Effects of the social environment on movement-integrated habitat selection

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Abstract

1. Movement links the distribution of habitats with the social environment of animals using those habitats; yet integrating movement, habitat selection, and socioecology remains an opportunity for further study.

2. Here, our objective was to disentangle the roles of habitat selection and social association as drivers of collective movement in a gregarious ungulate. To accomplish this objective, we (1) assessed whether socially familiar individuals form discrete social communities and whether social communities have high spatial, but not necessarily temporal, overlap; and (2) we modelled the relationship between collective movement and selection of foraging habitats using socially informed integrated step selection analysis.

3. We used social network analysis to assign individuals to social communities and determine short and long-term social preference among individuals. Using integrated step selection functions (iSSF), we then modelled the effect of social processes, i.e., nearest neighbour distance and social preference, and movement behaviour on patterns of habitat selection.

4. Based on assignment of individuals to social communities and home range overlap analyses, individuals assorted into discrete social communities, and these communities had high spatial overlap. By unifying social network analysis with iSSF, we identified movement-dependent social association, where individuals foraged with more familiar individuals, but moved collectively with any between foraging patches.

5. Our study demonstrates that social behaviour and space use are inter-related based on spatial overlap of social communities and movement-dependent habitat selection. 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

1. Introduction

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

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

and distribution of habitats on the landscape can constrain or promote collective movement

(Strandburg-Peshkin, Farine, Crofoot, & Couzin, 2017). For example, dense vegetation impedes

visibility, which could reduce the probability a group remains together. In addition, individual

movement and habitat selection are affected by the distribution of resources. For example,

patchily distributed foraging resources could facilitate large aggregations, whereas

homogenously distributed foraging resources could result in a reduction in social associations

(Spiegel, Leu, Bull, & Sih, 2017). The physical space an individual, or group, occupies and the

distribution and availability of foraging resources within that space are important drivers of

animal movement and the social environment an individual experiences (He et al., 2019).

 Animals typically select for habitats that maximize foraging and minimize risk of

predation; an important trade-off because most habitats do not accommodate both high quality

foraging *and* low predation risk. When animals aggregate in large groups, the per capita risk of

predation is lower. Thus, animals in larger groups reduce time spent vigilant (Creel, Schuette, &

Christianson, 2014). Furthermore, individuals in larger groups tend to select more risky habitats,

including foraging in open areas (Lima, 1995). However, not all social groups are equal; some

groups contain unfamiliar individuals (i.e., anonymous groups) (Harel, Spiegel, Getz, & Nathan,

2017), while others contain familiar individuals (Lachlan, Crooks, & Laland, 1998). For

anonymous and familiar groups, social foraging occurs when the costs and benefits of an

individual's foraging behaviour are linked with the foraging behaviour of conspecifics

(Giraldeau & Dubois, 2008). Social foraging can be most beneficial when social information

about foraging resources comes from familiar individuals (Patin, Fortin, Sueur, & Chamaillé-

Jammes, 2019). For example, when foraging resources are unpredictable, familiar individuals

obtain reliable information from conspecifics to increase foraging efficiency (Jones, Patrick,

Evans, & Wells, 2020; Spiegel & Crofoot, 2016), such that time searching for forage is reduced

in favour of more time spent foraging. In the context of movement and habitat selection, theory

on social foraging and the benefits of social familiarity provides a framework through which the

costs and benefits of collective movement can be explored (Giraldeau & Caraco, 2018; Giraldeau

& Dubois, 2008).

