in our GAMM of cougar diel activity because it was a better fit to the data than a model that ignored cougar sex \( \Delta AIC = 8.833 \). Our estimated diel patterns for male and female cougars were consistent with those described for other cougar populations (e.g., Beier et al. 1995; Sweanor et al. 2008; Wang et al. 2015).

**Modeling elk habitat selection**

We used a step-selection function (SSF; where each observed step was compared to a sample of available steps that originated from the same starting location) to estimate the parameters of a resource selection function, \( \exp(Z' \beta) \), where \( Z \) is a vector of environmental covariates, and \( \beta \) is a vector of selection coefficients (Forester et al. 2009; Merkle et al. 2017). The SSF describes the relative probability of a movement step, \( p \), which is the straight-line segment between successive locations at 5-hour intervals. Movement steps with a higher score \( (p) \) relative to the set of possible steps have higher odds of being chosen by an animal (Fortin et al. 2005; Kohl et al. 2018). We could not estimate the main effect of wolf or cougar diel movement rate because neither varied within a stratum owing to how used and available locations within a stratum share the same point in time. We did not include a distance function in our SSF model because the combination of an empirical distribution with linear splines contributed to an appreciable reduction in parameter estimate bias compared to that in other SSF designs (Forester et al. 2009).
Furthermore, empirical sampling produces relatively unbiased estimates when parameter estimates are as small ($\beta < 1.0$; Forester et al. 2009) as they were in our study (see Appendix S7).

We only tested thresholds that occurred within 1 – 99% of all used points for each spatial risk index in order to minimize the influence of outliers. The precision (i.e., decimal units) and scale (i.e., difference in minimum/maximum values) of the spatial risk index determined the number of candidate spline models (roughness = 99 models; openness = 269 models). We did not test for behavioral lags or leads (sensu Kohl et al. 2018) because in this case it required an excessive number of potentially meaningless candidate models that could produce spurious results (Johnson & Omland 2004).

To evaluate candidate models, we used the quasi-likelihood under independence criteria (QIC; Pan 2001) because it considers independent clusters of observations while also accounting for non-independence between subsequent observations (Craiu et al. 2008). We have shown previously that results from our elk habitat selection model do not differ between models fitted to all clusters and models fitted to every other independent cluster ($n = 2$ independent data sets; Kohl et al. 2018). We therefore present results from the analysis of all clusters to maximize sample size. We performed SSF analyses and $k$-folds cross validations (see main text) in R using the SURVIVAL and HAB packages, respectively.

We recognize that reproductive status (e.g., calf at heel) may influence the anti-predator responses of ungulates including elk (e.g., Laundré et al. 2001). We were unable to control for this potential effect in our study because the annual reproductive status of our sample of radio-collared elk was not known. Nevertheless, we doubt that reproductive status was an important source of variation in our analysis because cow-calf ratios were low (< 16 calves per 100 cows [2001-2004]; Proffitt et al. 2014) during our study. This suggests relatively few of our radio-collared elk had a calf at heel, which minimizes the possibility that uncontrolled variation in reproductive status biased our results.

We assessed multicollinearity among covariates using variance inflation factors with the R package “HH”. We examined the VIF scores of covariates in ‘single-predator space x activity’ models that included a linear effect of spatial risk (openness or roughness). We did not assess VIF scores of covariates in models that included linear splines for spatial risk because variation in knot location led to incomparable VIF scores. The VIF scores for the roughness x female cougar activity model and the roughness x wolf activity model exceeded 10 (Table S2a). Such a high score indicates that female cougar activity and wolf activity were highly collinear. We therefore excluded from consideration any elk habitat selection model that included the activity of female cougars and wolves.

