Research Article

Modelling Marten (Martes americana) Movement Costs in a Boreal Forest: Effects of Grain Size and Thematic Resolution

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Received 24 October 2011; Revised 30 March 2012; Accepted 17 April 2012

Academic Editor: Daniel Rubenstein

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We investigated landscape resistance to movements of American marten (Martes americana) based on snow-tracking data. We generated movement cost maps of the study area with different grain size, thematic resolution, and habitat-specific resistance to movements. We compared simulated tracks obtained from resistance maps to real tracks plotted along transects that we surveyed in winters 2004 to 2008 at the Montmorency Forest, Quebec, Canada. Simulated tracks were located at the intersection between least-cost paths simulated across the study area and transects. We used nearest-neighbour distances between simulated and real tracks to assess the performance of resistance maps and estimate landscape resistance parameters. Simulations with specified costs to movement for open areas, young forest, and mature forest performed better than simpler resistance scenarios, suggesting that resistance to marten movements differed among those landscape attributes that were considered. Simulations with a map grain size of 100 m performed significantly better than 5, 25, and 300 m, possibly because of gap crossing avoidance. Model performance (compared to null model) was maximal when resistance to movement in open habitat was set to 20 times higher than in mature forest, but uncertainty around this estimate was large. This research demonstrates that presence-only (point) data can be used to parameterize movements using spatially explicit modelling.

1. Introduction

Habitat selection studies have long assumed that access to habitats is a negligible constraint for animal movements within landscapes [1], but access limitations that are attributable to roadways and forest fragmentation are increasingly recognized [2]. Movement constraints are governed by spatial patterns of resource used to maximize fitness (e.g., access to critical resources and avoidance of predation). They reduce connectivity [3] within the established home range of animals and limit accessibility of habitat patches to dispersing organisms [4], thereby affecting population dynamics. Thus, the occurrence of an animal in a habitat may depend on the surrounding matrix, and the consideration of movement constraints appears to be crucial in the development of habitat selection models [1, 5, 6]. Spatially explicit simulations including least-cost paths [7, 8] have become increasingly popular for modelling optimum movement routes of animals. Least-cost models can simulate animal movements based on the configuration of different spatial features within a heterogeneous landscape, and the hypothetical resistance to movement associated with them. In testing various hypothetical management scenarios, Sutcliffe et al. [9] concluded that least-cost path simulation is a powerful tool in landscape management planning. Results of Broquet et al. [10] showed that inclusion of the resistance of landscape features in models better explained marten (Martes americana) genetic patterns than did Euclidean distances. Desrochers et al. [11] also better evaluated ovenbird (Seiurus aurocapilla) movement resistance in open areas at small and large scales with least-cost paths simulations than with Euclidean distances. These studies show that resistance to movement across the landscape contributes to shaping movement and gene flow in heterogeneous landscapes.

Predicting costs to movement is essential for identifying and preserving habitat configurations that promote gene flow and, thus, species persistence [9]. When Hargis et al. [12] found lower capture rates of American martens in small
forest patches, they suggested four factors that contributed to decreased occupancy, roughly corresponding to (1) lack of adequate shelter; (2) absence of prey; (3) exposure to harsh weather; (4) interspersion with low-quality habitats. The combination of these four factors could translate into a global “perceived” movement resistance by martens [13]. However, such an aggregate, “perceived” movement resistance has rarely been quantified, leading many wildlife scientists to estimate input parameters in least-cost path models purely as expressions of expert opinion [14]. Some authors have criticized this practice of using habitat suitability models to build cost surfaces [15], arguing that it may induce errors in predictions if models are not validated [16]. To circumvent this problem, methods are needed to provide truly empirical movement resistance estimates. Recent studies have attempted to estimate resistance by comparing model fit to real data [17–19]. Recently, Driezen et al. [20] presented a method of validating the best least-cost path tested model with real dispersal paths, and reiterated the necessity to work with habitat specialist species to estimate cost values throughout the landscape. Cushman and Lewis [21] predicted landscape resistance from movement data and expressed the need for applying correct landscape variables and thematic resolution values to resistance models. Cushman et al. [22, 23] used a path randomization approach in which the reference path’s topology is preserved by randomly shifting and rotating it to produce a group of available paths to which it can be compared.

