Habitat Relations

Crucial Nesting Habitat for Gunnison Sage-Grouse: A Spatially Explicit Hierarchical Approach

CAMERON L. ALDRIDGE,1 Department of Ecosystem Science and Sustainability, NREL, Colorado State University, in cooperation with U.S. Geological Survey, 2150 Centre Avenue, Building C, Fort Collins, CO 80526-8118, USA

D. JOANNE SAHER, NREL, Colorado State University, in cooperation with U.S. Geological Survey, 2150 Centre Avenue, Building C, Fort Collins, CO 80526-8118, USA

THERESA M. CHILDERS, National Park Service, Black Canyon of the Gunnison National Park and Curecanti National Recreation, 102 Elk Creek, Gunnison, CO 81230, USA

KENNETH E. STAHLNECKER, National Park Service, Black Canyon of the Gunnison National Park and Curecanti National Recreation, 102 Elk Creek, Gunnison, CO 81230, USA

ZACHARY H. BOWEN, U.S. Geological Survey, 2150 Centre Avenue, Building C, Fort Collins, CO 80526-8118, USA

ABSTRACT Gunnison sage-grouse (Centrocercus minimus) is a species of special concern and is currently considered a candidate species under Endangered Species Act. Careful management is therefore required to ensure that suitable habitat is maintained, particularly because much of the species’ current distribution is faced with exurban development pressures. We assessed hierarchical nest site selection patterns of Gunnison sage-grouse inhabiting the western portion of the Gunnison Basin, Colorado, USA, at multiple spatial scales, using logistic regression-based resource selection functions. Models were selected using Akaike Information Criterion corrected for small sample sizes (AICc) and predictive surfaces were generated using model averaged relative probabilities. Landscape-scale factors that had the most influence on nest site selection included the proportion of sagebrush cover >5%, mean productivity, and density of 2 wheel-drive roads. The landscape-scale predictive surface captured 97% of known Gunnison sage-grouse nests within the top 5 of 10 prediction bins, implicating 57% of the basin as crucial nesting habitat. Crucial habitat identified by the landscape model was used to define the extent for patch-scale modeling efforts. Patch-scale variables that had the greatest influence on nest site selection were the proportion of big sagebrush cover >10%, distance to residential development, distance to high volume paved roads, and mean productivity. This model accurately predicted independent nest locations. The unique hierarchical structure of our models more accurately captures the nested nature of habitat selection, and allowed for increased discrimination within larger landscapes of suitable habitat. We extrapolated the landscape-scale model to the entire Gunnison Basin because of conservation concerns for this species. We believe this predictive surface is a valuable tool which can be incorporated into land use and conservation planning as well the assessment of future land-use scenarios. © 2011 The Wildlife Society.

KEY WORDS Centrocercus minimus, Colorado, extrapolation, Gunnison sage-grouse, habitat, hierarchical model, nesting, resource selection, sagebrush.

To properly manage habitats for wildlife species of concern, knowledge about resource requirements is necessary. This information can inform management activities, lead to the provision of suitable habitat across life stages, and ensure long-term persistence for the species (Boyce and McDonald 1999, Morrison 2001, Aldridge and Boyce 2007). Although the abundance of high quality habitats is essential for species persistence, increasingly, the spatial distribution of those resources and continued disturbance from anthropogenic developments have been recognized as important drivers affecting habitat quality (Bock and Jones 2004, Doherty et al. 2008), animal fitness (Aldridge and Boyle 2007), and population persistence (Akçakaya et al. 2004, Haines et al. 2006). Characterizing the response of animals to anthropogenic disturbances (i.e., behavioral avoidance), and resultant effects (direct or indirect) on fitness components within a spatial framework, ultimately affords one the ability to model and map critical habitat requirements (see Aldridge and Boyle 2007). Maps based on models developed from empirical data assessing wildlife-habitat associations can provide managers with the tools necessary to make informed management decisions about where and what types of
management actions (i.e., protection, enhancement, or rehabilitation) might be required (Haines et al. 2006, Aldridge and Boyce 2007).

Capturing the correct spatial and temporal scales at which organisms operate has been recognized as one of the fundamental problems in ecology (Johnson 1980, Morris 1987, Levin 1992). The need to embrace and understand the influence of spatial scale (grain and extent) on habitat selection is central to understanding habitat requirements for species (Wiens 1989; Orians and Wittenberger 1991; Boyce 2006; Aldridge and Boyce 2007, 2008). Intuitively, selection is hierarchical (Rettie and Messier 2000), as originally proposed by Johnson (1980), when he described the concepts of nested orders of selection. Here, we define hierarchical selection as the processes where finer-scale (referring to extent only) habitat selection (i.e., nest or den site selection) occurs within the context of a priori selection for landscape-scale resources (i.e., annual or lifetime home range selection). Although hierarchical selection has rarely been assessed (but see Harvey and Weatherhead 2006), it is recognized as a potentially important constraint that should be considered with habitat selection assessments (Johnson 1980, Rettie and Messier 2000, Mayor et al. 2009).

Gunnison sage-grouse (Centrocercus minimus) is considered a species of special concern in Colorado and a sensitive species in Utah, the only 2 states where the species occurs. The International Union for Conservation of Nature (IUCN) includes the Gunnison sage-grouse on their Red List, considering the species endangered (IUCN 2010). Approximately 4,023 birds remain across 7 populations, with approximately 87% residing within the Gunnison Basin population (Fig. 1; A. Pfister, U.S. Fish and Wildlife Service, personal communication). Continued loss and fragmentation of sagebrush habitat from urban development and associated infrastructure, as well as direct conversion of habitats, threatens populations and limits connectivity (Oyler-McCance et al. 2001, Gunnison Sage-grouse Rangewide Steering Committee [GSGRSC] 2005). Human populations within the Gunnison Basin have been increasing since 1980, with projections estimating more than a 2-fold (2.3 times) increase in the number of people in the Gunnison River Basin by 2050 (Colorado Water Conservation Board 2009). Existing and future threats coupled with small and/or declining populations of Gunnison sage-grouse (GSGRSC 2005) contributed to the species being listed in September 2010, as a candidate species under the Endangered Species Act of 1973 (ESA; 65 Federal Register 82310).

Spatial models assessing habitat selection have been developed for greater sage-grouse (C. urophasianus; Doherty et al. 2008, Yost et al. 2008) and several studies have evaluated and identified the importance of selection at multiple spatial scales, across various life stages (Aldridge and Boyce 2007, 2008; Aldridge et al. 2008; Kolada et al. 2009; Carpenter et al. 2010; Doherty et al. 2010). To date, however, no spatial habitat assessments have been conducted for the closely related Gunnison sage-grouse, despite concern over the species’ future viability. The goal of our study was to develop spatially explicit nesting habitat models for Gunnison sage-grouse. We define crucial habitat as resources necessary for the survival and long-term viability of Gunnison sage-grouse. Specifically, our objectives were to: 1) develop landscape models that identify crucial nesting habitat across large spatial extents; 2) develop patch models that, after being constrained by landscape-scale selection (objective 1), identify crucial nesting habitat patches within larger nesting areas; 3) spatially apply both models to our study area, allowing us to map crucial nesting habitat for Gunnison sage-grouse, integrating both landscape- and patch-level selection; 4) assess the predictive capacity of these models using independent data; and 5) apply the landscape model to a broader range extent within the Gunnison Basin and evaluate model generality using lek (communal breeding grounds) data and the potential to use these models for direct management and conservation purposes for this population, and the species in general.

STUDY AREA

The Gunnison Basin is a high-elevation valley on the eastern edge of the Colorado Plateau. The study area was approximately 84,500 ha, of which, 49,200 ha (approx. 58%) were public lands. Elevation within the study area ranges from approximately 2,180 m to 3,100 m above sea level. The Gunnison Basin had an average annual temperature of 3.1°C and an average annual precipitation of 27 cm. Gunnison sage-grouse range within the Gunnison Basin was estimated at 240,000 ha, of which 70% were public lands (GSGRSC 2005). Gunnison sage-grouse nest location data were collected in the western portion of the Gunnison Basin, focused on the Curecanti National Recreation Area managed by the National Park Service (NPS), and its surrounding area (Fig. 1). Sagebrush-steppe was prevalent throughout the basin, with big sagebrush (Artemisia tridentata) dominant. Intermixed within the sagebrush community were riparian areas and drainages containing Narrowleaf Cottonwood (Populus angustifolia), juniper (Juniperus scopolorum), gamble oak (Quercus gambelii), serviceberry (Amelanchier alnifolia), and wild rose (Rosa woodsii). Most of the valley bottoms along major drainages had been converted to hay and pastureland. Ponderosa pine (Pinus ponderosa), Douglas fir (Pseudotsuga menziesii), spruce (Picea spp.), and aspen (Populus tremuloides) forests were found at higher elevations.

