

1 Effects of the social environment on movement-integrated habitat selection

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8 Abstract

- 9 1. Movement links the distribution of habitats with the social environment of animals using
10 those habitats; yet integrating movement, habitat selection, and socioecology remains an
11 opportunity for further study.
- 12 2. Here, our objective was to disentangle the roles of habitat selection and social association
13 as drivers of collective movement in a gregarious ungulate. To accomplish this objective,
14 we (1) assessed whether socially familiar individuals form discrete social communities
15 and whether social communities have high spatial, but not necessarily temporal, overlap;
16 and (2) we modelled the relationship between collective movement and selection of
17 foraging habitats using socially informed integrated step selection analysis.
- 18 3. We used social network analysis to assign individuals to social communities and
19 determine short and long-term social preference among individuals. Using integrated step
20 selection functions (iSSF), we then modelled the effect of social processes, i.e., nearest
21 neighbour distance and social preference, and movement behaviour on patterns of habitat
22 selection.
- 23 4. Based on assignment of individuals to social communities and home range overlap
24 analyses, individuals assorted into discrete social communities, and these communities
25 had high spatial overlap. By unifying social network analysis with iSSF, we identified
26 movement-dependent social association, where individuals foraged with more familiar
27 individuals, but moved collectively with any between foraging patches.
- 28 5. Our study demonstrates that social behaviour and space use are inter-related based on
29 spatial overlap of social communities and movement-dependent habitat selection.
- 30 Movement, habitat selection, and social behaviour are linked in theory. Here, we put

31 these concepts into practice to demonstrate that movement is the glue connecting
32 individual habitat selection to the social environment.

33 **Keywords:** caribou, integrated step selection analysis, movement ecology, social preference,
34 social network analysis

35 1. Introduction

36 Movement is defined by a change in spatial location and is the behavioural link between the
37 physical space an animal occupies and the resources available to them (Van Moorter, Rolandsen,
38 Basille, & Gaillard, 2016). In the context of the social environment, movement represents the
39 connection between the distribution of resources and the social structure of animals that consume
40 those resources (He, Maldonado-Chaparro, & Farine, 2019). Disentangling the social and spatial
41 drivers of movement is a formidable challenge within behavioural ecology. In many cases,
42 research omits the social contexts within which animals move to, from, and within the areas that
43 contain foraging resources (Spiegel, Leu, Bull, & Sih, 2017; Strandburg-Peshkin, Papageorgiou,
44 Crofoot, & Farine, 2018). Spatially-explicit models of sociality highlight that some gregarious
45 species aggregate at areas associated with profitable foraging resources (Chamaillé-Jammes,
46 Fritz, Valeix, Murindagomo, & Clobert, 2008), whereas some territorial species only interact at
47 territory edges (Spiegel, Sih, Leu, & Bull, 2018). Sharing space, either at foraging sites, territory
48 edges, or elsewhere within an animal's range is required to form the social environment. For
49 example, animals are predicted to select habitat as a function of the profitability and availability
50 of the habitat (van Beest et al., 2014). A logical extension can be made to conspecifics;
51 individuals form groups based on their familiarity with conspecifics and the profitability of
52 associating with familiar conspecifics. We aim to quantify the relative importance of habitat and
53 conspecifics by developing a socially informed integrated step selection analysis, a movement-
54 based method that accounts for the relative intensity of selection for habitats and neighbours.

55 For social animals, individual movement shapes social encounters and subsequent
56 interactions with conspecifics that can affect collective movement (Jolles, King, & Killen, 2020).
57 Further complicating our understanding of collective movement is the idea that the type, quality,

58 and distribution of habitats on the landscape can constrain or promote collective movement
59 (Strandburg-Peshkin, Farine, Crofoot, & Couzin, 2017). For example, dense vegetation impedes
60 visibility, which could reduce the probability a group remains together. In addition, individual
61 movement and habitat selection are affected by the distribution of resources. For example,
62 patchily distributed foraging resources could facilitate large aggregations, whereas
63 homogenously distributed foraging resources could result in a reduction in social associations
64 (Spiegel, Leu, Bull, & Sih, 2017). The physical space an individual, or group, occupies and the
65 distribution and availability of foraging resources within that space are important drivers of
66 animal movement and the social environment an individual experiences (He et al., 2019).

67 Animals typically select for habitats that maximize foraging and minimize risk of
68 predation; an important trade-off because most habitats do not accommodate both high quality
69 foraging *and* low predation risk. When animals aggregate in large groups, the per capita risk of
70 predation is lower. Thus, animals in larger groups reduce time spent vigilant (Creel, Schuette, &
71 Christianson, 2014). Furthermore, individuals in larger groups tend to select more risky habitats,
72 including foraging in open areas (Lima, 1995). However, not all social groups are equal; some
73 groups contain unfamiliar individuals (i.e., anonymous groups) (Harel, Spiegel, Getz, & Nathan,
74 2017), while others contain familiar individuals (Lachlan, Crooks, & Laland, 1998). For
75 anonymous and familiar groups, social foraging occurs when the costs and benefits of an
76 individual's foraging behaviour are linked with the foraging behaviour of conspecifics
77 (Giraldeau & Dubois, 2008). Social foraging can be most beneficial when social information
78 about foraging resources comes from familiar individuals (Patin, Fortin, Sueur, & Chamaillé-
79 Jammes, 2019). For example, when foraging resources are unpredictable, familiar individuals
80 obtain reliable information from conspecifics to increase foraging efficiency (Jones, Patrick,

81 Evans, & Wells, 2020; Spiegel & Crofoot, 2016), such that time searching for forage is reduced
82 in favour of more time spent foraging. In the context of movement and habitat selection, theory
83 on social foraging and the benefits of social familiarity provides a framework through which the
84 costs and benefits of collective movement can be explored (Giraldeau & Caraco, 2018; Giraldeau
85 & Dubois, 2008).