The first objective of this study was to evaluate the performance of various landscape resistance maps, based on an approach that can be easily applied to other study areas without perturbing animals during a critical period. In spatially explicit models, the application of movement resistance values to different land cover types involves creating grid maps and choosing their parameterization values. The second objective of this study was to optimize marten movement resistance, map grain size, and thematic resolution to estimate parameterization values and help in building future models of landscape-level genetics related to this species. Map resolution of resistance surfaces should be based on structural and functional connectivity rather than being determined solely from landscape or environmental features [14]. Previous studies have shown that movements of dispersing juveniles or adults foraging within their home-ranges depend on the fragmentation level of the landscape [22]. In resistance surfaces, map resolution cannot include connectivity of habitats across the landscape, hence the interest in simulating least-cost paths that progress through landscapes according to the configuration of different spatial features.

To meet the objectives of this study, we compared the resistance of open, young, and closed forest stands to movements of American martens during winter at four map grain sizes. We estimated relative movement resistance inside versus outside of forest cover by comparing the spatial proximity of least-cost path simulations to occurrence data, with different resistance parameters assigned to forests, roads, and clear cuts. We assumed that the marten tracks that we observed were a result of martens following least-cost-paths across the study area. Thus, we simulated least-cost-paths (LCPs) across the landscape, simulated tracks where those LCPs intersected transects sampled for real tracks, and compared simulated tracks with the observed marten tracks. Our approach illustrates the relevance of presence-only data for the comparison of mechanistic hypotheses pertaining to animal movement.

2. Materials and Methods

2.1. Study Area. We carried out field work in Montmory Forest, Quebec, Canada (47°19’ N, 71°9’ W; Figure 1). This 66 km² boreal forest is managed for timber and recreation, but it has a long history of use for teaching and research by Laval University. Balsam fir (Abies balsamea) and black spruce (Picea mariana) were the dominant tree-species, followed by white birch (Betula papyrifera), white spruce (Picea glauca), and trembling aspen (Populus tremuloides). From January to March (1971–2000), daily mean temperatures were $-15.8 \pm 2.5^\circ C$, $-14.1^\circ C \pm 2.9^\circ C$, and $-7.9^\circ C \pm 2.5^\circ C$, respectively [24]. Annual precipitation as snow (November to April) generally exceeded 6 m [24]. Peak snow accumulation varied from 1 to 2 m approximately, depending on years, forest cover, and topography (unpubl. data from snow depth surveys and automated weather station). The topography is hilly, with elevations ranging from 750–1000 m. One of the management objectives of Montmory Forest is to recreate a forest stand mosaic that reflects natural disturbances [25]. To reach this goal, harvest patches are separated into three size groups ($<10$ ha, 10–30 ha, and 30–100 ha) and each size group has the same total area, which results in the creation of many small patches and few large patches. Moreover, the managers of Montmory Forest have staggered harvests through time to obtain four main groups of stand ages (based on sylvicultural practice): (1) regenerating: 0–20 y, (2) young: 21–40 y, (3) mature: 41–60 y, and (4) old-growth: >60 y. Fine-grained sylvicultural practices have resulted in a very dense road network (2.91 km/km²) across the area. Trapping and hunting have not been allowed, but recreational activities such as cross-country skiing and walking are permitted.
2.2. Forest Data. All of the forest inventory data (stand age, cover type, stand limits, road networks) were provided by the managers of the Montmorency Forest who have compiled datasets from aerial photographs (scale 1 : 10 000) and GPS. Every year, stand characteristics (age, height, density, etc.) are ground-truthed at control points.