Apparent social familiarity or preference is the long-term repeated social association due

to shared space at the same time. Although individuals often interact with many conspecifics,

non-random repeated social interactions or associations with certain individuals form the basis

for social preference (Mourier, Vercelloni, & Planes, 2012). Proximately, long-term social

relationships can influence collective movement via the reliability of information transfer about

foraging resources or predator risk (Best, Seddon, Dwyer, & Goldizen, 2013; Muller, Cantor,

Cuthill, & Harris, 2018), while ultimately they can enhance fitness (Silk, 2007). The social

environment can be influenced by the availability of foraging resources, but social communities

can also be composed of individuals with similar physiological or nutritional requirements that

occupy the same locations. Apparent social preference may therefore arise as a function of

spatial constraints (Spiegel, Leu, Sih, & Bull, 2016), including physical barriers, such as rivers or

mountains. Disentangling social preference from spatial constraint could inform our

understanding of collective movement and habitat selection (Croft, Darden, & Wey, 2016;

Pinter-Wollman et al., 2013).

Here, we develop a unified framework to bridge the gap between social network analysis

and movement ecology. We disentangle the roles of social preference and collective movement

on habitat selection behaviour by parameterizing socially informed integrated step selection

models (Figure 1). Animal social networks often comprise distinct sub-networks, or social

communities, defined by the existence of social preference among discrete clusters of individuals

(Mourier et al., 2012). Using a social ungulate as a model system, our objective was to

disentangle the roles of habitat selection and social association as drivers of collective movement

in a gregarious ungulate (*Rangifer tarandus*) when the availability and distribution of foraging

resources are variable. We calculated three distinct measures of social preference. First, we

assigned individuals to social communities based on a community detection algorithm. Second,

we assessed the temporal stability of social association among individuals. Third, we estimated

spatial overlap of social communities using home range analyses. Due to variance in the

distribution of foraging resources on the landscape, we expected that access to social information

via close proximity to conspecifics should influence patterns of selection for foraging resources.

Specifically, individuals with stronger social preference should select foraging habitat

collectively. The corollary is that individuals should also take short steps in the presence of

conspecifics, given that from a movement ecology perspective, shorter steps typically represent

foraging behaviour and longer steps represent searching behaviour (Owen-Smith, Fryxell, &

Merrill, 2010).

2. Materials and Methods

2.1 Caribou as a model system

We investigated patterns of movement, space use, and social behaviour for caribou

(*Rangifer tarandus*) on Fogo Island, Newfoundland, Canada. Fogo Island is a small (~237km^2)

island off the northeastern coast of Newfoundland with a humid continental climate (see

Supplementary Materials for details). Between 1964-1967, 26 caribou were introduced to Fogo

Island from the Island of Newfoundland (Bergerud & Mercer, 1989). Currently, Fogo Island has

a population of approximately 300 caribou (Newfoundland and Labrador Wildlife Division,

unpublished data). Caribou live in fission-fusion societies (Lesmerises, Johnson, & St-Laurent,

2018), and throughout much of their range, caribou forage primarily on lichen, grasses, sedges,

and other deciduous browse with access to these resources changing between the seasons

(Bergerud, 1974). During winter (January to March), the landscape is covered by snow, and

caribou forage primarily on lichen (Webber, Ferraro, Hendrix, & Vander Wal, 2022). Lichen is

heterogeneously distributed, and access is impeded by snow and ice cover. Caribou dig holes in

the snow, termed craters, to access lichen in the winter, often where snow depth is relatively

shallow (~30–60 cm deep). Consequently, caribou have limited access to lichen buried under the

snow and tend to re-use established craters. To cope with this limitation, caribou use conspecific

attraction and social information transfer to gain access to foraging opportunities (Peignier et al.,

2019). In addition, caribou typically avoid forested habitats due to deep snow in forests and lack

of access to forage opportunities (Fortin, Courtois, Etcheverry, Dussault, & Gingras, 2008),

whereas most open habitats on Fogo Island are windswept in the winter, facilitating foraging and

movement (Bergerud, 1974).