**Visualizing elk selection for vacant hunting domains**

We visualized elk selection for vacant hunting domains in three steps. First, we combined our models of spatial kill probability and diel predator activity to describe the hunting domains of wolves and cougars with respect to openness and roughness. We represented hunting domains as the top five percentiles of fitted values from the predator-specific models of spatial kill probability (Fig. 3) and diel activity (Fig. 4b). The 90th percentile equals the top 10% of fitted
values, which indicate where the probability of a kill and when the level of predator activity were highest. For example, the 90th percentile cougar hunting domain with respect to vegetation openness (Fig. 6a) corresponds to values of openness (e.g., 164 – 192; Fig. 3a) and time of day (e.g., 0200 - 0415 hours; Fig. 4b) where kill probability and diel activity were highest. For each set of fitted values, we identified the upper and lower values of each percentile (50th to 90th). This provided two pairs of values corresponding to the upper and lower predicted spatial kill probability and diel activity level for each of the five percentiles. We calculated ellipses around each set of two pairs to illustrate the spatiotemporal hunting domain of each predator. The smallest ellipse equals the 90th percentile, and the largest ellipse equals the 50th percentile. We refer to these as the ‘high-risk’ and ‘low-risk’ zones of the predator hunting domain.

Second, we illustrated elk habitat selection with respect to space and time of day. To do so, we used our best models of elk habitat selection (Appendix S7) to predict the probability of selection at each integer value of openness (or roughness; Fig. 3a, 3b in Appendix S1) at each hour-specific level of average predator-specific activity (wolves and male cougars; Fig. 4b). This approach is sufficient for producing SSF-based maps because our elk GPS sampling rate was low and our spatial metrics (openness and roughness) were relatively simple (Singer et al. 2017). Next, we rescaled the fitted values to range from 0 to 1 for each hour of the day. We did this by (i) subtracting the minimum fitted value for a given hour from each fitted value for that hour, and (ii) dividing each difference by the maximum difference. We smoothed these rescaled fitted values using the default parameters of the stat_contour function in the ggplot2 package of R version 3.3.3.

Third, we overlaid the predator-specific hunting domains (i.e., ellipses) on the elk habitat selection illustration. This yielded a representation of elk habitat selection relative to predator hunting domains (Fig. 6) that was comparable to the vacant domain hypothesis illustrated in Figure. 2. We truncated roughness to a value of 150 because relatively few kills and elk locations occurred beyond it. Specifically, 1% of 657 kills extended to a maximum roughness of 198, and 0.13% of 10,199 elk locations extended to a maximum roughness of 209. By contrast, kills and elk locations were distributed across the entire range of openness.

We emphasize that Figure 6 does not represent habitat selection across our entire study area, but rather demonstrates the relative probability of use assuming that all values of openness and roughness were available to an animal at a given step. To demonstrate the validity of this assumption, we first quantified the distance an elk could travel within a 5-hour time step (Appendix S2. Fig. S1). Next, we identified locations within our study area that contained the full range of openness and roughness values within the typical distance of a 5-hour time step (Appendix S2. Fig. S2, S3). We found that 95% of all step lengths were < 1,637 m which was a sufficient distance for elk to plausibly use any value of openness or roughness illustrated in Figure 6.
Appendix S2. Table S1. Timeline of data collection periods for elk, cougars, and wolves in northern Yellowstone National Park and adjacent areas of Montana. Winter is defined as 1 November to 30 April for elk and 1 November to 31 May for cougars and wolves (see Study Period above).

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Appendix S2. Table S2. Variance Inflation Factor scores for the activity of male cougars (M), female cougars (F), and wolves (W) in elk habitat selection models that included a linear effect of roughness (R) or openness (O). This analysis considered ‘single-predator space x activity’ models (R/O x M/F/W).

<table>
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<tr>
<th></th>
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<th>R x W</th>
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<table>
<thead>
<tr>
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<th>O x F</th>
<th>O x W</th>
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<td></td>
<td>2.2</td>
<td>3.3</td>
<td>4.1</td>
</tr>
</tbody>
</table>
Appendix S2. Figure S1. Distribution of step lengths for 27 radio-collared elk relocated every 5 hours in northern Yellowstone during winter (2000-2004). Ninety-five percent of all step lengths were less than 1,637 m (identified in red).
Appendix S2. Figure S2. Areas within northern Yellowstone that contain the full range of openness values (black to white) within a distance potentially transversed by elk within a 5-hour step length (see Appendix S2. Fig. S1) during winter (2000-2004).
Appendix S2. Figure S3. Areas within northern Yellowstone that contain the full range of roughness (black to white) values within a distance potentially transversed by elk within a 5-hour step length (see Appendix S2. Fig. S1) during winter (2000-2004).
References