2.3. Snow-Tracking. We documented the spatial distribution of marten tracks by snow-tracking each winter (1 January–31 March) from 2004 to 2008. Snow-tracking consisted of locating marten tracks along transects traversing all habitat types in the Montmorency Forest. Transects were conducted on all unplowed forestry roads and trails (length 200 m to 5 km) and along a random sample of straight N-S or E-W lines (length 500 m to 2 km) taken from 67 lines placed along a grid (500 m spacing) superimposed upon the study area. Each transect was inventoried only once per winter, on snowshoes or on a snow-mobile operating at very low speed. Snow-tracking occurred only when neither snow falls > 3 cm nor wind speeds > 30 km/h were forecast, because in these cases, snow tracks are covered or swept away [26]. Track quality was described as low (single feet not distinguished), good (feet distinguished), or high (toes distinguished), with only good- and high-quality tracks being included in analyses. We did not follow marten tracks, but we recorded each marten path that crossed a transect as a point feature with a GPS receiver (Trimble GeoExplorer; Figure 2). Paths that crossed the transect several times over short distances (< 100 m) were likely to be from the same individual but we did not consider this to be a statistical issue because we used mean nearest-neighbour distances from several tracks as statistical units (see Section 2.10). All tracks showing the mustelid two-print track pattern, with two-print widths ranging from 8–13 cm, were assumed to be marten [27]. The presence of fisher (Martes pennanti) is unusual in the area, so we assumed that all tracks with this pattern were marten.

2.4. Resistance in Open Areas. We arbitrarily ascribed the highest resistance to movement in open areas, with ten values ranging from 4 to 1000 (constant set of values in mature forest, see [28]). Minimal resistance to open areas was set to 4 in order to vary young forest resistance value relative to the resistance in open path with ratios of 14, 12 and 34. Resistances > 1000 often generated simulated movements along the border of the study area and were assumed to be out of a biologically realistic range for marten. Due to habitat selection by American marten [29], we assumed that mature forest stands offered the lowest resistance to marten movements, and set its value to 1 in all resistance maps [28].

2.5. Thematic Map Resolution. We created three scenarios of movement resistance for marten that varied in the cost values that we assigned to different habitat types (Figure 3). Each scenario was described with either two or three habitat categories. American marten in Quebec are known to be a specialist of dense and mature forest cover because they select habitat types according to vertical structure and structural features like snags and stumps [29]. Thematic map resolution was based on known marten habitat requirements in winter at the Montmorency Forest, that is, avoidance of 0–10 y-old stands, significant preference for > 30 y-old stands, and weak avoidance of 11–30 y-old stands [30]. A first scenario partitioned the landscape into “Open” areas (0 to 10 y-old stands, roads, rivers, lakes, peat bogs, gravel-pits, and islands) and “Forest” (> 10 y-old forests). A second scenario partitioned the landscape into “Open” areas (> 25 m from forest), “Edge” (Open areas < 25 m from forest), and “Forest” (> 10 y-old). Resistance values for forest edges were set as half the resistance of open areas. This second scenario was aimed to evaluate forest border effect on marten movements, based on field observations that marten tracks were more often recorded on narrow roads and in the first few meters of open areas than in “core” open habitats. A third scenario partitioned the landscape into: “Open habitat,” “Young forest” (11 to 30 y-old), and “Mature” (> 30 y-old), with no edge effect. The latter scenario was subdivided in three subgroups, according to the resistance of young stands relative to that of open areas (ratio = 1/4, 1/2, or 3/4; Table 1; relative cost weights [28]). Each scenario was represented by its own series of 10 maps differing by resistance in open areas (see above).

2.6. Map Grain Size. Map grain can be an important issue in least-cost path modelling [31]. Thus, we ran all analyses with four map grain sizes. We converted each map layer from vector to raster form, with map grain sizes of 5, 25, 100, and 300 m. Small map grains led to uninterrupted forest roads and rivers, while coarser map grains created bridges across those linear features (Figure 4).
Figure 3: Three habitat resistance scenarios, applied to a selected part of the Montmorency Forest, Quebec, Canada.

Figure 4: Landscape matrix-grid in a portion of the Montmorency Forest, with varied grain. Map grain = 5 m (a), map grain = 25 m (b), map grain = 100 m (c) and map grain = 300 m (d). White: open area, light grey: young forest and dark grey: mature forest.
Table 1: Example of resistance values in young forest when the landscape layer is open/young/mature and the resistance value in the open is set to 12.

| Resistance ratio (Young forest : Open areas) | Resistance | Mature forest | Young forest | Open areas |
|---------------------------------------------|------------|---------------|--------------|------------|
| 1 : 4                                       | 1          | 3             | 12           |
| 1 : 2                                       | 1          | 6             | 12           |
| 3 : 4                                       | 1          | 9             | 12           |

Table 2: Model composition. Each line in the Table led to 10 cost surfaces, since 10 values were assigned to resistance in open areas for each of these combinations. Without null models, 200 cost surfaces were thus created.