We used GPS location data collected from Fogo Island caribou (2017–2019) to assess the

relationship between social behaviour, habitat selection, and movement (see supplementary

information for details on collaring procedures). For all analyses, we restricted locations to only

include relocations from the first 75 days of each year (1 January–16 March). Each relocation

was assigned to a given habitat classification that was extracted from Landsat images with 30m $\times$

30m pixels (Integrated-Informatics, 2014). Locations were categorized as one of open foraging

(lichen barrens), open moving (wetland, rocky outcrops, and water/ice), or forest (conifer scrub,

mixed wood, and conifer forest). We then calculated the proportion of each habitat type (i.e.,

open foraging, open moving, or forest) within 200 m around each used and available point

location (see below). Adult female caribou (n = 26 individual caribou, n = 72 caribou-years)

were immobilized and fitted with global positioning system (GPS) collars (Lotek Wireless Inc.,

Newmarket, ON, Canada, GPS4400M collars, 1,250 g). Prior to analyses, we removed all

erroneous and outlier GPS locations following Bjørneraas et al. (Bjørneraas, Van Moorter,

Rolandsen, & Herfindal, 2010). We did not collar all female caribou in the herds, however, and

collared individuals were randomly selected from the population. We therefore assume that our

sample of collared animals was randomly distributed. Although associations between collared

and uncollared animals were unrecorded, we assumed that our networks (see below) were

unbiased representations of the relative degree of social association among all caribou. All

animal captures and handling procedures were consistent with the American Society of

Mammologist guidelines and were approved by Memorial University Animal Use Protocol No.

20152067.

2.2 Formulating integrated step selection models

Integrated step selection function (iSSF) simultaneously incorporates movement and

habitat selection within a conditional logistic regression framework (Figure 1) (Avgar, Potts,

Lewis, & Boyce, 2016; Basille et al., 2015; Duchesne, Fortin, & Rivest, 2015). As in other

resource and step selection analyses (Fortin et al., 2005), iSSF models habitat selection as a

binomial response variable where 'use' represents the location an animal was observed and

'availability' represents the geographical area an animal could potentially use but was not

necessarily observed (Figure S1). iSSF defines availability based on empirically fitted

distributions of step lengths and turn angles (Avgar et al., 2016), where a step is the linear

connection between consecutive relocations, and turn angle is the angular deviation between the

headings of two consecutive steps (Prokopenko, Boyce, & Avgar, 2017). We generated available

steps and turn angles based on the distributions informed by observed population-level

movement behaviour using the *amt* package in R (Signer, Fieberg, & Avgar, 2019). First, we

sampled step lengths from a gamma distribution of observed step lengths for the study

population; values were log-transformed for analysis. The statistical coefficient of log-

transformed step length is a modifier of the shape parameter from the gamma distribution

originally used to generate available steps (Avgar et al., 2016). Second, we sampled turn angles

(measured in radians) for available steps from observed values between $-\pi$ and π following a

Von Mises distribution. Each observed relocation was paired through a shared start point with 20

available steps generated from step-length and turn-angle distributions and compared in a

conditional logistic regression framework (see section 2.7). In addition to generating available

movement parameters, we also generated an available social environment (see below). To

evaluate the predictive performance of our model, we used k-fold (k = 5) cross validation

(Roberts et al., 2017) following the methods of Fortin et al. (2009). For details on k-fold cross

validation see Appendix 2.

2.3 Social network analysis

We used the R (R Core Team, 2019) packages *spatsoc* (Robitaille, Webber, & Vander

Wal, 2019) and *igraph* (Csárdi & Nepusz, 2006) to generate proximity-based social association

networks from GPS location data. Nodes in the networks represented individual caribou and

edges represented the frequency of association based on proximity between individuals. We

generated social networks at two scales based on proximity of locations between individual

caribou: (1) seasonal winter networks to assign individuals to social communities and assess

long-term social preference and (2) weekly networks to assess the role of short-term social

preference on patterns of habitat selection (section 2.2). Social communities represent a subset of

individuals within a network that are more closely connected with each other than with the rest

of the network. For networks at both seasonal and weekly scales, we assumed association

between two individuals when simultaneous locations (i.e. GPS relocations that occurred within

5 minutes of each other) were within 50 m of one another (Lesmerises et al., 2018; Peignier et

al., 2019). We selected the 50 m threshold based on the standard distance applied to assign

individuals to groups in studies of ungulate group size and social behaviour (Kasozi &