METHODS

Field Sampling
We used spotlighting techniques (Giesen et al. 1982) to capture female Gunnison sage-grouse from 7 of 10 known active leks within the study area from 2000 to 2009. We fit captured hens with a necklace-style radio transmitter (RI-2BM, Holohil Systems, Carp, Ontario, Canada; A4050, Advanced Telemetry Systems, Isanti, MN) and tracked the hens throughout the breeding season using radio telemetry to identify nesting locations. We estimated nest
locations using triangulation vectors processed using LOCATE II (Locate II Version 1.82, http://nsac.ca/envsci/staff/vnams/Locate.htm, accessed 14 Feb 2001). We checked nest status regularly (minimum twice per week) via telemetry from an inconspicuously marked location (rock cairn or similar natural marker) 20 m from the estimated nest location. Once the nest was vacated, we recorded the Global Position System (GPS) coordinates of the nest site. We post-processed locations using Trimble Pathfinder Office software (Trimble Navigation Limited, Sunnyvale, CA) to increase spatial location accuracy (approx. 0.5 m). All animal handling methods and protocols were approved by the U.S. Geological Survey (USGS) Fort Collins Science Center’s Institutional Animal Care and Use Committee.

Spatial Predictor Variables
We developed spatial variables thought to be important predictors of Gunnison sage-grouse nest site selection, based on resource selection studies for greater sage-grouse, and to a lesser extent, for Gunnison sage-grouse. We grouped these variables into 8 categories, 1) shrub, 2) other vegetation, 3) conifer, 4) terrain, 5) residential, 6) roads, 7) water, and 8) vegetation indices, which are detailed below. Greater sage-grouse select habitat across spatial scales (Aldridge and Boyce 2007, 2008; Doherty et al. 2008, 2010) and recent research has shown that anthropogenic disturbances can affect greater sage-grouse habitat selection (Aldridge and Boyce 2007; Doherty et al. 2008, 2010), lek dynamics (Walker et al. 2007, Holloran and Anderson 2005, Harju et al. 2010), individual movements and nesting activities (Lyon and Anderson 2003, Holloran et al. 2010), and fitness components (Aldridge and Boyce 2007, Holloran et al. 2010), all of which occur at varying spatial extents. Thus, we evaluated spatial variables affecting Gunnison sage-grouse nest site selection over 6 different spatial circular moving window sizes, all of which had some support within the sage-grouse literature. Landscape scales included the window extents of 1-km, 1.5-km, 3-km, and 6.4-km radii (see Table S1, available online at www.onlinelibrary.wiley.com). For patch models, we assessed metrics across 2 window extents; 0.045-km and 0.564-km radii, as well as at the pixel level for some variables (30 m; Table S1). In this case, a moving window is a geospatial calculation where the value of a given pixel is generated based on the values of all surrounding pixels within a specified spatial extent.

We derived shrub variables from recently developed spatial products estimating the percent cover of shrubs (all species), sagebrush (Artemisia spp.; all species combined), all big sagebrush (all subspecies combined), and Wyoming sagebrush (A. t. wyomingensis; see Homer et al. 2008, 2012). Importantly, these cover estimates are not directly proportional to typical vegetation cover estimates measured at much smaller spatial scales using quadrats or measuring tapes on the ground (see Homer et al. 2008, 2012). From these products, we calculated the mean estimated percent cover and the standard deviation (SD) over all window extents. We used SD as a surrogate for habitat heterogeneity. We also calculated the proportion of each window that was estimated to have 1–5%, 6–10%, 11–15%, 16–20%, and 21–25% cover, as well as greater than 5%, 10%, 15%, or 20% cover for each shrub variable. The category break points align with the
estimated error associated with the original mapping products (approx. 5%; Homer et al. 2008, 2012). We also calculated the mean estimated percent cover and associated SD of bare ground, herbaceous vegetation, litter, and shrub height (Homer et al. 2008, 2012) over all window extents (Table S1).

We also calculated the mean and associated SDs across window scales for some terrain and vegetation indices variables. Terrain variables included Compound Topographic Index (CTI) and Terrain Ruggedness Index (TRI). The CTI is a function of both slope and upstream contributing area and can be used as a surrogate for soil moisture and vegetation productivity (Gessler et al. 1995). We based our TRI on the vector ruggedness measure (VRM) developed by Sappington et al. (2007). Low ruggedness values indicate flat areas (low slope), moderate values reflect steep but even terrain (high slope, low ruggedness), and high ruggedness values indentify areas that are steep and uneven (high ruggedness). Computational complexities prevented generation of these variables at some larger scales (see Table S1). We only assessed SD (heterogeneity) for CTI, given that TRI directly models terrain heterogeneity. We assessed vegetation indices over all scales including brightness, greenness, wetness, and Normalized Difference Vegetation Index (NDVI) indices using a tasseled cap transformation (see Table S1).

We calculated density and distance metrics for a number of potentially important drivers of Gunnison sage-grouse habitat use. We used Landfire products (Landfire 2006, Version LF_1.1.0, http://www.landfire.gov/, accessed 02 Oct 2009) to assess potential influence of conifer habitats on nesting habitat selection. We derived the density of conifer and conifer with pinyon (Pinus spp.) and juniper (Juniperus spp.) classed habitats over the suite of moving windows. Similarly, we developed density metrics across all scales to assess the influence of residential development on habitat selection and understand the scale of the effect. We defined residential development as all areas within the Gunnison County, any home site visible from 2005 NAIP imagery, or any other visible development (2005 NAIP 1-m imagery) occurring outside subdivision boundaries but where the county had recorded address locations. We modified address points manually to align with observed developments on the NAIP imagery and then buffered by 0.075 km. Finally, we developed road density (linear km/km²) estimates across all spatial window extents (Table S1). We used a recently developed Bureau of Land Management (BLM) road classification based on 2005 NAIP, aided by local knowledge of road use. Road products were collaboratively developed by BLM, U.S. Forest Service, NPS, Gunnison County, and USGS across the Gunnison sage-grouse range within the Gunnison Basin.

We assessed selection in relation to proximity of conifer, roads, residential habitat, and water using straight line (Euclidean) distance, as well as quadratic terms and exponential decays as a function of Euclidean distance (Nielsen et al. 2009) to assess nonlinearities. We developed 5 decay variables using the form $e^{-d^a}$ where $d$ was the distance in meters from each pixel to a landscape feature, and $a$ was set at 50, 100, 250, 500, and 1,000. This scaled each distance variable between 0 and 1, with highest values close to the feature of interest, capturing potential nonlinear responses from birds to a given landscape factor as the distance from that feature increases. We considered distance metrics in both landscape- and patch-scale models (see Table S1 for a complete list of the variables assessed).

Model Development

We assessed nest site selection hierarchically, first developing a landscape model using information from across the study area, and then assessing patch selection by restricting availability of resources using a threshold defined by the landscape model predictive surface, identifying habitats available for selection at the patch scale. We developed resource selection function (RSF; Manly et al. 2002) models using logistic regression (Hosmer and Lemeshow 2000) to characterize nesting habitat for Gunnison sage-grouse. We generated random locations across our study area at a density of 5 per km² ($n = 3,392$) to represent the variation in habitat availability, and compared these with nest locations to produce what we termed our landscape models. We did not consider the Blue Mesa Reservoir as available habitat and restricted available habitat (and thus model prediction space) to the distribution of spatial data from all sources. For instance, the sagebrush map products in this area were not developed above 3,100 m elevation (see Homer et al. 2012) and we did not consider these areas. Since models were heavily biased toward the larger sample of available resource units, we used an importance weight that gave full weighting to used resource units but down weighted available resource units proportional to the ratio of sampled points to available points (STATACORP 2007, Aldridge and Boyce 2007).

