1   Foraging with friends, moving with anyone: effects of the social environment on movement-

2   integrated habitat selection

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Abstract

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. 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. 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 movement ecology, we identified movement-dependent social association, where

individuals foraged with more familiar individuals, but moved collectively with any between

foraging patches. 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 these

concepts into practice to demonstrate that movement is the glue connecting individual habitat

selection to the social environment.

Keywords: caribou, integrated step selection analysis, movement ecology, social preference,

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 [1]. 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 [2]. 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 [but see 3,4]. Spatially-explicit models of sociality highlight that some gregarious species aggregate at areas associated with profitable foraging resources [5], whereas some territorial species only interact at territory edges [6]. 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 [7]. 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 [8]. 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 [9]. 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

[10]. 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 [2].

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 [11]. Furthermore,

individuals in larger groups tend to select more risky habitats, including foraging in open areas

[12]. However, not all social groups are equal; some groups contain unfamiliar individuals (i.e.,

anonymous groups) [13], while others contain familiar individuals [14]. 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 [15]. Social foraging can be

most beneficial when social information about foraging resources comes from familiar

individuals [16]. For example, when foraging resources are unpredictable, familiar individuals

obtain reliable information from conspecifics to increase foraging efficiency [17,18], 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 [15,19].

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 [20]. Proximately, long-term social relationships can influence collective

movement via the reliability of information transfer about foraging resources or predator risk

[21,22], while ultimately they can enhance fitness [23]. 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 [24], including physical barriers, such as rivers or mountains. Disentangling social

preference from spatial constraint could inform our understanding of collective movement and

habitat selection [25,26].

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

[20]. 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 [27].

2. Materials and Methods

(a) 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 [28]. Currently, Fogo Island has a population of

approximately 300 caribou (Newfoundland and Labrador Wildlife Division, unpublished data).

Caribou live in fission-fusion societies [29], 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 [30]. During winter (January to March), the landscape is covered

by snow, and caribou forage primarily on lichen. 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 [31]. In winter, caribou activity

budgets suggest that caribou spend ~50% of their time foraging, while ~40% of their time is

spent lying down or ruminating, 7% of their time is spent walking or trotting, and 3% of their

time is spent standing [32,33]. In addition, caribou typically avoid forested habitats due to deep

snow in forests and lack of access to forage opportunities [34], whereas most open habitats on

Fogo Island are windswept in the winter, facilitating foraging and movement [30].

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 [35]. 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. [36]. 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.

(b) Formulating integrated step selection models

Integrated step selection function (iSSF) simultaneously incorporates movement and

habitat selection within a conditional logistic regression framework (Figure 1) [37–39]. As in

other resource and step selection analyses [40], 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 [37], where a step is the linear connection between consecutive

relocations, and turn angle is the angular deviation between the headings of two consecutive

steps [41]. We generated available steps and turn angles based on the distributions informed by

observed population-level movement behaviour using the *amt* package in R [42]. 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 [37]. 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 10 available

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

logistic regression framework (see section *g*). In addition to generating available movement

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

(c) Social network analysis

We used the R [43] packages *spatsoc* [44] and *igraph* [45] 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 *b*). 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 [29,31]. 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 [46]. 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 [47], 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 [48].

(d) Detecting social communities: long-term social preference

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

social communities [49]. 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 [49]. 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 [24]. Specifically, we generated null models based on GPS fixes to reduce

potential for type II error typically associated with node-based permutations [50]. Following

Spiegel et al. [24], 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 [24]. 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 [20].

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 [51]. 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.

(e) 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 iSSF models (see section *b*). 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 [44]. 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 *b*). Therefore,

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

available weekly dyadic SRI values (see section *b*).

In addition to incorporating social preference directly within iSSF models, 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

proximity-based social networks in each winter using the *asnipe* package in R [52]. We

calculated LARs for each winter by generating social networks following the procedure

described above (section *e*) using GPS relocations. LARs measure the probability that pairs of

individuals associating at a given relocation would still associate at subsequent relocations [53].

We generated annual LARs to compare temporal stability to assess potential for within-season

patterns of association among individuals. In addition, we also compared within-season 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 (see Figure S4).

(f) 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 [54]

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 [55].

(g) Modelling collective movement and habitat selection

We fit two iSSF models with a series of fixed and random effects using the *glmmTMB*

package in R following Muff et al. [56], through which 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. For robust standard errors, all variables included in the fixed

effect structure were also included in the random effect structure [56]. Our first model was the

nearest neighbour model and 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. 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 in *spatsoc* [44]. 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, nearest neighbour distance 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.Our

second model was the simple ratio index model, and included the same set of covariates, but

instead of nearest neighbour distance, we included weekly dyadic simple ratio index (for details

on calculating SRI, see section *c*) and interactions between simple ratio index and step length,

and simple ratio index and each of the proportion of lichen, forest, and open habitats within 200

m of the point location.

(h) Calculating effect sizes

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

habitat features influenced selection [57]. 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 1). 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 1). 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 S1), thus facilitating the potential for association between

social communities.