| Map grain | Thematic resolution         | Ratio of the resistance in young forest compared to open areas |
|-----------|----------------------------|---------------------------------------------------------------|
| Open-forest | —                         | —                                                             |
| Open-edge-forest | —                 | —                                                             |
| 5 m    | Open-young-mature | 1/4                      |                                    |
|        | Open-young-mature | 1/2                      |                                    |
|        | Open-young-mature | 3/4                      |                                    |
| Open-forest | —                         | —                                                             |
| Open-edge-forest | —                 | —                                                             |
| 25 m    | Open-young-mature | 1/4                      |                                    |
|        | Open-young-mature | 1/2                      |                                    |
|        | Open-young-mature | 3/4                      |                                    |
| Open-forest | —                         | —                                                             |
| Open-edge-forest | —                 | —                                                             |
| 100 m   | Open-young-mature | 1/4                      |                                    |
|        | Open-young-mature | 1/2                      |                                    |
|        | Open-young-mature | 3/4                      |                                    |
| Open-forest | —                         | —                                                             |
| Open-edge-forest | —                 | —                                                             |
| 300 m   | Open-young-mature | 1/4                      |                                    |
|        | Open-young-mature | 1/2                      |                                    |
|        | Open-young-mature | 3/4                      |                                    |

2.7. Null Model. A null model was also generated, for which resistance values were set to one in all habitat types [32]. To summarize, we generated 201 resistance maps for each year, composed of 10 resistance levels in the open × 5 scenarios × 4 map grains, plus a null model (constant resistance) (Table 2).

2.8. Least-Cost Path Analysis. We used the least-cost path method [7] to simulate movements of American marten across cost surfaces, where grid cells are parameterized according to predicted costs of marten movement. ArcGIS 9.3 COSTDISTANCE and COSTPATH tools [33] were used to calculate the cumulative cost of each cell in the landscape and to link source and destination cells with the minimal cumulative cost. The aerial photos of the Montmorency forest are updated each year, which enables us to have year-specific maps. Every year and for each of the 201 resistance maps (varying thematic resolution, map grains, and cost assignments), we simulated marten movement paths connecting 100 random points along the western edge of the study area to 100 random points on the eastern edge of the study area. Simulated paths may represent the paths of several real martens, because some marten home-ranges (1–96.6 km², depending on conditions [34, 35]) cannot cover the width of the forest (3–16.5 km). Therefore, the method did not assume that the animal had perfect knowledge of the whole landscape.

2.9. Simulated versus Observed Track Locations. Once least-cost paths were obtained, we generated “simulated tracks” at each point where least-cost paths intersected a transect sampled in the field. Each cost surface was then tested by comparing location of its simulated tracks with location of our observed tracks. For each of these simulated tracks, we first calculated the distance to the nearest observed track (minimum distance). Second, we weighted each distance by the inverse of the number of hours passed between the last climatic disturbance in the forest (wind speed >20 km/h or snow fall >3 cm) and the time when the marten track was observed. Climatic data were recorded hourly by an automated weather station [24] and were consulted every day prior to working in the field. Had we not weighted distances between observed and simulated tracks, a long time between tracking and climatic disturbance would have artificially brought real tracks close to simulated tracks, because the number of tracks increases as the snowless days accumulate, thereby decreasing the mean distance between simulated and observed tracks.

2.10. Comparing Resistance Scenarios. The performance of each cost surface was evaluated from the 100 weighted-mean distances between real and simulated tracks that had been acquired with the 100 least-cost paths. Total length of covered transects varied between years, which led to a difference in the number of marten tracks recorded every year (Table 3). The years with a larger number of observed tracks resulted in smaller distances between simulated and observed tracks. To eliminate the year effect due to marten track density, performance of each model was compared with performance of the null model of the corresponding year. The performance of null model was measured as the median of the 100 weighted-mean distances between observed and

Table 3: Sampling effort of snow-tracking from 2004 to 2008.