Appendix S3. Spatial distribution of elk (adult females and calves) killed by male and female cougars (1998-2006) during winter in northern Yellowstone relative to vegetation openness (a) and topographic roughness (b). Lines are fitted values with 95% confidence intervals (shaded areas) from four separate generalized additive models (Appendix S2).
Appendix S4. Diel movement rate of GPS-collared cougars and wolves in northern Yellowstone during winter. Symbols are observed step lengths and lines are fitted values from generalized additive mixed models. Panel (a) shows the diel movement rate of 2 female and 3 male cougars based on a step-length of 3 hours, which was the finest temporal resolution of our cougar data. Panel (b) shows the mean diel movement rates of female and male cougars estimated from a GAMM fitted to the data in panel (a). The patterns in panel (b), which covers 21 hours of the diel cycle, compares reasonably well with the diel movement rate of male cougars we used in our elk habitat selection analysis (Fig. 4 - main text), which was based on a step-length of 6 hours for 2 female and 6 male cougars, panel (c). One male cougar was excluded from the 3-hr analysis due to an incompatible fix schedule. Panel (d) shows the 6-hr step lengths of 21 GPS-collared wolves. All step-lengths were jittered to enhance visualization.
c. [Graph showing cougar movement over time of day.]

- Step length (m)
- Mean diel activity (m)
- Step length (f)
- Mean diel activity (f)

Time of day

Cougar movement (km/6 hr)

Diel activity

0 5 10 15 20

0 2 4 6 8

d. [Graph showing wolf movement over time of day.]

- Step length
- Mean diel activity

Time of day

Wolf movement (km/5 hr)

Diel activity

0 5 10 15 20

0 5 10 15
Model selection results for a step-selection function describing the interactive effect of spatial risk (openness) and diel predator activity of male cougars (M; km travelled/6-hr), female cougars (F; km travelled/6-hr), and/or wolves (W; km travelled/5-hr) on elk habitat selection in northern Yellowstone, 2001-2004. Variable ‘Open’ represents a simple linear effect of openness. Variables O1 and O2 contain a linear spline for openness at the indicated threshold. Each spline model is the top model from a grid search of thresholds. For example, model set O1 + O2 + (O1 x W) + (O2 x W) with a threshold at 288 was the top model from 269 possible models with a threshold from 20 to 288. Number of parameters (K), QIC, and differences in QIC compared to the best model (ΔQIC) are given for each model. The best-fit model is identified in bold.

<table>
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<th>ΔQIC</th>
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Appendix S6. Model selection results for a step-selection function describing the interactive effect of spatial risk (roughness) and diel predator activity of male cougars (M; km travelled/6-hr), female cougars (F; km travelled/6-hr), and/or wolves (W; km travelled/5-hr) on elk habitat selection in northern Yellowstone, 2001-2004. Variable ‘Rough’ represents a simple linear effect of roughness. Variables R1 and R2 contain a linear spline for roughness at the indicated threshold. Each spline model is the top model from a grid search of thresholds. For example, model set R1 + R2 + (R1 x W) + (R2 x W) with a threshold at 29 was the top model from 99 possible models with a threshold from 1 to 99. Number of parameters (K), QIC, and differences in QIC compared to the best scoring model (ΔQIC) are given for each model. The best-fit model is identified in bold.

<table>
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Appendix S7. Coefficient estimates for best-fit step-selection function models describing the interactive effects of spatial risk (openness [a] and topographic roughness [b]) and diel activity of male cougars (C; km/6-hrs) and wolves (W; km/5-hrs) on elk habitat selection in northern Yellowstone, 2001-2004. Variables O1 and O2, and R1 and R2 are the slopes before and after each index-specific threshold. Model selection results are presented in Appendix S5 and Appendix S6. Confidence intervals were computed using robust standard errors.

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