Montgomery, 2020). We applied the 'chain rule', where each discrete GPS fix was buffered by

50 m and we considered individuals in the same group if 50 m buffers for two or more

individuals were contiguous, even if some individuals were beyond 50 m of one another. We

weighted edges of social networks by the strength of association between dyads of caribou using

the simple ratio index (Cairns & Schwager, 1987), SRI:

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

where x is the number of times individuals A and B were within 50 m of each other and y_{AB} is the

number of simultaneous fixes from individuals A and B that were separated by >50 m (Farine &

Whitehead, 2015).

2.4 Detecting social communities: long-term social preference

For seasonal winter social networks, we used a community detection algorithm to define

social communities (Newman, 2006). We assessed social community structure for each winter to

determine the broadest extent of social structure. Modularity is a commonly used measure that

defines how well-connected social communities are to one another. It is calculated from the

weighted proportion of edges that occur within a community, minus the expected proportion of

edges, if edges were distributed randomly in the network (Newman, 2006). A modularity value

close to 1 indicates a network with a strong clustered structure in which interactions of

individuals belonging to different clusters do not occur. We quantified modularity (Q) for

observed annual winter networks. To ensure observed social structure did not occur at random,

we compared these values to null models (Spiegel et al., 2016). Specifically, we generated null

models based on GPS fixes to reduce potential for type II error typically associated with node-

based permutations (Farine, 2014). Following Spiegel et al. (2016), we re-ordered daily GPS

movement trajectories for each individual while maintaining the temporal path sequence within

each time block (e.g., day 1 and day 2 may be swapped). This technique is a robust network

randomization procedure for GPS data because: 1) it maintains the spatial aspects of an

individual's movement; 2) by randomizing movement trajectories of individuals independent of

one another, temporal dependencies of movement are decoupled (Spiegel et al., 2016). We

repeated this procedure 100 times for annual winter networks and re-calculated modularity at

each iteration. We then compared observed modularity (Q) values to the null distribution and

determined whether the observed Q value fell within the 95% confidence interval of the

distribution of Q values (Mourier et al., 2012).

In addition to comparing observed Q values from annual winter networks to a null

distribution, we also calculated a community assortativity coefficient (R_{com}) to assess confidence

in the assignment of an individual to a given community (Shizuka & Farine, 2016). Specifically,

$R_{com} = 0$ indicates no confidence in the assignment of an individual to a community, while R_{com}

$= 1$ indicates certainty in the assignment of an individual to its community.

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

We iteratively generated weekly social networks using a moving window approach and

calculated the observed SRI to be included as a covariate in our iSSF model (see section 2.2).

The first network was calculated for 1 January to 7 January, the second was 2 January to 8

January, and so on. Weekly networks contained 84 relocations per individual (12 relocations per

day). For each of these networks, we used dyadic values of SRI as a proxy for short-term social

preference. We used a three-step process. First, to incorporate SRI within the iSSF framework,

we determined the identity and distance (m) of each individual's nearest neighbour at each

relocation. Second, for each focal individual and their nearest neighbour at each relocation, we

matched the dyadic SRI value for the prior week. For example, for individual A at 12:00 on 8

January, we determined the nearest neighbour was individual B and we extracted the dyadic SRI

value for these individuals for the previous week. Third, we repeated steps one and two for all

'available' relocations defined by random steps generated in the iSSF (section 2.2). Therefore,

each individual at each relocation had an observed weekly dyadic SRI value and a series of

available weekly dyadic SRI values (see section 2.2).

In addition to incorporating social preference directly within the iSSF model, we also

assessed social preference by estimating within-season temporal patterns in associations between

individuals by calculating the lagged association rate (LAR). We calculated the LAR for social

networks using the *asnipe* package in R (Farine, 2013). LARs measure the probability that pairs

of individuals associating at a given relocation would still associate at subsequent relocations

(Whitehead, 2008). We generated annual LARs to compare temporal stability to assess potential

for within-season patterns of association among individuals. In addition, we also compared

seasonal LARs for individuals in the same annual winter social community to LARs for

individuals in different annual winter social communities to assess potential for within-season

patterns of association among individuals (Figure S4).