| Year | Total transect length (km) | Number of marten tracks |
|------|----------------------------|-------------------------|
|      | On forestry roads and trails | Outside of roads and trails |
| 2004 | 89                        | 39                      | 259                   |
| 2005 | 86                        | 19                      | 260                   |
| 2006 | 138                       | 46                      | 291                   |
| 2007 | 116                       | 28                      | 247                   |
| 2008 | 145                       | 43                      | 610                   |
| Total| 574                       | 175                     | 1667                  |
simulated tracks. We used medians instead of means because the overall distribution of weighted-mean distances was skewed to the right (skew = 1.71, Agostino test: z = 22.8, P < 0.001). The performance of each of the 201 cost surfaces was expressed as the proportion of 500 least-cost paths (5 years, 100 paths in each year) that showed a mean distance that was smaller than the performance for the null model of the corresponding year. The performance of each simulated path (better than null model or not) was considered as an independent Bernoulli trial. Thus, we analyzed model performance as a proportion of models that were better than the null model with generalized linear models using a binomial error structure and logit link function (PROC GENMOD, SAS v. 9.1, Cary, NC, USA).

3. Results

Open areas, young and mature stands, respectively, covered 19%, 25%, and 56% in 2004 (variation of 1.6% max for each habitat type between 2004 and 2008). Each year we recorded locations of >247 marten tracks on over 100 km of transects (Table 3). Mean time without snow or strong winds before sampling was 95.7 h ± 29.8 h (N = 749 transects). Because transect length varied from year to year, the number of tracks that were simulated each year ranged from 131 791 to 240 643. Some simulated paths did not cross transects and, therefore, did not generate the creation of simulated tracks. For these models, the numbers of simulated paths that were used to estimate performance were less than 500.

3.1. Thematic Map Resolution. Varying thematic resolution affected the performance of least-cost path models. Of the 200 cost surfaces, 127 (63.5%) performed better than the corresponding null model, and 53 of these performed significantly better (P < 0.05). Cost surfaces with open and forest areas (OF), and simulations with forest, edge, and open areas (OEF) did not perform significantly better than null models. After accounting for map grain and movement resistance in open areas (Type 3 tests, SAS PROC GENMOD), models considering edges as less resistant to movements than “core” open areas did not perform better than models with edges as resistant as “core” open areas (linear contrast OF versus OEF, χ² < 0.01, df = 1, P > 0.9). After accounting for map grain and resistance in the open, adding movement resistance estimates that were specific to young stands significantly improved the performance of models (linear contrast, OF/OEF versus OYM simulations, χ² = 318.52, df = 1, P < 0.0001; see Table 4 for results with 100 m grain). After accounting for grain size and resistance in open areas, models with movement resistance in young stands set at 75% of that in open habitat performed significantly better than when the ratio was set at 25% (linear contrast, χ² = 4.73, df = 1, P = 0.03). Changes in relative movement cost in young stands had a very small impact on model performance compared to grain size (Figure 5). Simulations with resistance in young stands set at 50% of that in open areas yielded intermediate results (linear contrasts with 25% and 75% resistances: χ² < 1.5, df = 1, P > 0.22).

3.2. Map Grain Size. Map grain size had a substantial effect on movement path selection after accounting for edge, thematic resolution, and movement resistance in open canopy locations (Figure 5). Map grain size of 100 m performed better than those of 5, 25, or 300 m (linear contrasts: χ² > 80.0, df = 1, P < 0.0001). Simulations with grains of 25 m also performed better than those with grains of 5 m (χ² = 5.1, df = 1, P = 0.02), but the difference between 25 m and 5 m simulations was trivial (Figure 5). Grain size of 300 m did not perform better than any other grain size (linear contrasts with grains of 5 and 25 m: χ² < 1.50, df = 1, P > 0.2).

3.3. Movement Resistance in Open Areas. After accounting for forest edge, thematic map resolution, and map grain, model performance was the highest when hypothetical resistance to movement in open areas was 20 times higher than in mature forest (Figure 6). The latter resistance of 20× did

### Table 4: Percentage of the 500 simulation for which mean distance is smaller than the median of mean distances for null model of the corresponding year. Fractions in parentheses represent resistance in young forest relative to resistance in open areas. Percentages significantly greater than 50 (better than null model) are shown in bold typeface. Results with 100 m grain are shown.