86 Apparent social familiarity or preference is the long-term repeated social association due
87 to shared space at the same time. Although individuals often interact with many conspecifics,
88 non-random repeated social interactions or associations with certain individuals form the basis
89 for social preference (Mourier, Vercelloni, & Planes, 2012). Proximately, long-term social
90 relationships can influence collective movement via the reliability of information transfer about
91 foraging resources or predator risk (Best, Seddon, Dwyer, & Goldizen, 2013; Muller, Cantor,
92 Cuthill, & Harris, 2018), while ultimately they can enhance fitness (Silk, 2007). The social
93 environment can be influenced by the availability of foraging resources, but social communities
94 can also be composed of individuals with similar physiological or nutritional requirements that
95 occupy the same locations. Apparent social preference may therefore arise as a function of
96 spatial constraints (Spiegel, Leu, Sih, & Bull, 2016), including physical barriers, such as rivers or
97 mountains. Disentangling social preference from spatial constraint could inform our
98 understanding of collective movement and habitat selection (Croft, Darden, & Wey, 2016;
99 Pinter-Wollman et al., 2013).

100 Here, we develop a unified framework to bridge the gap between social network analysis
101 and movement ecology. We disentangle the roles of social preference and collective movement
102 on habitat selection behaviour by parameterizing socially informed integrated step selection
103 models (Figure 1). Animal social networks often comprise distinct sub-networks, or social

104 communities, defined by the existence of social preference among discrete clusters of individuals
105 (Mourier et al., 2012). Using a social ungulate as a model system, our objective was to
106 disentangle the roles of habitat selection and social association as drivers of collective movement
107 in a gregarious ungulate (*Rangifer tarandus*) when the availability and distribution of foraging
108 resources are variable. We calculated three distinct measures of social preference. First, we
109 assigned individuals to social communities based on a community detection algorithm. Second,
110 we assessed the temporal stability of social association among individuals. Third, we estimated
111 spatial overlap of social communities using home range analyses. Due to variance in the
112 distribution of foraging resources on the landscape, we expected that access to social information
113 via close proximity to conspecifics should influence patterns of selection for foraging resources.
114 Specifically, individuals with stronger social preference should select foraging habitat
115 collectively. The corollary is that individuals should also take short steps in the presence of
116 conspecifics, given that from a movement ecology perspective, shorter steps typically represent
117 foraging behaviour and longer steps represent searching behaviour (Owen-Smith, Fryxell, &
118 Merrill, 2010).

119 **2. Materials and Methods**

120 **2.1 Caribou as a model system**

121 We investigated patterns of movement, space use, and social behaviour for caribou
122 (*Rangifer tarandus*) on Fogo Island, Newfoundland, Canada. Fogo Island is a small (~237km²)
123 island off the northeastern coast of Newfoundland with a humid continental climate (see
124 Supplementary Materials for details). Between 1964-1967, 26 caribou were introduced to Fogo
125 Island from the Island of Newfoundland (Bergerud & Mercer, 1989). Currently, Fogo Island has
126 a population of approximately 300 caribou (Newfoundland and Labrador Wildlife Division,

127 unpublished data). Caribou live in fission-fusion societies (Lesmerises, Johnson, & St-Laurent,
128 2018), and throughout much of their range, caribou forage primarily on lichen, grasses, sedges,
129 and other deciduous browse with access to these resources changing between the seasons
130 (Bergerud, 1974). During winter (January to March), the landscape is covered by snow, and
131 caribou forage primarily on lichen (Webber, Ferraro, Hendrix, & Vander Wal, 2022). Lichen is
132 heterogeneously distributed, and access is impeded by snow and ice cover. Caribou dig holes in
133 the snow, termed craters, to access lichen in the winter, often where snow depth is relatively
134 shallow (~30–60 cm deep). Consequently, caribou have limited access to lichen buried under the
135 snow and tend to re-use established craters. To cope with this limitation, caribou use conspecific
136 attraction and social information transfer to gain access to foraging opportunities (Peignier et al.,
137 2019). In addition, caribou typically avoid forested habitats due to deep snow in forests and lack
138 of access to forage opportunities (Fortin, Courtois, Etcheverry, Dussault, & Gingras, 2008),
139 whereas most open habitats on Fogo Island are windswept in the winter, facilitating foraging and
140 movement (Bergerud, 1974).

141 We used GPS location data collected from Fogo Island caribou (2017–2019) to assess the
142 relationship between social behaviour, habitat selection, and movement (see supplementary
143 information for details on collaring procedures). For all analyses, we restricted locations to only
144 include relocations from the first 75 days of each year (1 January–16 March). Each relocation
145 was assigned to a given habitat classification that was extracted from Landsat images with 30m ×
146 30m pixels (Integrated-Informatics, 2014). Locations were categorized as one of open foraging
147 (lichen barrens), open moving (wetland, rocky outcrops, and water/ice), or forest (conifer scrub,
148 mixed wood, and conifer forest). We then calculated the proportion of each habitat type (i.e.,
149 open foraging, open moving, or forest) within 200 m around each used and available point

150 location (see below). Adult female caribou ($n = 26$ individual caribou, $n = 72$ caribou-years)
151 were immobilized and fitted with global positioning system (GPS) collars (Lotek Wireless Inc.,
152 Newmarket, ON, Canada, GPS4400M collars, 1,250 g). Prior to analyses, we removed all
153 erroneous and outlier GPS locations following Bjørneraas et al. (Bjørneraas, Van Moorter,
154 Rolandsen, & Herfindal, 2010). We did not collar all female caribou in the herds, however, and
155 collared individuals were randomly selected from the population. We therefore assume that our
156 sample of collared animals was randomly distributed. Although associations between collared
157 and uncollared animals were unrecorded, we assumed that our networks (see below) were
158 unbiased representations of the relative degree of social association among all caribou. All
159 animal captures and handling procedures were consistent with the American Society of
160 Mammologist guidelines and were approved by Memorial University Animal Use Protocol No.
161 20152067.