Model structure followed the form:

$$w(x) = \exp(\beta_1 x_1 + \beta_2 x_2 + \ldots \beta_k x_k)$$

where $w(x)$ is the relative probability of selection for the predictor variables, $x_i$, and $\beta_i$'s are the coefficient estimate for each predictor variable $(1 \ldots k)$.

We defined availability for the patch model by a threshold classification (geometric binning) from the landscape model defining crucial nesting areas for Gunnison sage-grouse. We based this classification on a threshold that maximized the inclusion of nest locations. We used a subset of the landscape-scale available points reflecting the reduction in spatial extent for patch-scale modeling. We developed landscape models first and subsequently developed patch models. We felt it was inappropriate to assess individual variables without some a priori consideration of habitat, given the species’ affinity for sagebrush. Therefore, we initially assessed shrub metrics (i.e., shrub or sagebrush cover, etc.) as individual predictors of Gunnison sage-grouse nest occurrence for both landscape and patch models. When graphical inspection of data indicated a nonlinear variable form would more accurately represent the data, we assessed both linear and
quadratic forms. We also evaluated whether the inclusion of the SD of the same spatial extent window explained additional variation. The most predictive metric (form and window extent) based on Akaike's Information Criterion corrected for small sample sizes (AIC$_c$; Burnham and Anderson 2002) was carried forward as the base model and was maintained throughout the assessment of other variables and candidate models at each scale.

**Variable selection.**—The number of variables considered for inclusion in models predicting the occurrence of Gunnison sage-grouse nests was large (see Table S1). To ensure balanced participation by variables belonging to each category, we first assessed variables within a category. We only considered the top performing variable(s) from each category for the development of the final candidate set of models. We evaluated each variable in combination with the shrub base using AIC$_c$. Within each category, we identified the most predictive window extent and form of each variable type (i.e., distance or density), as described above. If more than 1 variable was selected from a category, we again assessed variables for correlation (Pearson’s $r > [0.7]$) to prevent multicollinearity issues (Aldridge and Boyce 2007). When correlated, variables carried forward had the lowest AIC$_c$ score (most predictive) and intuitive biological interpretation. We compared the change in AIC$_c$ from the base shrub model to each simple model (variable of interest plus base) to assess the relative contribution for each variable. We carried forward selected variables (extent and form) from each category to develop a final candidate set of models. Because of small sample sizes, we limited candidate models to 7 variables, or roughly 1 variable per 10 observations (Hosmer and Lemeshow 2000). All possible combinations of uncorrelated variables that met this criterion were included in the final candidate set of models. We ranked candidate models based on the difference in AIC$_c$, values ($\Delta$ AIC$_c$) and we used Akaike weights ($w_i$) to assess the strength of evidence that a particular model was the best of those in the candidate set. We repeated this process for patch models. We conducted all analyses in STATA 10.1 (STATACORP 2007).

**Multi-model inference and spatial prediction.**—At both the landscape and patch scales, we model averaged over the 90% confidence set to produce more robust spatial predictions and strengthen inference (Burnham and Anderson 2002). The estimated relative probabilities generated by individual models were scaled between 0 and 1 incorporating the estimated $\beta_0$. We re-calculated weights to sum to 1 for models within the 90% confidence set, and applied these adjusted weights to spatial predictions for each model, because model weights ($w_i$) apply to models as a whole, not their individual components (Candolo et al. 2003, Murray and Conner 2009). We added weighted predictions together in a Geographic Information System (GIS) to produce the final model averaged relative probability surface.

Greater sage-grouse nest-site selection is influenced by both landscape- and patch-level factors (Aldridge and Boyce 2007). To capture this for Gunnison sage-grouse, we multiplied the model averaged continuous relative probability surfaces for patch and landscape scales within the NPS study area, but also provide individual surfaces for visual assessment. We geometrically binned all surfaces for display purposes and to allow for summarization of classified habitats.

**Model Assessment and Generalization**

**Model evaluation.**—Various approaches have been suggested for evaluating RSFs (see Johnson et al. 2006 for a review), but most necessitate subjective binning of the RSF predictions. We used Kendall’s c (Harrell et al. 1996) to describe the probability of concordance between model predictions (RSF relative probabilities) and observed data (nest and available locations; Harrell et al. 1996, Harrell 2001), preventing the need to bin data. In our case, Kendall’s c is an assessment of a model’s ability to discriminate between selected nest site locations and available habitats (Newson 2006). We consider probabilities of concordance $>0.9$ to be excellent, 0.8–0.9 to be good, and 0.7–0.8 to be fair. We assessed model discrimination for both training and testing nest datasets, contrasting nests against the available dataset used to build models, and did so for each modeling scale (patch and landscape). Using the geometric binning process described for the landscape model above, we summarized the proportion of nests that would be captured and the landscape that would be implicated with a given threshold. We followed a similar summary approach for the patch model.

**Model generalization.**—Since we did not have nest locations basin-wide, we used all known active lek locations in 2007 (Colorado Division of Wildlife, unpublished data) to evaluate the extrapolated landscape-scale model. Lek sites are centers of breeding activities, and their status and population demography are linked to proximity of high-quality nesting habitat (Holloran and Anderson 2005, Doherty et al. 2010). Lek location is strongly influenced by the amount of and proximity to quality nesting habitat (Connelly et al. 2000). If the landscape model is predictive, lek locations should have higher than average predicted nest occurrence, capturing nesting habitat surrounding the lek (landscape variables ranged in scale from 1- to 6.4-km radii windows). We used Kendall’s c to assess how well our landscape model could predict lek locations, contrasting lek locations to a sample of available locations from across the Gunnison Basin.

**Threshold response.**—We assessed nesting habitat selection responses relative to changes in sagebrush habitat and key anthropogenic variables, by plotting the model averaged nest occurrence probabilities against the variable of interest. We calculated the mean probability of occurrence across intervals of the identified predictor variables to summarize predictions for all pixels in the landscape (see Hanser et al. 2011). We developed threshold response curves for landscape-scale crucial nest areas across the range of values available throughout the NPS study area, and assessed the patch-scale crucial nesting model within the modified patch-scale study extent.
Table 1. A comparison of variables that were carried forward to model the occurrence of Gunnison sage-grouse nests (2006–2009; \( n = 73 \)) in the Gunnison Basin, Colorado at either the landscape or patch scale. We evaluated variables with the top shrub variable at each scale (landscape or patch base models), as indicated by the number of parameters in the model (including the shrub base model). \( \Delta \text{AIC} \) indicates the change in Akaike Information Criterion corrected for small sample size (AICc) value from either the landscape or patch base.