2.6 Home range overlap between social communities

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

social communities using the area of the 95% isopleths from fixed kernel density estimates

(Worton, 1989) for each social community in each year with the *href* smoothing parameter in the

adehabitatHR package in R. Data from all individuals in a given social community were pooled

to estimate the community home range. We estimated home range overlap between social

communities with the utilization distribution overlap index (UDOI), where higher values of

UDOI represent a greater proportion of overlap and lower values represent lower proportion of

overlap (Fieberg & Kochanny, 2005).

2.7 Modelling collective movement and habitat selection

We fit a single iSSF model with a series of fixed and random effects using the *glmmTMB*

package in R following Muff et al. (Muff, Signer, & Fieberg, 2020). We took advantage of the

fact that the conditional logistic regression model is a likelihood-equivalent to a Poisson model

with stratum☐specific fixed intercepts. The approach outlined by Muff et al. (2020) uses a mixed

modelling approach which allows intercepts and/or slopes to vary by individual, while also

incorporating shared information that is present in the data from different individuals (Fieberg,

Rieger, Zicus, & Schildcrout, 2009). For social species that may move collectively, and therefore

have correlated movement trajectories, varying intercepts by individual is recommended to

account for correlation within nested groupings of locations (Hebblewhite & Merrill, 2008).

Following Muff et al. (2020), all variables included in the fixed effect structure were also

included in the random effect structure. Our model included the proportion of lichen, forest, and

open habitat within 200 m of the point location, the natural log-transformed step length, natural

log-transformed nearest neighbour distance, and weekly dyadic simple ratio index (section 2.3).

Nearest neighbour distance (m) was measured as the distance between a focal individual and the

nearest collared conspecific and was calculated for all used and available steps. We also included

interactions between step length and each of the proportion of lichen, forest, and open habitats

within 200 m of the point location, nearest neighbour distance and step length, and simple ratio

index, nearest neighbour distance and each of the proportion of lichen, forest, and open habitats

within 200 m of the point location, and simple ratio index and each of the proportion of lichen,

forest, and open habitats within 200 m of the point location (see Table S1). For interactions that

included nearest neighbour distance, we used either distance at the start of a step or at the end of

the step, depending on the other variable in the interaction (Figure S1). Specifically, for the

interaction between step length and nearest neighbour distance, we used distance at the start of

the step because the likelihood of taking a shorter or longer step is predicted to vary based on the

distance to conspecifics before the step is taken. By contrast, for interactions between habitat

variables and nearest neighbour distance, we used distance at the end of the step because the

likelihood of selecting a given habitat is predicted to vary based on the distance to conspecifics

when that habitat is being selected, i.e., at the end of the step.

2.8 Calculating effect sizes

We calculated individual-level relative selection strength (RSS) to demonstrate how

habitat features influenced selection (Avgar, Lele, Keim, & Boyce, 2017). We calculated the

strength for selecting one step over another that differed in the habitat value where those steps

ended. RSS was calculated for each habitat type (i.e., forest, lichen, or open habitats) as a

function of nearest neighbour distance and the shared dyadic simple ratio index between nearest

neighbours.

3. Results

We found that individuals associated with members of multiple communities, and

associations were stronger among members of a given community. Depending on the year, social

networks comprised 2–6 social communities, and although community assortativity (R_{com}) was

similar across years, there was high certainty (range = 0.95–1.00) of an individual's assignment

to a given community in a given year (Table S1). In addition, lagged association rates (LAR)

within each winter confirmed temporal stability of community assortment, where association

rates for members of the same winter community remained higher than association rates for

members of different communities in each year (Figure 2). Seasonal winter values of modularity

(Q) were significantly lower than the distribution of Q generated from null models (Figure S2),

suggesting that social networks were structured weakly into communities with frequent inter-

community social associations (Table S1). In support of our expectation, we observed relatively

high spatial overlap between different winter social communities (average UDOI = 0.37, SD =

0.34, range = 0–0.98; Figure S3; Table S2), thus facilitating the potential for association between

social communities.