162 **2.2 Formulating integrated step selection models**

163 Integrated step selection function (iSSF) simultaneously incorporates movement and
164 habitat selection within a conditional logistic regression framework (Figure 1) (Avgar, Potts,
165 Lewis, & Boyce, 2016; Basille et al., 2015; Duchesne, Fortin, & Rivest, 2015). As in other
166 resource and step selection analyses (Fortin et al., 2005), iSSF models habitat selection as a
167 binomial response variable where ‘use’ represents the location an animal was observed and
168 ‘availability’ represents the geographical area an animal could potentially use but was not
169 necessarily observed (Figure S1). iSSF defines availability based on empirically fitted
170 distributions of step lengths and turn angles (Avgar et al., 2016), where a step is the linear
171 connection between consecutive relocations, and turn angle is the angular deviation between the
172 headings of two consecutive steps (Prokopenko, Boyce, & Avgar, 2017). We generated available

173 steps and turn angles based on the distributions informed by observed population-level
174 movement behaviour using the *amt* package in R (Signer, Fieberg, & Avgar, 2019). First, we
175 sampled step lengths from a gamma distribution of observed step lengths for the study
176 population; values were log-transformed for analysis. The statistical coefficient of log-
177 transformed step length is a modifier of the shape parameter from the gamma distribution
178 originally used to generate available steps (Avgar et al., 2016). Second, we sampled turn angles
179 (measured in radians) for available steps from observed values between $-\pi$ and π following a
180 Von Mises distribution. Each observed relocation was paired through a shared start point with 20
181 available steps generated from step-length and turn-angle distributions and compared in a
182 conditional logistic regression framework (see section 2.7). In addition to generating available
183 movement parameters, we also generated an available social environment (see below). To
184 evaluate the predictive performance of our model, we used k-fold ($k = 5$) cross validation
185 (Roberts et al., 2017) following the methods of Fortin et al. (2009). For details on k-fold cross
186 validation see Appendix 2.

187 **2.3 Social network analysis**

188 We used the R (R Core Team, 2019) packages *spatsoc* (Robitaille, Webber, & Vander
189 Wal, 2019) and *igraph* (Csárdi & Nepusz, 2006) to generate proximity-based social association
190 networks from GPS location data. Nodes in the networks represented individual caribou and
191 edges represented the frequency of association based on proximity between individuals. We
192 generated social networks at two scales based on proximity of locations between individual
193 caribou: (1) seasonal winter networks to assign individuals to social communities and assess
194 long-term social preference and (2) weekly networks to assess the role of short-term social
195 preference on patterns of habitat selection (section 2.2). Social communities represent a subset of

196 individuals within a network that are more closely connected with each other than with the rest
197 of the network. For networks at both seasonal and weekly scales, we assumed association
198 between two individuals when simultaneous locations (i.e. GPS relocations that occurred within
199 5 minutes of each other) were within 50 m of one another (Lesmerises et al., 2018; Peignier et
200 al., 2019). We selected the 50 m threshold based on the standard distance applied to assign
201 individuals to groups in studies of ungulate group size and social behaviour (Kasozi &
202 Montgomery, 2020). We applied the ‘chain rule’, where each discrete GPS fix was buffered by
203 50 m and we considered individuals in the same group if 50 m buffers for two or more
204 individuals were contiguous, even if some individuals were beyond 50 m of one another. We
205 weighted edges of social networks by the strength of association between dyads of caribou using
206 the simple ratio index (Cairns & Schwager, 1987), SRI:

$$SRI = \frac{x}{x + y_{AB}}$$

207 where x is the number of times individuals A and B were within 50 m of each other and y_{AB} is the
208 number of simultaneous fixes from individuals A and B that were separated by >50 m (Farine &
209 Whitehead, 2015).

210 **2.4 Detecting social communities: long-term social preference**

211 For seasonal winter social networks, we used a community detection algorithm to define
212 social communities (Newman, 2006). We assessed social community structure for each winter to
213 determine the broadest extent of social structure. Modularity is a commonly used measure that
214 defines how well-connected social communities are to one another. It is calculated from the
215 weighted proportion of edges that occur within a community, minus the expected proportion of
216 edges, if edges were distributed randomly in the network (Newman, 2006). A modularity value
217 close to 1 indicates a network with a strong clustered structure in which interactions of

218 individuals belonging to different clusters do not occur. We quantified modularity (Q) for
219 observed annual winter networks. To ensure observed social structure did not occur at random,
220 we compared these values to null models (Spiegel et al., 2016). Specifically, we generated null
221 models based on GPS fixes to reduce potential for type II error typically associated with node-
222 based permutations (Farine, 2014). Following Spiegel et al. (2016), we re-ordered daily GPS
223 movement trajectories for each individual while maintaining the temporal path sequence within
224 each time block (e.g., day 1 and day 2 may be swapped). This technique is a robust network
225 randomization procedure for GPS data because: 1) it maintains the spatial aspects of an
226 individual's movement; 2) by randomizing movement trajectories of individuals independent of
227 one another, temporal dependencies of movement are decoupled (Spiegel et al., 2016). We
228 repeated this procedure 100 times for annual winter networks and re-calculated modularity at
229 each iteration. We then compared observed modularity (Q) values to the null distribution and
230 determined whether the observed Q value fell within the 95% confidence interval of the
231 distribution of Q values (Mourier et al., 2012).

232 In addition to comparing observed Q values from annual winter networks to a null
233 distribution, we also calculated a community assortativity coefficient (R_{com}) to assess confidence
234 in the assignment of an individual to a given community (Shizuka & Farine, 2016). Specifically,
235 $R_{com} = 0$ indicates no confidence in the assignment of an individual to a community, while R_{com}
236 = 1 indicates certainty in the assignment of an individual to its community.

237 **2.5 Weekly networks and lagged association rates: short-term social preference**

238 We iteratively generated weekly social networks using a moving window approach and
239 calculated the observed SRI to be included as a covariate in our iSSF model (see section 2.2).
240 The first network was calculated for 1 January to 7 January, the second was 2 January to 8

241 January, and so on. Weekly networks contained 84 relocations per individual (12 relocations per
242 day). For each of these networks, we used dyadic values of SRI as a proxy for short-term social
243 preference. We used a three-step process. First, to incorporate SRI within the iSSF framework,
244 we determined the identity and distance (m) of each individual's nearest neighbour at each
245 relocation. Second, for each focal individual and their nearest neighbour at each relocation, we
246 matched the dyadic SRI value for the prior week. For example, for individual A at 12:00 on 8
247 January, we determined the nearest neighbour was individual B and we extracted the dyadic SRI
248 value for these individuals for the previous week. Third, we repeated steps one and two for all
249 'available' relocations defined by random steps generated in the iSSF (section 2.2). Therefore,
250 each individual at each relocation had an observed weekly dyadic SRI value and a series of
251 available weekly dyadic SRI values (see section 2.2).