We found the simple ratio index model was more likely than the nearest neighbour model

was on AIC (ΔAIC = 70969; Δlog-likelihood = 35484; Table S2). 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 they shared a high SRI value with their nearest neighbour,

suggesting potential that conspecific familiarity influenced foraging-related movement (Table

S1). However, relative to its availability, caribou moved more quickly through lichen habitat,

perhaps due to the relatively high proportion of lichen habitat on the landscape (Figure 3).

Meanwhile, relative selection strength for all habitats decayed as nearest neighbours were further

away, however, relative selection for lichen and open habitats was stronger than forest (Figure

4).

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, as well as 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 familiarity, or not, 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.

320   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 [58]. 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 [59]. 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 [60]. 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

[24]. 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 [5]. 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 [27]. Our results therefore

suggest familiar individuals collectively move slowly through foraging habitat. Individuals are

more likely to trust social information about food sources and predation risks 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 foraging socially

with familiar individuals but spacing apart from less familiar individuals to reduce competition.

Moreover, individuals selected lichen habitat relative to its availability close to nearest

neighbours. Taken together, these results suggest that not only do individuals select foraging

habitat close to familiar individuals, but they also move between foraging sites with any

individual. 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 [63].

Moreover, in the bison system, the costs and benefits of foraging in a group are moderated by

collective decision making [64] and collective movement [65], 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 [66]. Assamese macaques (*Macaca*

assamensis) distance from one another during foraging, but move collectively between foraging

sites [67], while individual giraffes (*Giraffa camelopardalis*) show social preference for

conspecifics during foraging, but not during movement [22]. 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 [68,69], 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 foraging with familiar conspecifics is the result of information transfer

about the location or quality of cratering sites, although competition among individual caribou

for craters in the winter can be substantial [68]. Moreover, selection for open habitat relative to

its availability in groups could also reflect the use of social information about foraging sites [29]

or predation [70]. Craters can vary in size and distribution [30]; however, craters may only be

large enough for a single individual to forage at a time [71]. Foraging near familiar 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

[31,72], groups vary in size based on movement and habitat selection behaviour presumably to

balance the trade-off between competition and information acquisition. Unfamiliar individuals

spread out such that competition at foraging sites is limited, although group members with high

familiarity appear to remain in close proximity. 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 [68,73]. 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

[74], movement ecology [75], and habitat selection [76]. We also demonstrate how social

association is context-dependent, where individuals forage with familiar conspecifics, but moved

collectively with any conspecific 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 [9]. Moreover, individual variation in phenotypes attributable to movement or

habitat selection may affect how individuals experience the social environment [77]. Movement,

habitat selection, and social behaviour are clearly linked; as van Moorter et al. [1] 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.

Ethics. 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.

Data accessibility. 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.

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Table 1. Summary of winter community assignment for three caribou social networks (2017–

2019), including the number of individuals (N), number of communities, community size,

modularity (Q), and the community assortativity coefficient ($R_{com} = 0$ indicates no confidence in

community assignment; $R_{com} = 1$ indicates certainty in community assignment).

Year	N	Community	Community size	Home range area (km^2)	Q	R_{com}
2017	10	1	3	84.71	0.13	0.95
		2	7	43.43		
2018	16	1	5	113.09	0.17	1.00
		2	7	77.31		
		3	1	13.84		
		4	1	23.11		
		5	1	4.94		
		6	1	19.06		
2019	12	1	2	32.39	0.13	1.00
		2	9	33.87		
		3	1	1.26		

Figure captions

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.

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.

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.

References

1. Van Moorter B, Rolandsen CM, Basille M, Gaillard J-M. 2016 Movement is the glue connecting home ranges and habitat selection. *J. Anim. Ecol.* **85**, 21–31. (doi:10.1111/1365-2656.12394)

2. He P, Maldonado-Chaparro AA, Farine DR. 2019 The role of habitat configuration in shaping social structure: a gap in studies of animal social complexity. *Behav. Ecol. Sociobiol.* **73**, 9.

3. Spiegel O, Leu ST, Bull CM, Sih A. 2017 What's your move? Movement as a link between personality and spatial dynamics in animal populations. *Ecol. Lett.* **20**, 3–18. (doi:10.1111/ele.12708)

4. Strandburg-Peshkin A, Papageorgiou D, Crofoot MC, Farine DR. 2018 Inferring influence and leadership in moving animal groups. *Philos. Trans. R. Soc. B* **373**, 20170006. (doi:10.1098/not)

5. Chamaillé-Jammes S, Fritz H, Valeix M, Murindagomo F, Clobert J. 2008 Resource variability, aggregation and direct density dependence in an open context: The local regulation of an African elephant population. *J. Anim. Ecol.* **77**, 135–144. (doi:10.1111/j.1365-2656.2007.01307.x)

6. Spiegel O, Sih A, Leu ST, Bull CM. 2018 Where should we meet? Mapping social network interactions of sleepy lizards shows sex-dependent social network structure. *Anim. Behav.* **136**, 207–215. (doi:10.1016/j.anbehav.2017.11.001)