<table>
<thead>
<tr>
<th>Category</th>
<th>Variable</th>
<th>Definition</th>
<th>( K )</th>
<th>( \Delta \text{AIC} )</th>
</tr>
</thead>
<tbody>
<tr>
<td>Landscape variables—null model AIC = 204.393</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>Shrub</td>
<td>Proportion of pixels with &gt;5% sagebrush cover across a 1.5-km radius window</td>
<td>2</td>
<td>159.822</td>
</tr>
<tr>
<td></td>
<td>Other veg</td>
<td>Mean % ground cover (quadratic) across a 1.5-km radius window and standard deviation</td>
<td>5</td>
<td>153.538</td>
</tr>
<tr>
<td></td>
<td>NDVI</td>
<td>Mean NDVI across a 1.5-km radius window and standard deviation</td>
<td>4</td>
<td>137.001</td>
</tr>
<tr>
<td></td>
<td>Terrain</td>
<td>Mean compound topographic index across 1-km radius window</td>
<td>3</td>
<td>138.442</td>
</tr>
<tr>
<td></td>
<td>Roads</td>
<td>Linear density (km/km²) of roads classed 1–4 within a 6.4-km radius window. BLM classification: 1 = primary paved highway, 2 = secondary paved highway, 3 = light duty road, constructed and regularly maintained, 4 = primitive road, sedan clearance, not regularly maintained</td>
<td>3</td>
<td>128.444</td>
</tr>
<tr>
<td></td>
<td>Residential</td>
<td>Euclidian distance (km) to residential development (quadratic)</td>
<td>4</td>
<td>168.263</td>
</tr>
<tr>
<td></td>
<td>Conifer</td>
<td>Proportion of conifer/pinyon juniper habitat across a 0.56-km radius window</td>
<td>4</td>
<td>178.292</td>
</tr>
<tr>
<td></td>
<td>Water</td>
<td>Euclidian distance (km) to water (quadratic)</td>
<td>5</td>
<td>175.933</td>
</tr>
<tr>
<td>Patch variables—null model AIC = 204.398</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>Shrub</td>
<td>Proportion of pixels with &gt;10% big sagebrush (quadratic) across a 0.56-km radius window</td>
<td>3</td>
<td>179.397</td>
</tr>
<tr>
<td></td>
<td>Other veg</td>
<td>Mean % ground cover (quadratic) across a 0.56-km radius window and standard deviation</td>
<td>5</td>
<td>177.896</td>
</tr>
<tr>
<td></td>
<td>NDVI</td>
<td>Mean NDVI across a 0.56-km radius window and standard deviation</td>
<td>4</td>
<td>177.235</td>
</tr>
<tr>
<td></td>
<td>Terrain</td>
<td>Mean compound topographic index value across 0.56-km radius window</td>
<td>4</td>
<td>170.328</td>
</tr>
<tr>
<td></td>
<td>Roads</td>
<td>Linear density (km/km²) of all roads within 0.56-km radius window</td>
<td>4</td>
<td>177.235</td>
</tr>
<tr>
<td></td>
<td>Residential</td>
<td>Euclidian distance (km) to residential development (quadratic)</td>
<td>5</td>
<td>171.287</td>
</tr>
<tr>
<td></td>
<td>Conifer</td>
<td>Proportion of conifer/pinyon juniper habitat across a 0.56-km radius window</td>
<td>5</td>
<td>180.792</td>
</tr>
<tr>
<td></td>
<td>Water</td>
<td>Euclidian distance (km) to water (quadratic)</td>
<td>5</td>
<td>175.933</td>
</tr>
</tbody>
</table>

* Normalized difference vegetation index.

b Bureau of Land Management.
RESULTS
We located 102 nests from 75 different individuals between 2000 and 2009 (mean 1.35 ± 0.58 SD nests per individual; range 1–3). Sample sizes were low during the early years of study. We used nests from 2006 to 2009 to develop nest habitat selection models (total \( n \) = 73; 2006 = 14; 2007 = 21; 2008 = 14; 2009 = 24). This sample included repeat nests from 15 females (12 = 2 nests, 3 = 3 nests, 40 = 1 nest), which captured expected female fidelity to nesting areas. We used 29 nests located in 2000–2005 for model evaluation.

The maximum distance a radio marked female traveled from the lek of capture to a nesting site was 10.1 km. We applied 10.1-km circular buffers around all capture leks within the study area to define available nesting habitat for marked females at the landscape scale (Fig. 1).

Landscape Models
After screening variables for form, extent, and correlations, we retained 15 variables (including SDs) across all categories for modeling nesting habitat at the landscape scale (Table 1). The proportion of a 1.5 km-radius moving window that contained sagebrush cover greater than 5% (\( \text{PG}_{e, 	ext{SG}_{gt5}} \)) was the top shrub predictor and used as the landscape base, with a \( \text{AIC} \), from the null model of 44.571 (Table 1). We assessed variables in the remaining categories with the inclusion of the landscape base (Table 1). All shrub height variables were correlated with the landscape base and were not considered further. The most predictive variables (largest decrease in \( \text{AIC} \), from base; Table 1) included linear density of roads classes 1–4 within a 6.437-km radius (\( f_{\text{rd}}^{1-4} \)), mean productivity within a 1-km radius (\( \text{cti}_{\text{cm}} \)), and Euclidian distance to nearest road classed 1 or 2 (\( \text{dist}_{\text{rd}}^{1-2} \)). In combination with the landscape shrub base, each of these simple models had a decrease in \( \text{AIC} \), values from the base model of 25.889, 22.821, and 21.380, respectively.

We developed 344 candidate models (all subsets where \( K \leq 7 \)) using the selected variables and assessed at the landscape scale. All \( w_k \) were low due to the number of models in the candidate set and the nested nature of models (Table 2). Thirty-five models were included in the landscape model 90% confidence set. These models had good fit to the data (likelihood-ratio [\( LR \]) \( \chi^2 \) range = 90.80–106.39, \( D_f = 3–7, P < 0.001 \)). The model averaged standard errors suggest that Gunnison sage-grouse nests in the western portion of the Gunnison Basin are found in areas with a higher proportion of sagebrush cover greater than 5% (1.5-km radius), higher mean productivity within 1 km (CTI), lower road density (classes 1–4) within 6.4 km, are a moderate distance from water (quadratic function), and are further from conifer-juniper forest (0.1-km decay function; Table 3).

### Landscape Model Evaluation
The model averaged surface validated well based on Kendall’s \( \tau \), with excellent (\( \tau_{\text{train}} = 0.936, \text{CI}: 0.912–0.956 \)) and good (\( \tau_{\text{test}} = 0.800, \text{CI}: 0.720–0.881 \)) probability of concordance between model predictions and observed nest or available locations (\( \text{nest}_{\text{train}} = 73; \text{nest}_{\text{test}} = 29, n_{\text{available}} = 3,369; \text{Fig. 2} \)). Of the 102 identified nests, 99 (97%) occurred within the top 5 landscape-scale habitat class bins (bins 6–10), with the 3 omitted nests belonging to the testing group. Bins 6–10 (RSF values >0.119) encompass approximately 57% of the NPS study area (Fig. 2). We considered habitat falling within these bins as crucial nesting areas based on the landscape model.

### Generality of Landscape Model
We used all active lek locations as of 2007 that fell within the predictive surface (\( n = 44 \)) to evaluate the predictive capacity

Table 2. A comparison of the top 10 landscape-scale habitat models used to characterize Gunnison sage-grouse nest occurrence (2006–2009; \( n = 73 \)) in the Gunnison Basin, Colorado. Models are ranked by the change in Akaike Information Criterion corrected for small sample size (\( \Delta \text{AIC} \)). Akaike weights \( (w_k) \) indicate the likelihood of the model being the best of those evaluated (\( n = 344 \)). \( K \) indicates the number of parameters in the model. The top 35 models were included in the 90% confidence set and incorporated in the model averaging procedure.