| Resistance assigned to Open Areas | Open, forest | Open, edge, forest | Open, young (1/4), mature | Open, young (1/2), mature | Open, young (3/4), mature |
|----------------------------------|--------------|-------------------|--------------------------|--------------------------|--------------------------|
| 4                                | 51.1         | 49.7              | 49.7                     | 62.4                     | 64.4                     |
| 8                                | 49.7         | 49.9              | 61.8                     | 61.2                     | 61.6                     |
| 12                               | 49.7         | 49.5              | 62.2                     | 61.6                     | 62.0                     |
| 16                               | 49.7         | 49.7              | 61.4                     | 62.2                     | 61.4                     |
| 20                               | 49.9         | 49.7              | 61.8                     | 62.2                     | 61.6                     |
| 60                               | 51.5         | 49.5              | 61.4                     | 60.6                     | 61.0                     |
| 100                              | 49.7         | 49.7              | 60.4                     | 60.2                     | 60.6                     |
| 200                              | 49.3         | 49.9              | 60.2                     | 60.0                     | 59.8                     |
| 600                              | 49.5         | 48.7              | 59.6                     | 59.8                     | 60.0                     |
| 1000                             | 49.5         | 49.7              | 60.2                     | 59.8                     | 60.0                     |
not perform significantly better than the other resistances (linear contrasts, $P > 0.05$), but the lowest five resistances to movement, as a group, performed better than the highest five resistances (linear contrast, $\chi^2 = 4.5, df = 1, P < 0.03$). Using all models weighted by their performance, the mean estimate of movement resistance in open habitats was 45.5, but 95% confidence limits covered the entire range of hypothetical resistances, making this estimate unreliable.

4. Discussion

We showed that simple assumptions about marten movement can explain a nonnegligible part of the distribution of marten activity in a managed forest, particularly at coarse levels of map grain. Performance of thematic map resolutions varied significantly depending on assumptions about resistance to movement in open areas and regenerating forest stands. Robinson [36] suggested that American martens avoided venturing far into open areas in winter. He found that they never went more than 25 yards ($\approx 23$ m) into open habitats despite the presence of bait. Yet, we found that considering forest edges as less resistant to movement than areas far into open habitat did not improve simulation performance. The absence of a difference in resistance between open areas and edges may have arisen from a difference between movement resistance of roads and clear-cut edges, both of which were subsumed in the “edge group” in our study. In support of this idea, marten tracks were seldom recorded on narrow roadways (O. Planckaert, pers. obs.). However, these field impressions might have arisen from the high sampling effort on paths compared to that in open areas. A concurrent study in the Montmorency Forest found that American martens approached forest road and clear-cut edges similarly [37].

The allocation of an intermediate resistance to young forest (11 to 30 y-old stands) at a 100 m grain size approximated best the observed track locations. This result suggests that stands of young forest are not as permeable to marten movement as mature stands are. Alternatively, by including both young and mature forests in our model, we may have incorporated factors that were indirectly related to movement cost, such as prey densities specific to stand ages. In winter, martens hunt mostly large prey such as Northern flying squirrels (Glaucomys sabrinus), American red squirrels (Tamiasciurus hudsonicus), and snowshoe hares (Lepus americanus) [38] and feed on microtine rodents only incidentally [39]. In our study area, squirrels and hares occurred in similar densities in young, regenerating, and old forest stands [40] and thus prey density in forest is unlikely to have affected our results. However, the lack of foraging opportunities in open areas may have increased movement cost in these habitats.

Thematic maps with 100 m grain size better described landscape structure that is perceived by martens than did maps with finer or coarser grains. Broquet et al. [10] estimated least-cost paths with six grain sizes (10, 25, 50, 75, 100, and 500 m) to evaluate the effect of landscape structure on dispersal of American martens, and did not find significantly different results between them, except at 500 m for which performance was lower. In our study, maps with 100 m grain size led to discontinuous roads and simulated more realistic marten track locations than did finer and coarser map grains, probably because of the creation of “cracks” (or bridges) in the maps. Rothley [41] defined “cracks” as the shortcuts across costly linear barriers that were represented in a raster form. This suggests that forestry roads are highly permeable to marten movements. In a forest where road network density is particularly high, such as was the case here, these “cracks” may represent locations on narrow roads that martens use to move from one forest stand to another.
Moreover, although we did not take habitat width explicitly into consideration, the application of a 100 m grain to the land-use map may have created large forest patches and eliminated narrow forest patches, forcing simulated cost paths to travel through large patches of suitable habitat. Model performance obtained at this grain size suggests that marten use large patches for movement, as observed by Hargis et al. [12]. The coarser map grain of 300 m may have generated large forested patches and also could have resulted in wide-open areas that were unsuitable for martens and which impeded movement, confirming that marten is sensitive to connectivity of suitable habitat [22]. We conclude that the least-cost path approach must be sensitive to grain because of gap crossing and sensitivity to connectivity of high quality habitat.