Overall, we found that caribou are highly social in nearly all circumstances and that

caribou prefer to select all habitats with familiar conspecifics (Figure 4). Despite these findings,

the effect of the social environment on selection was nuanced, and we found partial support for

our expectation of social foraging. Individuals moved more slowly when selecting lichen and

when they shared a high SRI value with their nearest neighbour, suggesting potential that

conspecific familiarity influenced foraging-related movement (Table S2). However, relative to

its availability, caribou moved more quickly through open habitat, perhaps to travel between

foraging sites (Figure 3). Meanwhile, relative selection strength for all habitats decayed as

nearest neighbours were further away, however, relative selection for lichen habitat was stronger

than forest and open habitats (Figure 4). Our k-fold cross-validation had high scores (rho = 0.80

SE ± 0.06), demonstrating our model was better than random at predicting where caribou moved

(see Figure S6 for coefficients for variables in each fold).

4. Discussion

Our study examined apparent social preference in the context of shared space use using socially

informed integrated step selection functions. We present a framework that unifies social

networks within a traditional movement ecology and habitat selection framework. Although

individual social associations were well mixed at the population level, we found that social

networks were structured into discrete communities. Despite spatial overlap between different

social communities, which suggests an opportunity for individuals to interact with members of

other communities, we highlight two forms of within-community social preference, including

long-term temporal stability of associations among individuals, and an effect of short-term social

preference on habitat selection. Further, we found that individuals tended to select foraging

habitat near familiar individuals but moved between foraging habitats with conspecifics

regardless of their degree of familiarity, suggesting the social environment can vary relative to

the speed animals are moving. The processes underlying community structure appear to be

social, and not spatial. Based on our unification of social network analysis with integrated step

selection functions, we highlight the influence of collective movement and preferred associations

on habitat selection and foraging.

Testing social preference as a driver of movement and habitat selection required

establishing the existence of discrete communities and long-term social associations within the

population-level network. Indeed, the formation of social communities, in combination with our

lagged association analysis, confirmed the existence of temporal stability in social associations

for members of the same social community. The loose formation of non-random social

communities is consistent with expectations of fission-fusion dynamics, where groups merge and

split through space and time (Sueur et al., 2011). Community formation was driven in part by

social preference, but aspects of space use, including shared space, could also influence the

formation of social communities, even if they are relatively weak (Daizaburo Shizuka et al.,

2014). We found high spatial overlap between social communities, suggesting that physical

barriers on the landscape do not explain the formation of discrete social communities. For social

communities to emerge from a well-mixed population, individuals in different communities must

have high spatial, but low temporal overlap in shared geographical space, thus revealing the

importance of space and time in the formation of social communities (Cantor et al., 2012).

Disentangling space and time within the social environment reveals distinct social communities

and groups of individuals that are more likely to associate than by chance (Spiegel et al., 2016).

On resource limited landscapes, individuals are expected to aggregate in close proximity to those

resources, for example, elephants (*Loxodonta africana*) aggregate near water-holes, which are a

limiting resource (Chamaillé-Jammes et al., 2008). At the population-level, social networks were

highly connected, thus providing the impetus to quantify socially informed patterns in movement

and habitat selection.