252 In addition to incorporating social preference directly within the iSSF model, we also
253 assessed social preference by estimating within-season temporal patterns in associations between
254 individuals by calculating the lagged association rate (LAR). We calculated the LAR for social
255 networks using the *asnipe* package in R (Farine, 2013). LARs measure the probability that pairs
256 of individuals associating at a given relocation would still associate at subsequent relocations
257 (Whitehead, 2008). We generated annual LARs to compare temporal stability to assess potential
258 for within-season patterns of association among individuals. In addition, we also compared
259 seasonal LARs for individuals in the same annual winter social community to LARs for
260 individuals in different annual winter social communities to assess potential for within-season
261 patterns of association among individuals (Figure S4).

262 **2.6 Home range overlap between social communities**

263 To determine spatial overlap of social communities we estimated home ranges for winter

264 social communities using the area of the 95% isopleths from fixed kernel density estimates
265 (Worton, 1989) for each social community in each year with the *href* smoothing parameter in the
266 *adehabitatHR* package in R. Data from all individuals in a given social community were pooled
267 to estimate the community home range. We estimated home range overlap between social
268 communities with the utilization distribution overlap index (UDOI), where higher values of
269 UDOI represent a greater proportion of overlap and lower values represent lower proportion of
270 overlap (Fieberg & Kochanny, 2005).

271 **2.7 Modelling collective movement and habitat selection**

272 We fit a single iSSF model with a series of fixed and random effects using the *glmmTMB*
273 package in R following Muff et al. (Muff, Signer, & Fieberg, 2020). We took advantage of the
274 fact that the conditional logistic regression model is a likelihood-equivalent to a Poisson model
275 with stratum-specific fixed intercepts. The approach outlined by Muff et al. (2020) uses a mixed
276 modelling approach which allows intercepts and/or slopes to vary by individual, while also
277 incorporating shared information that is present in the data from different individuals (Fieberg,
278 Rieger, Zicus, & Schildcrout, 2009). For social species that may move collectively, and therefore
279 have correlated movement trajectories, varying intercepts by individual is recommended to
280 account for correlation within nested groupings of locations (Hebblewhite & Merrill, 2008).
281 Following Muff et al. (2020), all variables included in the fixed effect structure were also
282 included in the random effect structure. Our model included the proportion of lichen, forest, and
283 open habitat within 200 m of the point location, the natural log-transformed step length, natural
284 log-transformed nearest neighbour distance, and weekly dyadic simple ratio index (section 2.3).
285 Nearest neighbour distance (m) was measured as the distance between a focal individual and the
286 nearest collared conspecific and was calculated for all used and available steps. We also included

287 interactions between step length and each of the proportion of lichen, forest, and open habitats
288 within 200 m of the point location, nearest neighbour distance and step length, and simple ratio
289 index, nearest neighbour distance and each of the proportion of lichen, forest, and open habitats
290 within 200 m of the point location, and simple ratio index and each of the proportion of lichen,
291 forest, and open habitats within 200 m of the point location (see Table S1). For interactions that
292 included nearest neighbour distance, we used either distance at the start of a step or at the end of
293 the step, depending on the other variable in the interaction (Figure S1). Specifically, for the
294 interaction between step length and nearest neighbour distance, we used distance at the start of
295 the step because the likelihood of taking a shorter or longer step is predicted to vary based on the
296 distance to conspecifics before the step is taken. By contrast, for interactions between habitat
297 variables and nearest neighbour distance, we used distance at the end of the step because the
298 likelihood of selecting a given habitat is predicted to vary based on the distance to conspecifics
299 when that habitat is being selected, i.e., at the end of the step.

300 **2.8 Calculating effect sizes**

301 We calculated individual-level relative selection strength (RSS) to demonstrate how
302 habitat features influenced selection (Avgar, Lele, Keim, & Boyce, 2017). We calculated the
303 strength for selecting one step over another that differed in the habitat value where those steps
304 ended. RSS was calculated for each habitat type (i.e., forest, lichen, or open habitats) as a
305 function of nearest neighbour distance and the shared dyadic simple ratio index between nearest
306 neighbours.

307 **3. Results**

308 We found that individuals associated with members of multiple communities, and
309 associations were stronger among members of a given community. Depending on the year, social

310 networks comprised 2–6 social communities, and although community assortativity (R_{com}) was
311 similar across years, there was high certainty (range = 0.95–1.00) of an individual's assignment
312 to a given community in a given year (Table S1). In addition, lagged association rates (LAR)
313 within each winter confirmed temporal stability of community assortment, where association
314 rates for members of the same winter community remained higher than association rates for
315 members of different communities in each year (Figure 2). Seasonal winter values of modularity
316 (Q) were significantly lower than the distribution of Q generated from null models (Figure S2),
317 suggesting that social networks were structured weakly into communities with frequent inter-
318 community social associations (Table S1). In support of our expectation, we observed relatively
319 high spatial overlap between different winter social communities (average UDOI = 0.37, SD =
320 0.34, range = 0–0.98; Figure S3; Table S2), thus facilitating the potential for association between
321 social communities.

322 Overall, we found that caribou are highly social in nearly all circumstances and that
323 caribou prefer to select all habitats with familiar conspecifics (Figure 4). Despite these findings,
324 the effect of the social environment on selection was nuanced, and we found partial support for
325 our expectation of social foraging. Individuals moved more slowly when selecting lichen and
326 when they shared a high SRI value with their nearest neighbour, suggesting potential that
327 conspecific familiarity influenced foraging-related movement (Table S2). However, relative to
328 its availability, caribou moved more quickly through open habitat, perhaps to travel between
329 foraging sites (Figure 3). Meanwhile, relative selection strength for all habitats decayed as
330 nearest neighbours were further away, however, relative selection for lichen habitat was stronger
331 than forest and open habitats (Figure 4). Our k-fold cross-validation had high scores ($\rho = 0.80$

332 SE \pm 0.06), demonstrating our model was better than random at predicting where caribou moved
333 (see Figure S6 for coefficients for variables in each fold).