<table>
<thead>
<tr>
<th>Model</th>
<th>Model structure</th>
<th>( K )</th>
<th>( \text{AIC} )</th>
<th>( \Delta \text{AIC} )</th>
<th>( w_k )</th>
<th>Rank</th>
</tr>
</thead>
<tbody>
<tr>
<td>LAND1</td>
<td>( \text{PG}<em>{e, 	ext{SG}</em>{gt5}} + \text{cti}<em>{\text{cm}} + \text{p}</em>{\text{res}} e + f_{\text{rd}}^{1-4} + \text{litter}<em>{\text{fm}} + (\text{litter}</em>{\text{fm}})^2 + \text{litter}_{\text{fs}} )</td>
<td>113.523</td>
<td>0.000</td>
<td>0.159</td>
<td>1</td>
<td></td>
</tr>
<tr>
<td>LAND2</td>
<td>( \text{PG}<em>{e, 	ext{SG}</em>{gt5}} + \text{dist}<em>{\text{water}} + (\text{dist}</em>{\text{water}})^2 + \text{cti}<em>{\text{cm}} + \text{p}</em>{\text{res}} e + f_{\text{rd}}^{1-4} + \text{cj}_{\text{d}}^{100} )</td>
<td>114.049</td>
<td>0.526</td>
<td>0.122</td>
<td>2</td>
<td></td>
</tr>
<tr>
<td>LAND3</td>
<td>( \text{PG}<em>{e, 	ext{SG}</em>{gt5}} + \text{cti}<em>{\text{cm}} + f</em>{\text{rd}}^{1-4} + \text{cj}<em>{\text{d}}^{100} + \text{litter}</em>{\text{fm}} + (\text{litter}<em>{\text{fm}})^2 + \text{litter}</em>{\text{fs}} )</td>
<td>114.311</td>
<td>0.788</td>
<td>0.107</td>
<td>3</td>
<td></td>
</tr>
<tr>
<td>LAND4</td>
<td>( \text{PG}<em>{e, 	ext{SG}</em>{gt5}} + \text{cti}<em>{\text{cm}} + \text{p}</em>{\text{res}} e + f_{\text{rd}}^{1-4} + \text{cj}_{\text{d}}^{100} )</td>
<td>115.364</td>
<td>1.841</td>
<td>0.063</td>
<td>4</td>
<td></td>
</tr>
<tr>
<td>LAND5</td>
<td>( \text{PG}<em>{e, 	ext{SG}</em>{gt5}} + \text{dist}<em>{\text{water}} + (\text{dist}</em>{\text{water}})^2 + \text{cti}<em>{\text{cm}} + \text{p}</em>{\text{res}} e + f_{\text{rd}}^{1-4} )</td>
<td>115.861</td>
<td>2.338</td>
<td>0.049</td>
<td>5</td>
<td></td>
</tr>
<tr>
<td>LAND6</td>
<td>( \text{PG}<em>{e, 	ext{SG}</em>{gt5}} + \text{dist}<em>{\text{water}} + (\text{dist}</em>{\text{water}})^2 + \text{cti}<em>{\text{cm}} + f</em>{\text{rd}}^{1-4} + \text{cj}_{\text{d}}^{100} )</td>
<td>115.936</td>
<td>2.413</td>
<td>0.047</td>
<td>6</td>
<td></td>
</tr>
<tr>
<td>LAND7</td>
<td>( \text{PG}<em>{e, 	ext{SG}</em>{gt5}} + \text{p}<em>{\text{res}} e + f</em>{\text{rd}}^{1-4} + \text{cj}<em>{\text{d}}^{100} + \text{litter}</em>{\text{fm}} + (\text{litter}<em>{\text{fm}})^2 + \text{litter}</em>{\text{fs}} )</td>
<td>116.201</td>
<td>2.678</td>
<td>0.042</td>
<td>7</td>
<td></td>
</tr>
<tr>
<td>LAND8</td>
<td>( \text{PG}<em>{e, 	ext{SG}</em>{gt5}} + \text{cti}<em>{\text{cm}} + f</em>{\text{rd}}^{1-4} + \text{litter}<em>{\text{fm}} + (\text{litter}</em>{\text{fm}})^2 + \text{litter}_{\text{fs}} )</td>
<td>116.553</td>
<td>3.030</td>
<td>0.035</td>
<td>8</td>
<td></td>
</tr>
<tr>
<td>LAND9</td>
<td>( \text{PG}<em>{e, 	ext{SG}</em>{gt5}} + \text{cti}<em>{\text{cm}} + \text{p}</em>{\text{res}} e + f_{\text{rd}}^{1-4} + \text{dj}_{\text{d}}^{1-2} )</td>
<td>117.380</td>
<td>3.857</td>
<td>0.023</td>
<td>9</td>
<td></td>
</tr>
<tr>
<td>LAND10</td>
<td>( \text{PG}<em>{e, 	ext{SG}</em>{gt5}} + \text{cti}<em>{\text{cm}} + \text{p}</em>{\text{res}} e + f_{\text{rd}}^{1-4} )</td>
<td>117.595</td>
<td>4.073</td>
<td>0.021</td>
<td>10</td>
<td></td>
</tr>
</tbody>
</table>

\( a \) Proportion of pixels with >5% sagebrush cover across 1.5-km radius window.
\( b \) Mean compound topographic index (productivity) across 1-km radius window.
\( c \) Proportion of residential classed habitat within a 1.5-km radius window.
\( d \) Linear density (km/ km²) of roads classed 1–4 within a 6.4-km radius window. Bureau of Land Management classification: 1 = primary paved highway, 2 = secondary paved highway, 3 = light duty road, constructed and regularly maintained, 4 = primitive road, sedan clearance, not regularly maintained.
\( e \) Mean % litter across a 6.4-km radius window.
\( f \) Standard deviation of mean % litter across a 6.4-km radius window.
\( g \) Euclidian distance (km) to water.
\( h \) Decay function of influence of conifer/pinyon juniper classed habitat, decay set at 100 m.
\( i \) Euclidian distance (km) to any road classed 1 or 2. Bureau of Land Management classification: 1 = primary paved highway, 2 = secondary paved highway.
Table 3. Model averaged coefficients ($\beta$) and standard errors (SE) associated with each variable represented in the 90% confidence set ($n = 35$) of landscape-scale models used to predict Gunnison sage-grouse nest occurrence (2006–2009, $n = 73$) across the western portion of the Gunnison Basin, Colorado. Note that the shrub variable (pg_e_sg_gt5) was included in all models.

<table>
<thead>
<tr>
<th>Variable</th>
<th>Model averaged $\beta$</th>
<th>Model averaged SE</th>
<th>Confidence intervals</th>
</tr>
</thead>
<tbody>
<tr>
<td>pg_e_sg_gt5$^a$</td>
<td>47.265</td>
<td>18.483</td>
<td>11.039</td>
</tr>
<tr>
<td>cti_cm$^b$</td>
<td>0.001</td>
<td>&lt;0.001</td>
<td>&lt;0.001</td>
</tr>
<tr>
<td>p_res_e$^c$</td>
<td>−36.784</td>
<td>28.503</td>
<td>−92.651</td>
</tr>
<tr>
<td>f_rd1_4$^d$</td>
<td>−10.590</td>
<td>3.143</td>
<td>−16.750</td>
</tr>
<tr>
<td>litter_fm$^e$</td>
<td>0.715</td>
<td>0.741</td>
<td>−0.738</td>
</tr>
<tr>
<td>(litter_fm)$^e$</td>
<td>−0.024</td>
<td>0.021</td>
<td>−0.064</td>
</tr>
<tr>
<td>p_res_ec</td>
<td>36.784</td>
<td>28.503</td>
<td>19.083</td>
</tr>
<tr>
<td>f_rd1_4$^d$</td>
<td>10.590</td>
<td>3.143</td>
<td>−16.750</td>
</tr>
<tr>
<td>litter_fm$^e$</td>
<td>0.715</td>
<td>0.741</td>
<td>−0.738</td>
</tr>
<tr>
<td>(litter_fm)$^e$</td>
<td>−0.024</td>
<td>0.021</td>
<td>−0.064</td>
</tr>
<tr>
<td>dist_water$^g$</td>
<td>1.152</td>
<td>0.519</td>
<td>1.354</td>
</tr>
<tr>
<td>(dist_water)$^g$</td>
<td>−0.473</td>
<td>0.230</td>
<td>−0.923</td>
</tr>
<tr>
<td>cj_d100$^h$</td>
<td>−1.648</td>
<td>0.789</td>
<td>−3.195</td>
</tr>
<tr>
<td>dist_rd1_2$^i$</td>
<td>−0.002</td>
<td>0.015</td>
<td>−0.031</td>
</tr>
<tr>
<td>ndvi_cm$^j$</td>
<td>−0.398</td>
<td>0.321</td>
<td>−1.028</td>
</tr>
<tr>
<td>ndvi_csk$^k$</td>
<td>0.016</td>
<td>0.353</td>
<td>0.027</td>
</tr>
<tr>
<td>cv_dist_res$^l$</td>
<td>0.036</td>
<td>0.027</td>
<td>0.017</td>
</tr>
<tr>
<td>(dist_res)$^l$</td>
<td>−0.006</td>
<td>0.005</td>
<td>−0.015</td>
</tr>
<tr>
<td>herb_fm$^m$</td>
<td>0.014</td>
<td>0.025</td>
<td>−0.035</td>
</tr>
<tr>
<td>(herb.fm)$^m$</td>
<td>−0.001</td>
<td>0.001</td>
<td>0.004</td>
</tr>
</tbody>
</table>

$^a$ Proportion of pixels with >5% sagebrush cover across 1.5-km radius window.
$^b$ Mean compound topographic index (productivity) over a 1-km radius window.
$^c$ Proportion of residential classed habitat within a 1.5-km radius window.
$^d$ Linear density (km/ km$^2$) of roads classified 1–4 within a 6.4-km radius window. Bureau of Land Management classification: 1 = primary paved highway, 2 = secondary paved highway, 3 = light duty road, constructed and regularly maintained, 4 = primitive road, sedan clearance, not regularly maintained.
$^e$ Mean % litter across a 6.4-km radius window.
$^f$ Standard deviation of % litter across a 6.4-km radius window.
$^g$ Euclidian distance (km) to water.
$^h$ Decay function of influence of conifer/pinyon juniper classed habitat, decay set at 100 m.
$^i$ Euclidian distance (km) to any road classified 1 or 2. Bureau of Land Management classification: 1 = primary paved highway, 2 = secondary paved highway.
$^j$ Mean normalized difference vegetation index (productivity) over a 1-km radius window.
$^k$ Standard deviation of the mean normalized difference vegetation index over a 1-km radius window.
$^l$ Euclidian distance (km) to residential development.
$^m$ Mean % herbaceous cover over a 6.4-km radius window.