Our findings reveal that caribou are social in nearly all circumstances, although we

observed a social hierarchy of movement-dependent social associations. Specifically, individuals

tended to select to be close to familiar nearest neighbours when moving slowly and, in general,

selected to be closer to nearest neighbours in lichen habitat relative to forest and open habitats

regardless of the familiarity of nearest neighbours. Within the movement ecology literature for

ungulates, there is an assumption that slower movement in a given habitat represents foraging

behaviour and faster movement represents searching behaviour (Owen-Smith et al., 2010). Our

results support this assumption. Individuals moved more slowly in lichen habitat and moved

more quickly in open habitats. Within a social context, individuals appear to collectively move

through open habitat with familiar individuals, perhaps to new foraging patches. Individuals are

more likely to trust social information about food sources and predation risk from familiar

individuals, but the potential costs are an increase in competition at foraging patches. Individuals

may balance the trade-off between competition and access to information by moving with

socially familiar individuals but spacing apart during foraging. Lichen habitat is typically open,

suggesting the possibility that individuals may remain in visual and vocal contact, thereby

facilitating social cohesion during foraging despite physically spacing apart (Jacobs, 2010). This

type of movement-dependent social association could contribute to the maintenance of social

communities described above. Our results are also corroborated by other ungulate systems. In

bison (*Bison bison*), the social environment in combination with recent knowledge of local

foraging options dictated whether individuals followed, or left, a group (Merkle, Sigaud, &

Fortin, 2015). Moreover, in the bison system, the costs and benefits of foraging in a group are

moderated by collective decision making (Sigaud et al., 2017) and collective movement (Courant

& Fortin, 2012), both of which are likely involved in the foraging decisions made by caribou.

Here, we elucidate potential behavioural mechanisms (i.e., foraging or moving) that influence the

frequency and magnitude of social associations.

The emergent geometry of collective movement and spatial arrangement of individuals in

a group appears to change as individuals adjust their behaviour based on the availability of

resources and the presence of familiar conspecifics (Morrell, Ruxton, & James, 2011). Assamese

macaques (*Macaca assamensis*) distance from one another during foraging, but move

collectively between foraging sites (Heesen, Macdonald, Ostner, & Schülke, 2015), while

individual giraffes (*Giraffa camelopardalis*) show social preference for conspecifics during

foraging, but not during movement (Muller et al., 2018). Interestingly, macaques foraged in

closer proximity to individuals of similar dominance rank, but for giraffes it was unclear whether

observed social preference was the result of passive or active assortment. For caribou,

dominance hierarchies are linear and typically driven by body size (Barrette & Vandal, 1986),

suggesting that social preference in caribou could also be related to dominance. Our ability to

delineate aspects of the social environment between collective movement and habitat selection

within a unified framework is useful for disentangling passive or active assortment, for example

dominance rank, conspecific attraction, or the transfer of information about foraging resources.

We assumed that moving with familiar conspecifics is the result of information transfer

about the location or quality of cratering sites, but spacing apart during foraging occurs because

competition among individual caribou for craters in the winter can be substantial (Barrette &

Vandal, 1986). Moreover, selection for open habitat relative to its availability in groups could

also reflect the use of social information about the location of foraging sites (Lesmerises et al.,

2018) or predation (Hamilton, 1971). Craters can vary in size and distribution (Bergerud, 1974);

however, craters may only be large enough for a single individual to forage at a time (Mayor,

Schaefer, Schneider, & Mahoney, 2009). Foraging apart from conspecifics reduce the costs of

competition at cratering sites, which may be limited on the landscape or relatively small. We

propose that while caribou generally have larger group sizes in winter (Webber & Vander Wal,

2021), groups vary in size based on movement and habitat selection behaviour presumably to

balance the trade-off between competition and information acquisition. Furthermore, female

caribou often have antlers, which unlike males, persist into winter. Females are hypothesized to

use their antlers to defend craters and exert dominance over both males and females without

antlers (Barrette & Vandal, 1986; Schaefer & Mahoney, 2001). This interpretation is

corroborated by theory used to explain fission-fusion dynamics, where individuals are expected

to split and merge through space and time to reduce conflict and competition during foraging.