334 **4. Discussion**

335 Our study examined apparent social preference in the context of shared space use using socially
336 informed integrated step selection functions. We present a framework that unifies social
337 networks within a traditional movement ecology and habitat selection framework. Although
338 individual social associations were well mixed at the population level, we found that social
339 networks were structured into discrete communities. Despite spatial overlap between different
340 social communities, which suggests an opportunity for individuals to interact with members of
341 other communities, we highlight two forms of within-community social preference, including
342 long-term temporal stability of associations among individuals, and an effect of short-term social
343 preference on habitat selection. Further, we found that individuals tended to select foraging
344 habitat near familiar individuals but moved between foraging habitats with conspecifics
345 regardless of their degree of familiarity, suggesting the social environment can vary relative to
346 the speed animals are moving. The processes underlying community structure appear to be
347 social, and not spatial. Based on our unification of social network analysis with integrated step
348 selection functions, we highlight the influence of collective movement and preferred associations
349 on habitat selection and foraging.

350 Testing social preference as a driver of movement and habitat selection required
351 establishing the existence of discrete communities and long-term social associations within the
352 population-level network. Indeed, the formation of social communities, in combination with our
353 lagged association analysis, confirmed the existence of temporal stability in social associations
354 for members of the same social community. The loose formation of non-random social

355 communities is consistent with expectations of fission-fusion dynamics, where groups merge and
356 split through space and time (Sueur et al., 2011). Community formation was driven in part by
357 social preference, but aspects of space use, including shared space, could also influence the
358 formation of social communities, even if they are relatively weak (Daizaburo Shizuka et al.,
359 2014). We found high spatial overlap between social communities, suggesting that physical
360 barriers on the landscape do not explain the formation of discrete social communities. For social
361 communities to emerge from a well-mixed population, individuals in different communities must
362 have high spatial, but low temporal overlap in shared geographical space, thus revealing the
363 importance of space and time in the formation of social communities (Cantor et al., 2012).
364 Disentangling space and time within the social environment reveals distinct social communities
365 and groups of individuals that are more likely to associate than by chance (Spiegel et al., 2016).
366 On resource limited landscapes, individuals are expected to aggregate in close proximity to those
367 resources, for example, elephants (*Loxodonta africana*) aggregate near water-holes, which are a
368 limiting resource (Chamaillé-Jammes et al., 2008). At the population-level, social networks were
369 highly connected, thus providing the impetus to quantify socially informed patterns in movement
370 and habitat selection.

371 Our findings reveal that caribou are social in nearly all circumstances, although we
372 observed a social hierarchy of movement-dependent social associations. Specifically, individuals
373 tended to select to be close to familiar nearest neighbours when moving slowly and, in general,
374 selected to be closer to nearest neighbours in lichen habitat relative to forest and open habitats
375 regardless of the familiarity of nearest neighbours. Within the movement ecology literature for
376 ungulates, there is an assumption that slower movement in a given habitat represents foraging
377 behaviour and faster movement represents searching behaviour (Owen-Smith et al., 2010). Our

378 results support this assumption. Individuals moved more slowly in lichen habitat and moved
379 more quickly in open habitats. Within a social context, individuals appear to collectively move
380 through open habitat with familiar individuals, perhaps to new foraging patches. Individuals are
381 more likely to trust social information about food sources and predation risk from familiar
382 individuals, but the potential costs are an increase in competition at foraging patches. Individuals
383 may balance the trade-off between competition and access to information by moving with
384 socially familiar individuals but spacing apart during foraging. Lichen habitat is typically open,
385 suggesting the possibility that individuals may remain in visual and vocal contact, thereby
386 facilitating social cohesion during foraging despite physically spacing apart (Jacobs, 2010). This
387 type of movement-dependent social association could contribute to the maintenance of social
388 communities described above. Our results are also corroborated by other ungulate systems. In
389 bison (*Bison bison*), the social environment in combination with recent knowledge of local
390 foraging options dictated whether individuals followed, or left, a group (Merkle, Sigaud, &
391 Fortin, 2015). Moreover, in the bison system, the costs and benefits of foraging in a group are
392 moderated by collective decision making (Sigaud et al., 2017) and collective movement (Courant
393 & Fortin, 2012), both of which are likely involved in the foraging decisions made by caribou.
394 Here, we elucidate potential behavioural mechanisms (i.e., foraging or moving) that influence the
395 frequency and magnitude of social associations.

396 The emergent geometry of collective movement and spatial arrangement of individuals in
397 a group appears to change as individuals adjust their behaviour based on the availability of
398 resources and the presence of familiar conspecifics (Morrell, Ruxton, & James, 2011). Assamese
399 macaques (*Macaca assamensis*) distance from one another during foraging, but move
400 collectively between foraging sites (Heesen, Macdonald, Ostner, & Schülke, 2015), while

401 individual giraffes (*Giraffa camelopardalis*) show social preference for conspecifics during
402 foraging, but not during movement (Muller et al., 2018). Interestingly, macaques foraged in
403 closer proximity to individuals of similar dominance rank, but for giraffes it was unclear whether
404 observed social preference was the result of passive or active assortment. For caribou,
405 dominance hierarchies are linear and typically driven by body size (Barrette & Vandal, 1986),
406 suggesting that social preference in caribou could also be related to dominance. Our ability to
407 delineate aspects of the social environment between collective movement and habitat selection
408 within a unified framework is useful for disentangling passive or active assortment, for example
409 dominance rank, conspecific attraction, or the transfer of information about foraging resources.

410 We assumed that moving with familiar conspecifics is the result of information transfer
411 about the location or quality of cratering sites, but spacing apart during foraging occurs because
412 competition among individual caribou for craters in the winter can be substantial (Barrette &
413 Vandal, 1986). Moreover, selection for open habitat relative to its availability in groups could
414 also reflect the use of social information about the location of foraging sites (Lesmerises et al.,
415 2018) or predation (Hamilton, 1971). Craters can vary in size and distribution (Bergerud, 1974);
416 however, craters may only be large enough for a single individual to forage at a time (Mayor,
417 Schaefer, Schneider, & Mahoney, 2009). Foraging apart from conspecifics reduce the costs of
418 competition at cratering sites, which may be limited on the landscape or relatively small. We
419 propose that while caribou generally have larger group sizes in winter (Webber & Vander Wal,
420 2021), groups vary in size based on movement and habitat selection behaviour presumably to
421 balance the trade-off between competition and information acquisition. Furthermore, female
422 caribou often have antlers, which unlike males, persist into winter. Females are hypothesized to
423 use their antlers to defend craters and exert dominance over both males and females without

424 antlers (Barrette & Vandal, 1986; Schaefer & Mahoney, 2001). This interpretation is
425 corroborated by theory used to explain fission-fusion dynamics, where individuals are expected
426 to split and merge through space and time to reduce conflict and competition during foraging.