Broquet et al. [10] applied a resistance value 50 times higher in unforest and open areas (<20 y-old) than in forest and showed that 50× cost was better than no differential cost. Model performance peaked in our analyses when resistance in open habitat was 20 times higher than in mature forest, but the location of this peak was unreliable. This result does not imply that open habitat is not resistant to movement marten relative to other landcover types. Instead, the Montmorency Forest might not be heterogeneous enough in terms of habitat types and configuration to allow us to evaluate precisely relative movement resistance in open areas compared to mature forests. In south central Wyoming (USA), Cushman et al. [22] concluded that nonforest cover was resistant to American marten movement when it fragmented landscape (24% clearcuts distributed heterogeneously). In a landscape that was highly dominated by closed-canopy forest, they showed that open areas were 16 times less avoided. Similarly, Schadt et al. [42] were unable to estimate movement resistance for Eurasian lynx (Lynx lynx) in central Europe and explained it as a lack of variability in landscape connectivity. Spatial variability of habitat features must affect their impact on movement and the power to detect these relationships [43]. Koen et al. [17], who found no significant effect of landscapes features on genetic distance, argued that open habitat is not very resistant to marten movement relative to other landcover types at province-wide scale (Ontario). They suggest that landscape features influencing marten movement are different among spatial scales [15].

We are aware that better models may exist which we did not test. Resistance values and map grain sizes were set arbitrarily, and nontested values could have led to higher least-cost path performance. However, resistance values for open areas could not be set higher because of the least-cost paths following forest borders, together with the range of grain sizes that was used based on forest-management scales. The simplification of map resolution to a maximum three classes might not be the strongest landscape model explaining marten movements [31]. We restricted our models to a few number of habitat categories to show that assumptions about two landscape criteria (e.g., forest age and edges) can be sufficient to model movement resistance of a pronounced habitat specialist like the American marten. Further research on marten movement resistance in more heterogeneous landscapes could lead to generalization of least-cost path factors estimates at the forest management scale (spatial replication; [43, 44]). Also, snow-tracking could be criticized, because it generates strongly heterogeneous sampling effort in space. Heterogeneous sampling effort could add bias to distances; where transects were rare, distances between simulated and real tracks were possibly higher. This problem was resolved here by comparing those distances rather than using absolute ones. One might argue that statistical analyses could also be biased due to the spatial clustering of tracks of the same individual, which could be recorded several times on the same transect. However, our sampling units were not single tracks, but least-cost paths, for which a single distance (fit) metric was calculated for a given set of parameters. We thus avoided pseudoreplication attributable to repeated tracks from the same individual.

Forest management scenarios that are based on stand quality differ greatly from scenarios based on access, with the former focusing mostly on tree attributes (species, size, snags, debris, etc.) and the latter focusing mostly on the configuration of stands. Our study contributes to forest management by giving weight to access and, thus, to stand configuration issues. We showed that presence-only data can aid in the understanding of movement costs for marten in the boreal forest and the parameterization of spatial models. The choice of the correct map grain size seems to be crucial in spatially explicit simulation models. Although it was probably not significant to population persistence in our study, matrix resistance might thwart animal movements through more fragmented landscapes. More studies about American marten movement at a forest management scale would be useful to refine these models in more heterogeneous landscapes.

Acknowledgments

Financial support for this work was provided by an NSERC (Natural Sciences and Engineering research Council of Canada) Discovery Grant to AD. The authors would like to thank all of the field assistants and former students who contributed to the collection of the 5-year dataset. Thanks are also due to the five anonymous reviewers who examined the manuscript in detail, and to Bill Parsons for a thorough review of the English of the revised version. Material support was possible in the field thanks to the employees of the Montmorency Forest, and the authors are grateful to them for their help. The authors of this paper do not have a direct financial relation with the commercial identities mentioned in this paper that might lead to a conflict of interest for any of the authors.

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