We demonstrate assortment of individuals into distinct social communities, despite high

range overlap with individuals in other communities. Integrating space and time revealed fine-

scale processes that form social communities and the socially mediated nature of movement

ecology and habitat selection. Within a unified socially informed integrated step selection

framework, we bridge the theoretical and methodological gap between social network analysis,

movement ecology, and habitat selection. We also demonstrate how social association is context-

dependent, where individuals forage spaced apart from one another, but move collectively with

familiar between foraging patches. Our synthesis of integrated step selection functions with

social networks to test hypotheses is an important step towards identifying the roles of physical

space and animal space use as factors influencing the social environment (Strandburg-Peshkin et

al., 2017). Moreover, individual variation in phenotypes attributable to movement or habitat

selection may affect how individuals experience the social environment (Webber et al., 2022;

Webber & Vander Wal, 2018). Movement, habitat selection, and social behaviour are clearly

linked; as van Moorter et al. (2016) described movement as the 'glue' connecting habitat

selection to the physical location of a given set of habitats, we posit that movement is the glue

connecting collective habitat selection to the social environment.

ACKNOWLEDGEMENTS

We respectfully acknowledge the territory in which data were collected and analyzed as the

ancestral homelands of the Beothuk and the Island of Newfoundland as the ancestral homelands

of the Mi'kmaq and Beothuk. We thank M. Laforge, M. Bonar, J. Hendrix, and R. Huang for

help in the field, and D. Wilson, K. Lewis, I. Fleming, G. Albery, members of the Wildlife

Evolutionary Ecology Lab, and three anonymous reviewers for helpful comments on previous

versions of this manuscript. We also thank members of the Newfoundland and Labrador Wildlife

Division, including S. Moores, B. Adams, W. Barney, and J. Neville for facilitating animal

captures and for logistical support in the field. Funding for this study was provided by Natural

Sciences and Engineering Research Council (NSERC) Vanier Canada Graduate Scholarships to

QMRW and CMP, a NSERC Canada Graduate Scholarship (Masters and Doctoral) to KAK, and

a NSERC Discovery Grant to EVW.

DATA AVAILABILITY STATEMENT

All code and data used for statistical analysis and figures are archived in Zenodo:

doi:10.5281/zenodo.4549509

AUTHORS CONTRIBUTIONS

QMRW and EVW conceived the ideas and designed method; QMRW conducted fieldwork

QMRW and generated social networks; QMRW, CMP, KAK, and JWT conducted movement

and spatial analysis; QMRW led the writing of the manuscript; EVW developed the research

program. All authors contributed critically to drafts and gave final approval for publication.

Figure 1. Summary of the data pipeline used to generate integrated step selection function

(iSSF) models. Primary data sources were landcover data and caribou GPS relocation data,

which were combined to determine the physical locations of animals on the landscape. The

pairing of animal locations and landcover data was used to generate the comparison of used to

available points (panel a), which is the response variable in iSSF models, as well as the habitat

type in which a given relocation occurred: lichen (defined in text as open-forage), open (defined

in text as open-movement), and forest (panel b). Caribou relocation data were also used to

generate two movement parameters (panel c) and aspects of the social environment (panels d and

f). Movement parameters included turn angle, which is the angular deviation between the

headings of two consecutive steps, and step length, which is the linear distance between

consecutive relocations. The social environment included nearest neighbour distance (panel d)

and weekly social networks and the dyadic simple ratio index generated based on a moving-

window as a proxy for short-term social preference (panel e). The bottom row represents a

graphical formulation of our iSSF models, where habitat selection (1:10 ratio of used to available

relocations) was regressed against habitat type (lichen, open, and forest), movement parameters

(step length and turn angle), nearest neighbour distance, and weekly dyadic simple ratio index.

Figure 2. Observed annual lagged association rate (LAR) for caribou, calculated as the

probability that any pair of individuals associated on a given day, are still associated on

subsequent days. Note, the time period for LAR analysis was 1 January to 16 March. Error bars

represent the standard error of all pairwise association rates calculated on each day.

487
488 **Figure 3.** Relationship between expected speed (m/hr) as a function of changes in the proportion

489 of lichen, forest, and open habitats within 200 m of a given point location for individual caribou.

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

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713
714   Figure 1.