427 We demonstrate assortment of individuals into distinct social communities, despite high
428 range overlap with individuals in other communities. Integrating space and time revealed fine-
429 scale processes that form social communities and the socially mediated nature of movement
430 ecology and habitat selection. Within a unified socially informed integrated step selection
431 framework, we bridge the theoretical and methodological gap between social network analysis,
432 movement ecology, and habitat selection. We also demonstrate how social association is context-
433 dependent, where individuals forage spaced apart from one another, but move collectively with
434 familiar between foraging patches. Our synthesis of integrated step selection functions with
435 social networks to test hypotheses is an important step towards identifying the roles of physical
436 space and animal space use as factors influencing the social environment (Strandburg-Peshkin et
437 al., 2017). Moreover, individual variation in phenotypes attributable to movement or habitat
438 selection may affect how individuals experience the social environment (Webber et al., 2022;
439 Webber & Vander Wal, 2018). Movement, habitat selection, and social behaviour are clearly
440 linked; as van Moorter et al. (2016) described movement as the ‘glue’ connecting habitat
441 selection to the physical location of a given set of habitats, we posit that movement is the glue
442 connecting collective habitat selection to the social environment.

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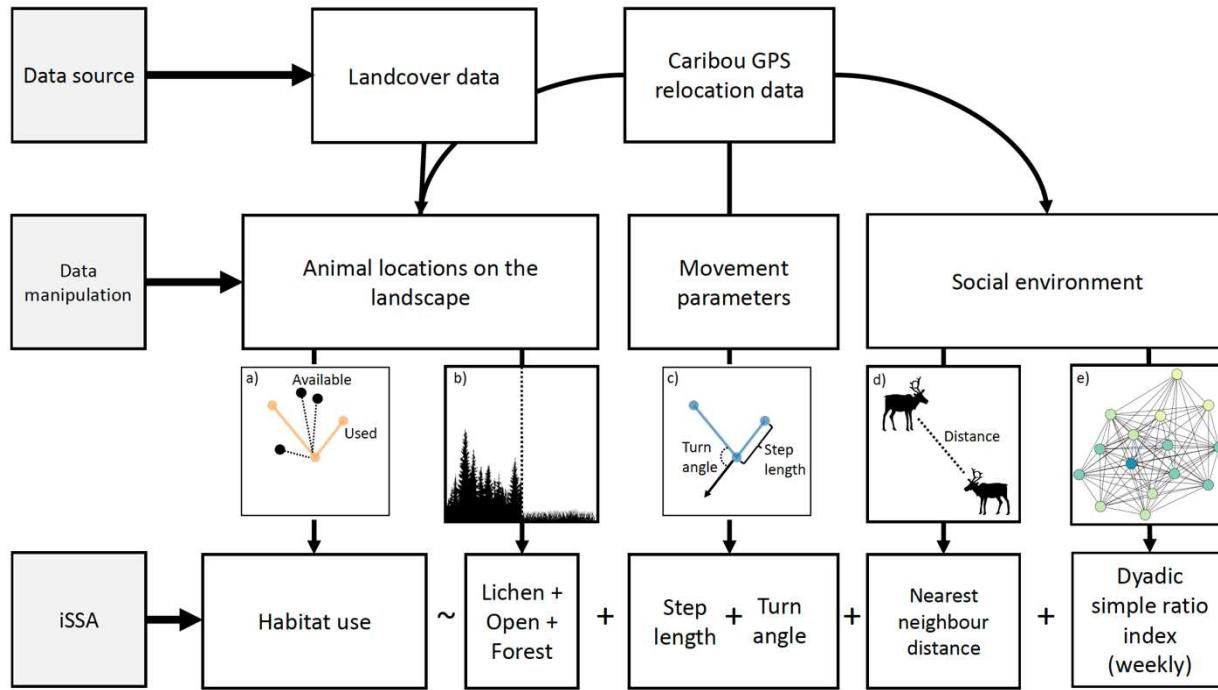
455 **DATA AVAILABILITY STATEMENT**

456 All code and data used for statistical analysis and figures are archived in Zenodo:
457 doi:10.5281/zenodo.4549509

458 **AUTHORS CONTRIBUTIONS**

459 QMRW and EVW conceived the ideas and designed method; QMRW conducted fieldwork
460 QMRW and generated social networks; QMRW, CMP, KAK, and JWT conducted movement
461 and spatial analysis; QMRW led the writing of the manuscript; EVW developed the research
462 program. All authors contributed critically to drafts and gave final approval for publication.