Figure 2. Model averaged relative probability surface predicting the crucial nest areas based on landscape-scale occurrence of Gunnison sage-grouse nest sites applied to the study area in the western portion of the Gunnison Basin, Colorado. We overlaid model development ($n = 73$) and model evaluation ($n = 29$) nest sites for comparison of model prediction.
of the landscape-scale model extrapolated to the entire Gunnison Basin. The model averaged surface applied across the entire basin had fair probability of concordance (\(c_{lek} = 0.721, \text{CI:} 0.642–0.800\)) between model predictions and observed lek or available locations \(n_{lek} = 44; n_{available} = 10,710; \text{Fig. 3}\). Of the 44 active leks identified within the basin, approximately 90% (40) occurred within bins 3–10, comprising approximately 62% of the landscape (Fig. 3).

**Patch Scale Models**

We defined availability for patch-scale models based on crucial nesting areas identified from the landscape-scale nesting model (bin 6 or greater). This left an area roughly half the size (approx. 36,380 ha) of the original study area considered to be available for patch models, subsequently reducing the number of available points to 1,870 for contrast against the 73 nest locations. After screening variables for form, extent, and correlations, we retained 15 variables (including SD) for patch modeling. The top AIC\(_r\)-selected shrub base variable was the quadratic form of the proportion of 0.564-km moving window that was classed as big sagebrush cover greater than 10%. This base model had a \(\Delta\text{AIC\(_r\)} = -25.932\). Other key predictor variables (largest decrease in AIC\(_r\) from base; Table 1) considered in conjunction with the patch-scale shrub base were mean productivity over 0.564-km extent (cti_bm), Euclidian distance to residential classed habitat (dist_res, quadratic function), and Euclidian distance to nearest road classed 1 or 2 (dist_rd1_2). In combination with the patch shrub base, each of these simple models showed a decrease in AIC\(_r\) values from the base model of 11.134, 9.110, and 9.069, respectively.

We evaluated 590 candidate models (all subsets \(K \leq 7\)) comprised of the selected variables with their forms and window scale extents maintained (Table 1; see Table 4 for the 10 most predictive models). Seventy-six models were included in the 90% confidence set and contributed to the model averaged surface. These models had good fit to the data (LR \(\chi^2 = 47.65–62.85, \text{Df} = 4–7, P < 0.001\)). The model averaged standard errors of variables contained within the patch scale models suggest that Gunnison sage-grouse nests are located in areas that have either lower or higher proportions (0.564-km moving window) of high big sagebrush cover (>10% big sagebrush cover; convex quadratic function), are a moderate distance from residential development and water sources (quadratic functions), are further from roads classed 1 or 2, and have higher vegetation productivity (ndvi_am and cti_bm; Table 5). Although the linear portion of the distance to residential development term had confidence intervals overlapping zero, a likelihood ratio test confirmed that the quadratic form of the variable was an important contributor determining Gunnison sage-grouse nest site selection (LR \(\chi^2 = 12.87, P < 0.002\)).

![Figure 3](image_url)

Figure 3. Relative probability of Gunnison sage-grouse (GUSG) nest occurrence predicted across all occupied and potential habitat within the Gunnison Basin. We developed the model using 73 nest locations collected from 2000 to 2005 within the western portion of the Gunnison Basin, Colorado. We used leks active in 2007 to test the generality of model predictions outside of the geographic space from where the model was developed, and show their locations for visual comparison. Non-predicted spaces are areas where model inputs were not available.
Table 4. A comparison of the top 10 patch-scale habitat models used to characterize Gunnison sage-grouse nest occurrence (2006–2009; \( n = 73 \)) in the Gunnison Basin, Colorado. Models are ranked by the change in Akaike weight \((\Delta AIC_c)\) values. Akaike weight \((w_i)\) indicate the likelihood of the model being the best of those evaluated \((\Delta AIC_c < 2)\). The top 76 models were included in the 90% confidence set and were incorporated in the model averaging procedure.

<table>
<thead>
<tr>
<th>Model</th>
<th>(\Delta AIC_c)</th>
<th>(w_i)</th>
<th>Rank</th>
</tr>
</thead>
<tbody>
<tr>
<td>PATCH1</td>
<td>0.000</td>
<td>1.000</td>
<td>1</td>
</tr>
<tr>
<td>PATCH2</td>
<td>1.564</td>
<td>0.975</td>
<td>2</td>
</tr>
<tr>
<td>PATCH3</td>
<td>1.995</td>
<td>0.950</td>
<td>3</td>
</tr>
<tr>
<td>PATCH4</td>
<td>2.101</td>
<td>0.933</td>
<td>4</td>
</tr>
<tr>
<td>PATCH5</td>
<td>2.302</td>
<td>0.918</td>
<td>5</td>
</tr>
<tr>
<td>PATCH6</td>
<td>3.382</td>
<td>0.628</td>
<td>6</td>
</tr>
<tr>
<td>PATCH7</td>
<td>3.828</td>
<td>0.505</td>
<td>7</td>
</tr>
<tr>
<td>PATCH8</td>
<td>4.056</td>
<td>0.449</td>
<td>8</td>
</tr>
<tr>
<td>PATCH9</td>
<td>4.098</td>
<td>0.438</td>
<td>9</td>
</tr>
<tr>
<td>PATCH10</td>
<td>4.682</td>
<td>0.108</td>
<td>10</td>
</tr>
</tbody>
</table>

a Proportion of pixels with \(>10\%) big sagebrush cover across 0.564-km radius window.
b Mean normalized difference vegetation index (productivity) over a \(3 \times 3\) (30-m cells) window.
c Euclidian distance to residential classed habitat.
d Euclidian distance (km) to any road classed 1 or 2. Bureau of Land Management classification: 1 = primary paved highway, 2 = secondary paved highway.
e Mean compound topographic index (productivity) across a 0.564-km radius window.
f Proportion of pixels across a 0.564-km window classed as conifer/pinyon juniper habitat.
g Slope in percent.

Patch Scale Model Evaluation

The model averaged surface validated well based on Kendall’s \(\tau\), with good \((\tau_{\text{train}} = 0.986, \text{CI: 0.982–0.993})\) and fair to good \((\tau_{\text{test}} = 0.794, \text{CI: 0.699–0.889})\) probability of concordance between model predictions and observed nest or available locations \((n_{\text{train}} = 73; n_{\text{test}} = 26, n_{\text{available}} = 1,870; \text{Fig. 4})\). Of the 99 nests sites contained within the patch model extent, 95 (approx. 96%) occurred in the top 6 bins (5–10), encompassing approximately 56% of the available habitat at the patch extent (previously restricted by the landscape model). Sixty-nine nests (approx. 70%) were located in the top 2 bins (approx. 13% of available habitat). Strong selection for the higher ranked predicted habitats suggests the model is predictive.

Threshold Response

Landscape models indicated that Gunnison sage-grouse selected nesting areas containing >93% (lower SD = 90%) of a 1.5-km area (radius moving window) with >5% sagebrush cover (all species; \text{Fig. 5a}). Probability of nesting approaches zero for all landscapes when the proportion of sagebrush cover >5% is less than roughly 90%. After accounting for the landscape-scale selection of sagebrush, shrubs still entered into the patch scale model, reflecting the hierarchical nature of selection. However, at the patch scale, moderate proportions of big sagebrush (all subspecies) cover >10% within 0.564-km radius patches were avoided (\text{Fig. 5d}).

At the landscape scale, Gunnison sage-grouse were more likely to nest in areas that contained less than roughly 0.55 km/km\(^2\) of roads (class 1–4) across a 6.4-km radius landscape (\text{Fig. 5b}). Despite avoidance of areas with high road density, females also avoided nesting in close proximity to major roads (class 1–2) at the patch scale. The patch–landscape threshold response curves highlight relatively low probabilities of nest occurrence until about 8 km from major roads (\text{Fig. 5e}). This pattern was similar for residential development, with threshold response curves capturing selection for large landscapes (1.5-km radius) with a low density of residential development (<1%; \text{Fig. 5c}). The patch–landscape threshold curves suggest maximum probabilities of nest site selection are reached at approximately 2.5 km from any given development (\text{Fig. 5f}).

Evaluation of the Hierarchical Surface

The hierarchical model surface validated well based on Kendall’s \(\tau\), with good \((\tau_{\text{train}} = 0.989, \text{CI: 0.859–0.938})\) and fair \((\tau_{\text{test}} = 0.789, \text{CI: 0.684–0.893})\) probability of concordance between model predictions and observed nest or available locations \((n_{\text{train}} = 73; n_{\text{test}} = 26, n_{\text{available}} = 1,870; \text{Fig. 4})\). Ninety-three percent of all nest sites (92/99) were located within approximately 50% of the patch landscape (bins 6–10; \text{Fig. 4}). Using this as a potential threshold to further refine crucial nesting habitat to include both scales of selection (patch and landscape), we identify 18,102 ha as crucial habitat. This is a refinement of 18,301 ha over what was identified as crucial habitat using the landscape model only (36,403 ha).
shrub variable \((pg_b_ba_gt10)\) was included in all models. Models used to predict Gunnison sage-grouse nest occurrence (2006–2009; Aldridge et al. 1980, Rettie and Messier 2000, Mayor et al. 2009).

Our landscape-scale results follow patterns seen with greater sage-grouse and reinforce that both species respond to landscape patterns and changes to resources when selecting nesting habitat (Aldridge and Boyce 2007, Doherty et al. 2004). Sage-grouse require large landscapes important for nesting females. Despite prior modeled selection for resources across large landscapes, we show that females are indeed making a hierarchical decision, selecting for patches of sagebrush habitat while avoiding proximity to residential developments and major roads (Table 5); a multi-scale decision process. Hierarchical selection is theorized to be an important biological construct, for which, assessment of these processes should increase understanding of resource needs for a species (Johnson 1980, Rettie and Messier 2000, Mayor et al. 2009).

Harvey and Weatherhead (2006) assessed hierarchical selection for eastern massasauga rattlesnakes \((Sistrurus c. catenates)\) and found that most landscape-scale selection patterns could be explained primarily by microhabitat availability within those landscapes. This was not the case for Gunnison sage-grouse, with resources at both scales affecting nest selection patterns.

Our landscape-scale results follow patterns seen with greater sage-grouse and reinforce that both species respond to landscape patterns and changes to resources when selecting nesting habitat (Aldridge and Boyce 2007, Doherty et al. 2010, Connelly et al. 2004). Sage-grouse require large expanses of contiguous sagebrush habitat across all life stages (Connelly et al. 2004). Threshold response curves allowed us to identify that large areas of sagebrush were used by Gunnison sage-grouse for nesting if roughly 95% or more of the area contained sagebrush with >5% cover (remotely estimated; Fig. 5a). Sagebrush habitat meeting these requirements is abundant within the Gunnison Basin, however much of it has been degraded due to fragmentation. Although others have found that fragmented habitats are
avoided by greater sage-grouse (Oyler-McCance et al. 2001, Lyon and Anderson 2003, Aldridge and Boyce 2007), response curves from our models identify thresholds of development, above which Gunnison sage-grouse are likely to avoid nesting. Using our landscape crucial nesting habitat cutoff probability (0.119) we can estimate that areas with road densities (roads classed 1–4) above 0.50 km/km² across a 6.4-km window area (Fig. 5b) are avoided. Similarly, if residential density exceeds approximately 2% of an area within a 1.5-km moving window (Fig. 5c), females avoid nesting in the area, regardless of the amount of sagebrush habitat it contains. Although the effect of residential density appears less influential (95% CIs for the coefficient overlap zero), the top landscape model (LAND1) is >3 AICc points better than LAND8, which is the same model excluding the residential variable (Table 2). A likelihood ratio test between the 2 models ($\chi^2 = 5.27, P = 0.022$) also suggests residential density is an important contributor to nest-site selection.

Although increased residential and road development directly remove habitat, Gunnison sage-grouse behavioral response to these features as a result of increased human activity, noise, and changes in predator distribution and abundance, may be equally important (Forman and Alexander 1998). Similarly, it has been suggested that greater sage-grouse avoidance of forest–sagebrush interface, as we found for Gunnison sage-grouse, is a behavioral response, attempting to minimize exposure to avian predators (Doherty et al. 2008, Freese 2009, Coates and Delehanty 2010).

Although conservation interest in extrapolating our landscape-model across the entire Gunnison Basin is high, and the model had fair prediction, such extrapolations should be undertaken with caution (see Miller et al. 2004). Most importantly, these extrapolations need to be challenged with independent data to ensure predictions are valid (Miller et al. 2004). Although nest sites were not available over the entire basin, lek locations, which have previously been shown to correlate well with greater sage-grouse nest habitat models (Doherty et al. 2010), were reasonably predicted by the extrapolated model. Noted differences in available habitats (i.e., agriculture was limited in the NPS study area, but locally abundant across the basin) certainly reduced the effectiveness of model extrapolation. Despite some locally reduced fit of the model extrapolation, we argue that this landscape model could be used as an initial management tool to identify crucial nesting habitat for Gunnison sage-grouse across the basin. However, as with any spatial model, predictions should be viewed as one of many possible tools, and should not replace local knowledge when making site-

---

**Figure 4.** Crucial Gunnison sage-grouse nesting habitat within the western portion of the Gunnison Basin, Colorado. We considered crucial nesting areas the top 5 bins (6 or higher) from the landscape-scale nest occurrence model (a), and a priori refined the assessment extent of the patch-scale model (b) predicting crucial nest site habitat. We multiplied the 2 surfaces to produce a composite crucial nesting habitat surface (c) which accounts for both landscape- and patch-level habitat requirements for Gunnison sage-grouse nesting.
When extrapolated across the Gunnison Basin, this would effectively capture 97% (99/102) of known nest sites. We feel that our conservative threshold of bin 6 or higher (>0.119 relative probability), was reasonable for identifying crucial habitat within the NPS study area, given it effectively captures 97% (99/102) of known nest sites. When extrapolated across the Gunnison Basin, this would suggest approximately 50% of the landscape is crucial nesting habitat required to support >90% of known nests within the study area. Although we have mapped crucial nesting habitat, this does not preclude the need to consider other life stages, the fitness of individuals using these habitats (i.e., source vs. sink habitats), or the connectivity between habitats and across life stages, all of which might require different management actions (protection vs. enhancement; Aldridge and Boyce 2007, Aldridge et al. 2008).

We feel that our hierarchical modeling approach more closely aligns with Johnson’s (1980) original perception on how animals select resources, choosing larger home ranges or areas that meet an animal’s general needs (our landscape scale) vs. selecting smaller areas that meet the species’ high priority as a conservation requirement (nest patch model). Accordingly, despite having influence at the landscape-scale, some habitat character-
istics continued to affect selection at the patch scale. For instance, although more sagebrush cover above 5% was selected at the landscape scale (1.5-km radius areas; Fig. 5a) sagebrush continued to influence selection at the patch scale, but the strongest predictor was proportion big sagebrush >10% (0.564-km radius). Even though models at both scales contained similar variables, correlation between the 2 prediction surfaces was minimal (Pearson’s $r = 0.411; n = 1,870$ patch-scale available points), suggesting significant additional variation is explained by the patch model.