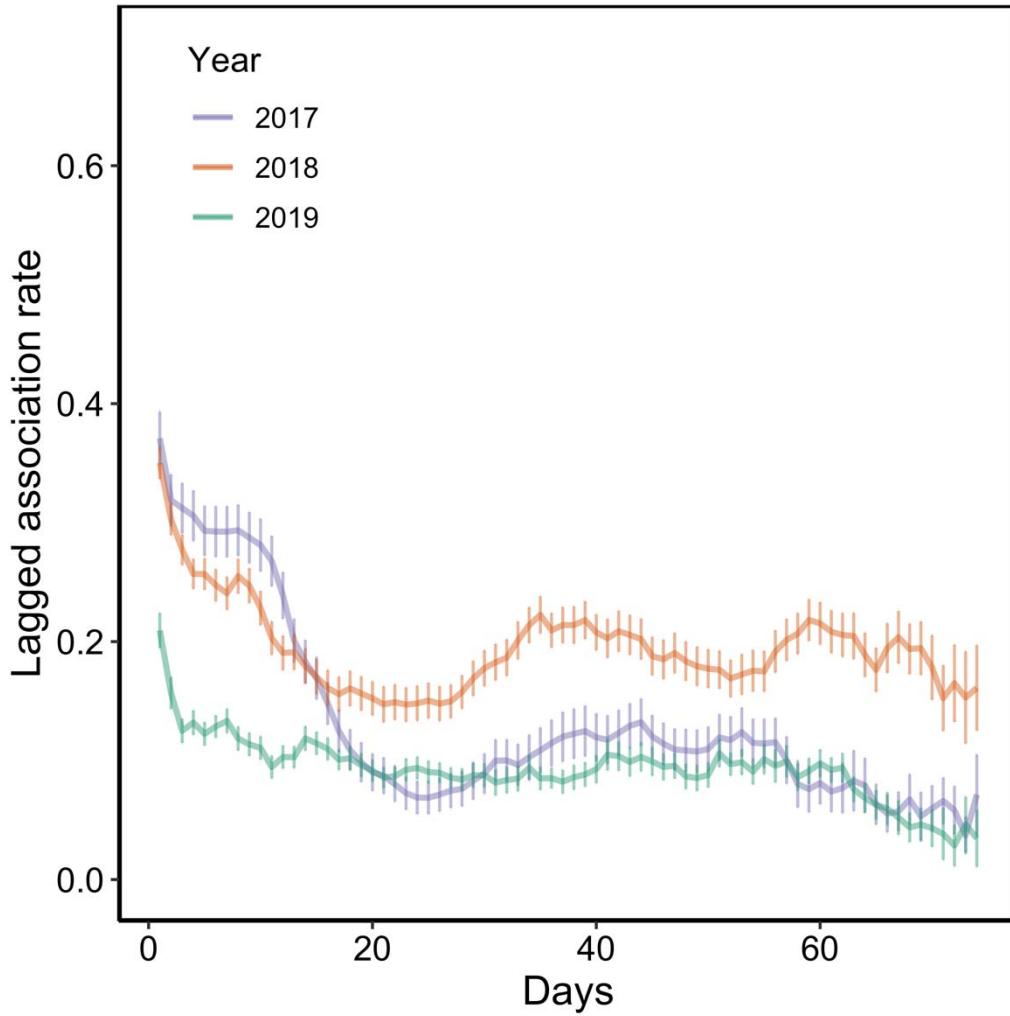
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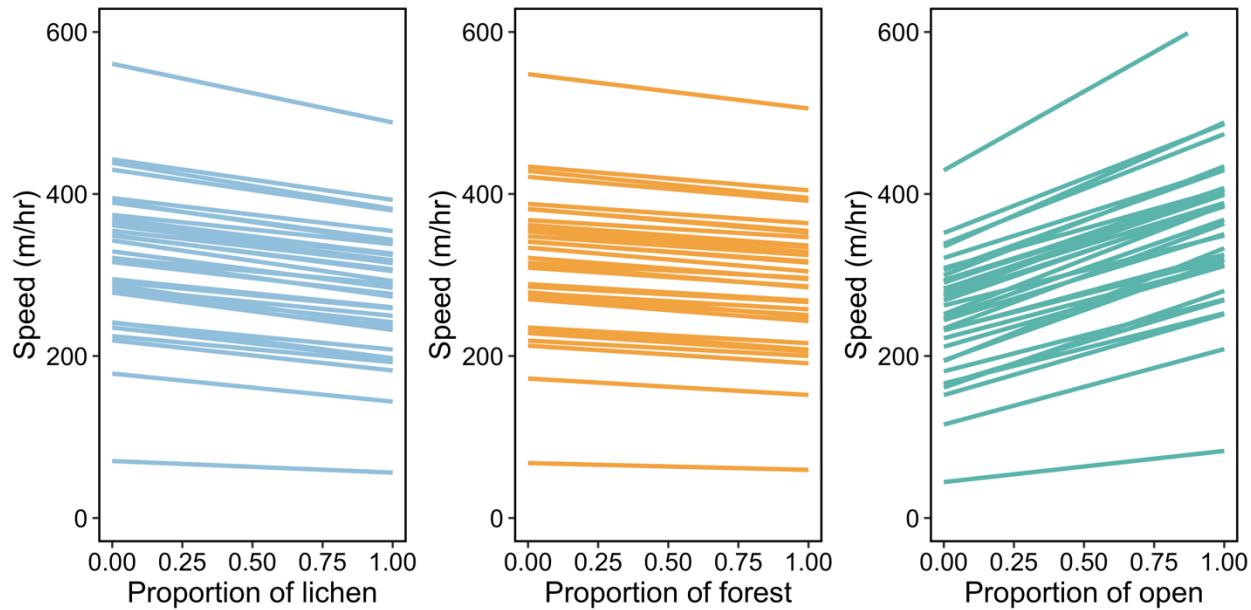
465 **Figure 1.** Summary of the data pipeline used to generate integrated step selection function
466 (iSSF) models. Primary data sources were landcover data and caribou GPS relocation data,
467 which were combined to determine the physical locations of animals on the landscape. The
468 pairing of animal locations and landcover data was used to generate the comparison of used to
469 available points (panel a), which is the response variable in iSSF models, as well as the habitat
470 type in which a given relocation occurred: lichen (defined in text as open-forage), open (defined
471 in text as open-movement), and forest (panel b). Caribou relocation data were also used to
472 generate two movement parameters (panel c) and aspects of the social environment (panels d and
473 f). Movement parameters included turn angle, which is the angular deviation between the
474 headings of two consecutive steps, and step length, which is the linear distance between
475 consecutive relocations. The social environment included nearest neighbour distance (panel d)
476 and weekly social networks and the dyadic simple ratio index generated based on a moving-
477 window as a proxy for short-term social preference (panel e). The bottom row represents a

478 graphical formulation of our iSSF models, where habitat selection (1:10 ratio of used to available
479 relocations) was regressed against habitat type (lichen, open, and forest), movement parameters
480 (step length and turn angle), nearest neighbour distance, and weekly dyadic simple ratio index.
481



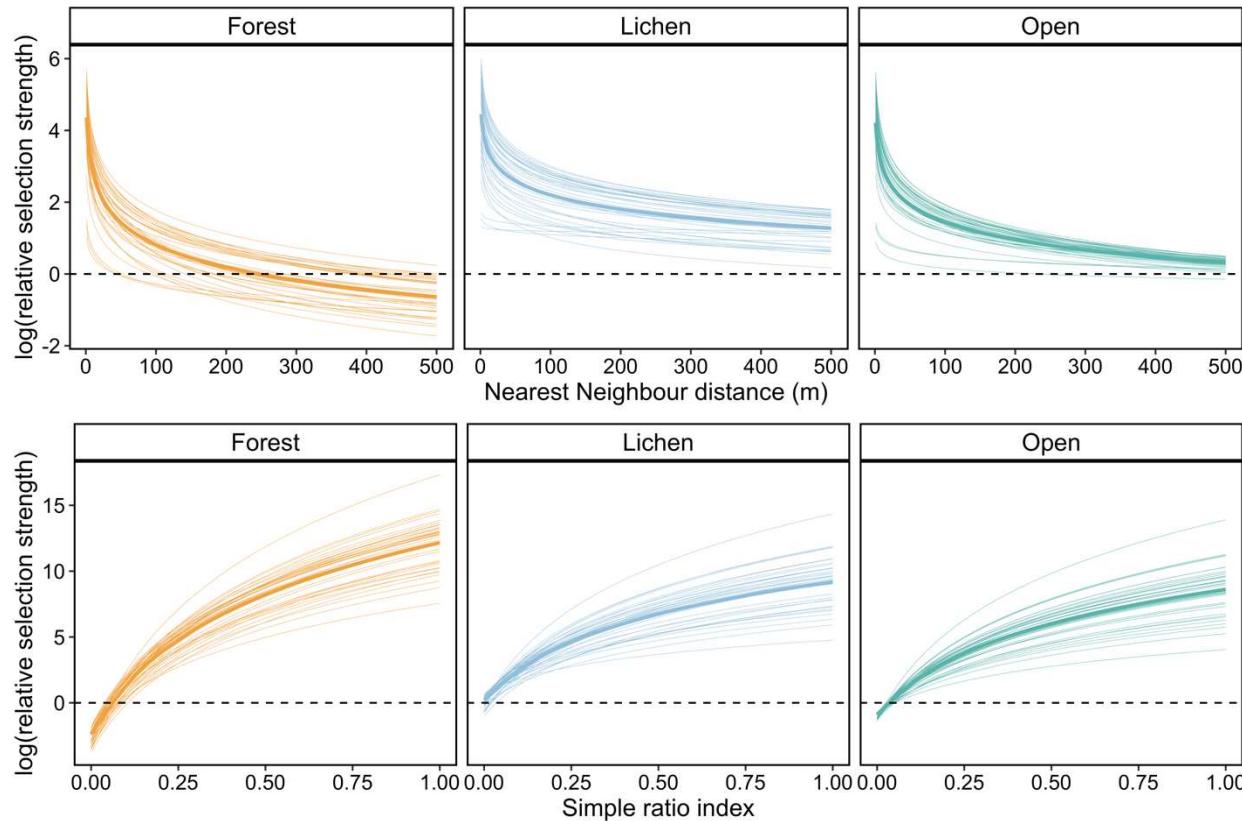
482

483 **Figure 2.** Observed annual lagged association rate (LAR) for caribou, calculated as the
484 probability that any pair of individuals associated on a given day, are still associated on
485 subsequent days. Note, the time period for LAR analysis was 1 January to 16 March. Error bars
486 represent the standard error of all pairwise association rates calculated on each day.



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Figure 3. Relationship between expected speed (m/hr) as a function of changes in the proportion of lichen, forest, and open habitats within 200 m of a given point location for individual caribou.



491

492 **Figure 4.** Relative selection strength of forest, lichen, and open habitats as a function of nearest
493 neighbour distance (m) (top panels) and shared dyadic simple ration index between nearest
494 neighbours (bottom panels). The dotted horizontal line represents no response, while values
495 above the line indicate the population is selecting to be closer to that habitat than expected or to
496 have higher shared dyadic simple ratio index than expected, and below the dotted line the
497 population is selecting to be farther from that habitat than expected or to have a lower shared
498 dyadic simple ratio index than expected. Interpretation for RSS values are that individuals
499 generally tend to select to be near to conspecifics when selecting lichen and open habitats
500 relative to their availability, whereas the response to nearest neighbours in forest habitat relative
501 to its availability is limited. Meanwhile, individuals select for familiar nearest neighbours in all
502 habitats.

503 References

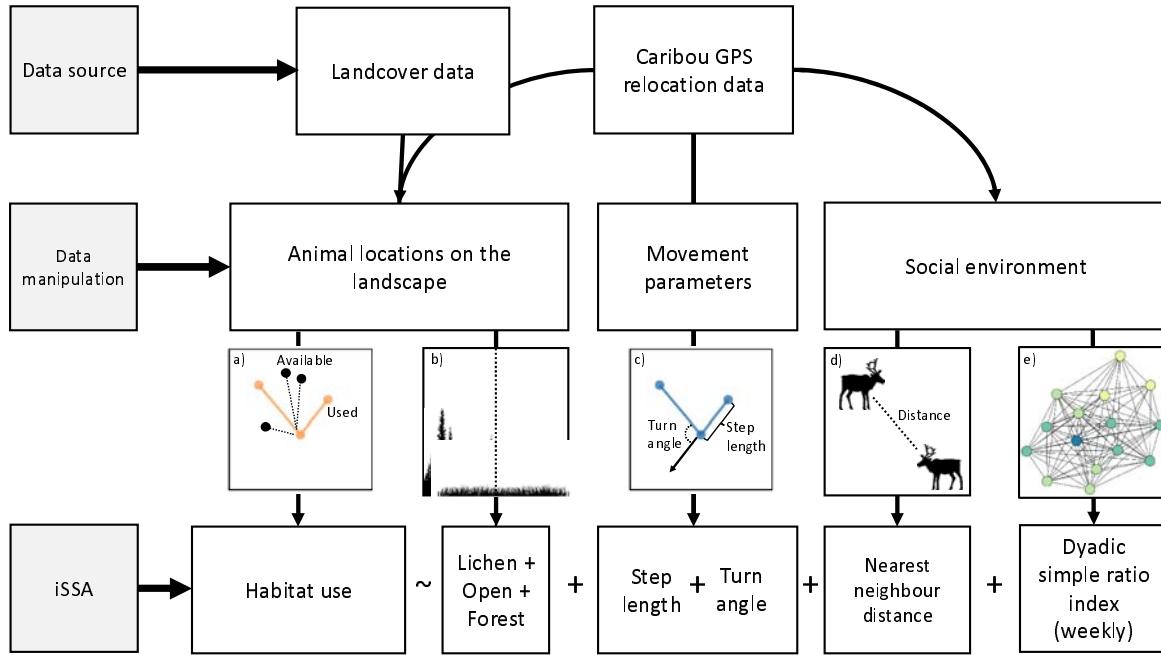
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714 Figure 1.