Predicted responses to big sagebrush at the patch scale suggest selection is bimodal, with birds selecting for low or high amounts of big sagebrush above 10% cover; proportions below 0.2 or above 0.4 (Fig. 5d). However, only 3 of 73 nests used for model building were at the high end of the distribution (>0.3 in Fig. 5d), and we caution against drawing strong inferences when interpreting this part of the curve. Several factors could be contributing to this overall predicted relationship. First, although big sagebrush appeared to have a strong contribution to selection (Table 5), the low sample sizes at the high end of the distribution in conjunction with the symmetric nature of the quadratic function magnify this relationship. Second, although the patch model shows some avoidance of moderate amounts of higher estimated (>10%) big sagebrush cover, prior selection for sagebrush in our landscape model redefined availability for the patch model. Patches had already been selected with suitable sagebrush; densities in these patches were just lower than 10% big sagebrush cover. This avoidance of dense cover may subsequently reflect a need for females to seek out local habitat patches with increased lateral nest cover provided by grass, forbs, or other shrubs, which are not likely available in large areas with greater big sagebrush cover (Aldridge and Brigham 2002, Crawford et al. 2004, Hagen et al. 2007). Finally, this does not refute the well supported evidence that sage-grouse will ultimately choose to place their nests in a dense sagebrush stand (for reviews see Connelly et al. 2004, Hagen et al. 2007). Big sagebrush summarized here was at a 564-m radius moving window extent, clearly capturing different vegetation characteristics than that of an individual shrub. Current limitations of remotely-sensed vegetation models prevent our ability to capture intimate vegetation characteristics within a sagebrush stand or individual plants (<15 m; Homer et al. 2008, 2012). Despite these scalar differences, our hierarchical model predictions accurately predicted independent nest locations.

Gunnison sage-grouse also exhibit a clear avoidance of paved, high traffic volume roads (classes 1 and 2; Fig. 5e) during nesting at the patch scale, in addition to prior avoidance of higher density 2-wheel drive accessible roads at the landscape scale (6.4-km radius; classes 1–4; Fig. 5b). Landscape-level response may reflect selection for less fragmented areas (Holloran 2005, Aldridge and Boyce 2007, Walker et al. 2007, Doherty et al. 2008). However, the direct avoidance of high volume roads in patch-scale models reinforces that Gunnison sage-grouse are selecting for resources hierarchically (Rettie and Messier 2000, Mayor et al. 2009).

This may be a behavioral response to disturbance (Forman and Alexander 1998) and avian predators (Knight and Kawashima 1993, Connelly et al. 2004, Bui et al. 2010) when choosing where to place a nest, despite prior selection for less fragmented habitats (landscape model). Road avoidance extends out to approximately 8 km from high volume roads, with the relative probability of nest occurrence increasing steadily beyond that point (Fig. 5e). This corresponds with a lek analysis in Wyoming and Utah which found that greater sage-grouse leks within 7.5 km of Interstate 80 appear to have declined at a much faster rate than those further away (Connelly et al. 2004). A recent analysis assessing the effects of urban development on greater sage-grouse used a 6.9-km foraging distance for mammalian and corvid predators (Knick et al. 2011). These numbers correspond with our findings of road avoidance during nesting by Gunnison sage-grouse.

Development in the Gunnison Basin is becoming increasingly exurban. This type of development results in a highly fragmented landscape as the number of roads and buildings (Theobald et al. 1996, Mitchell et al. 2002) in previously contiguous patches of sagebrush increases, clearly reducing nesting habitat quality for Gunnison sage-grouse. High density residential development was avoided at a landscape scale, and nesting females chose to place nests farther away from any single development at the patch scale. This avoidance was not linear, with a threshold at approximately 2.5 km (Fig. 5f). The joint effects of roads and residential developments within sagebrush habitats will have negative consequences on Gunnison sage-grouse nesting habitat. With future developments on the horizon for the Gunnison Basin, housing and associated road developments within 2.5 km of identified crucial habitat should be evaluated cautiously, due to the potential direct and functional loss of nesting habitat (Aldridge and Boyce 2007).

Overall, landscape models were more predictive, having higher probabilities of concordance between model predictions and independent data. This is expected, given the landscape nature of the species (Patterson 1952, Knick et al. 2003, Connelly et al. 2004), and the inability of our model inputs to characterize fine scale (<30 m) habitat characteristics known to be important for sage-grouse (Connelly et al. 2000, Hagen et al. 2007). Because of the truly nested nature of our modeling approach, some variation at the patch scale may already be captured by the landscape scale models, although this appears to be minimal. When the patch and landscape scale surfaces were multiplied to produce the hierarchical surface, evaluation was strong, and the area implicated as crucial habitat was reduced by half from considering landscape models only. We believe using the hierarchical surface, where available, will allow for better informed management decisions, particularly when sage-grouse habitat occurs on lands with multiple use mandates.

**MANAGEMENT IMPLICATIONS**

Our models depicting crucial nesting habitat of Gunnison sage-grouse should be seen as an initial tool to inform management and conservation across large landscapes, but
does not obviate the need for local, targeted assessments and adaptive management within these areas (Aldridge et al. 2004, Aldridge and Boyce 2007). Models identify priority conservation areas for protection (i.e., high quality red areas on maps) and areas where management actions might be warranted (i.e., moderate quality yellow areas on maps), providing an initial baseline spatial accounting system (Aldridge and Boyce 2007) to track crucial nesting habitat for the species. Realistically, managers will want to know how best to threshold these continuous RSF relative probability surfaces to explicitly identify crucial habitats for protection. Given the extreme fidelity of these imperiled birds to nesting areas (Fischer et al. 1993, Connelly et al. 2004), we argue that managers should indeed strive to protect as many of these habitats as possible, and reinforce our recommendation that crucial habitat capture \( \geq 90\% \) of known nest locations. Similarly, future developments (urban or roads) should be prevented within 2.5 km of identified crucial nesting habitat, if habitats, and thus populations, are to be maintained. Finally, because input data are spatial, managers could begin to assess the impacts of future landscape scenarios, such as evaluating the potential cost and benefit for Gunnison sage-grouse nesting habitat of future urban expansion or decommissioning of roads. These models and maps could subsequently inform travel management planning, environmental impact statements, records of decisions, and development and conservation planning in general. Future work should concentrate on developing similar models for brood-rearing and winter habitat, both important life stages for sage-grouse. Ultimately, management actions must not only consider protection of crucial habitat, but also connectivity between patches, both within, and across life stages (Aldridge and Boyce 2007), if long-term persistence of the species is desired.

ACKNOWLEDGMENTS

For their hard work and dedication to field data collection, we thank F. Bayouth, J. Blum, L. Brummer, B. Cooper, B. Detamore, S. Gregory, K. Kasik, P. Magee, T. Meyers, J. Podeszek, T. Root, C. Santana, J. Severson, M. Sewalt, M. Szczypiniski, J. Timmer, and G. Wann, with a special thanks to J. Frey and B. Palmer for their many years of work to keep the project running. This research has been supported by the U.S. Geological Survey and the National Park Service, with direct funding from the USGS Fort Collins Science Center, Black Canyon of the Gunnison National Park and Curecanti National Recreation Area, the USGS Natural Resources Preservation Program, National Park Service Challenge Cost Share Program, the Rocky Mountain Cooperative Ecosystem Studies Unit, the U.S. Fish and Wildlife Service, Earth Friends Conservation Fund, and the Western National Parks Association. We also thank the Colorado Division of Wildlife and specifically, M. Phillips, for guidance with data collection and discussions about Gunnison sage-grouse. We especially thank B. Cade for statistical advice, M. O'Donnell for assistance with spatial analyses, and A. Anderson, M. Pelletier, G. Gantt, and J. Sonderegard for assistance with the roads layer classification.

We thank J. Sedinger, J. Young, P. Coates, G. Chong, and D. Reinkensmeyer for helpful comments on earlier drafts of this manuscript. Finally, we thank members of the Gunnison Sage-Grouse Strategic Committee for their support of our research. Any use of trade, product, or firm names in this publication is for descriptive purposes only and does not imply endorsement by the United States government.

LITERATURE CITED


Harrell, F. E., Jr. 2001. Regression modeling strategies: with applications to linear models, logistic regression, and survival analyses. Springer-Verlag, New York, New York, USA.


STATACORP. 2007. Stata user’s guide: release 10. Stata Press, College Station, Texas, USA.