7. van Beest FM, Uzal A, Vander Wal E, Laforge MP, Contasti AL, Colville D, Mcloughlin PD. 2014 Increasing density leads to generalization in both coarse-grained habitat selection and fine-grained resource selection in a large mammal. *J. Anim. Ecol.* **83**, 147–156. (doi:10.1111/1365-2656.12115)

8. Jolles JW, King AJ, Killen SS. 2019 The role of individual heterogeneity in collective animal behaviour. *Trends Ecol. Evol.* **35**, 278–291. (doi:10.1016/j.tree.2019.11.001)

9. Strandburg-Peshkin A, Farine DR, Crofoot MC, Couzin ID. 2017 Habitat structure shapes individual decisions and emergent group structure in collectively moving wild baboons. *Elife* **6**, e19505. (doi:10.7554/eLife.19505)

10. Spiegel O, Leu ST, Bull CM, Sih A. 2017 What's your move? Movement as a link between personality and spatial dynamics in animal populations. *Ecol. Lett.* **20**, 3–18. (doi:10.1111/ele.12708)

11. Creel S, Schuette P, Christianson D. 2014 Effects of predation risk on group size, vigilance, and foraging behavior in an African ungulate community. *Behav. Ecol.* **25**, 773–784. (doi:10.1093/beheco/aru050)

12. Lima SL. 1995 Back to the basics of anti-predatory vigilance: the group-size effect. *Anim. Behav.* **49**, 11–20.

13. Harel R, Spiegel O, Getz WM, Nathan R. 2017 Social foraging and individual consistency in following behaviour: testing the information centre hypothesis in free-ranging vultures. *Proc. R. Soc. B* **284**, 20162654.

14. Lachlan RF, Crooks L, Laland KN. 1998 Who follows whom? Shoaling preferences and social learning of foraging information in guppies. *Anim. Behav.* **56**, 181–190. (doi:10.1006/anbe.1998.0760)

15. Giraldeau LA, Dubois F. 2008 Social foraging and the study of exploitative behavior. *Adv. Study Behav.* **38**, 59–104. (doi:10.1016/S0065-3454(08)00002-8)

515   16.   Patin R, Fortin D, Sueur C, Chamaillé-Jammes S. 2019 Space use and leadership modify
516   dilution effects on optimal vigilance under food-safety trade-offs. *Am. Nat.* **193**, E15–E28.
517   (doi:10.1086/700566)

518   17.   Spiegel O, Crofoot MC. 2016 The feedback between where we go and what we know—
519   information shapes movement, but movement also impacts information acquisition. *Curr.*
520   *Opin. Behav. Sci.* **12**, 90–96. (doi:10.1016/j.cobeha.2016.09.009)

521   18.   Jones TB, Patrick SC, Evans JC, Wells MR. 2020 Consistent sociality but flexible social
522   associations across temporal and spatial foraging contexts in a colonial breeder. *Ecol. Lett.*
523   **23**, 1085–1096. (doi:10.1111/ele.13507)

524   19.   Giraldeau L-A, Caraco T. 2018 *Social foraging theory*. Princeton University Press.

525   20.   Mourier J, Vercelloni J, Planes S. 2012 Evidence of social communities in a spatially
526   structured network of a free-ranging shark species. *Anim. Behav.* **83**, 389–401.
527   (doi:10.1016/j.anbehav.2011.11.008)

528   21.   Best EC, Seddon JM, Dwyer RG, Goldizen AW. 2013 Social preference influences female
529   community structure in a population of wild eastern grey kangaroos. *Anim. Behav.* **86**,
530   1031–1040. (doi:10.1016/j.anbehav.2013.09.008)

531   22.   Muller Z, Cantor M, Cuthill IC, Harris S. 2018 Giraffe social preferences are context
532   dependent. *Anim. Behav.* **146**, 37–49. (doi:10.1016/j.anbehav.2018.10.006)

533   23.   Silk JB. 2007 Social components of fitness in primate groups. *Science (80-.).* **317**, 1347–
534   1351.

535   24.   Spiegel O, Leu ST, Sih A, Bull CM. 2016 Socially interacting or indifferent neighbours?
536   Randomization of movement paths to tease apart social preference and spatial constraints.
537   *Methods Ecol. Evol.* **7**, 971–979. (doi:10.1111/2041-210X.12553)

538   25.   Pinter-Wollman N *et al.* 2013 The dynamics of animal social networks: Analytical,
539   conceptual, and theoretical advances. *Behav. Ecol.* **25**, 242–255.
540   (doi:10.1093/beheco/art047)

541   26.   Croft DP, Darden SK, Wey TW. 2016 Current directions in animal social networks. *Curr.*
542   *Opin. Behav. Sci.* **12**, 52–58. (doi:10.1016/j.cobeha.2016.09.001)

543   27.   Owen-Smith N, Fryxell JM, Merrill EH. 2010 Foraging theory upscaled: the behavioural
544   ecology of herbivore movement. *Philos. Trans. R. Soc. B* **365**, 2267–2278.
545   (doi:10.1098/rstb.2010.0095)

546   28.   Bergerud AT, Mercer WE. 1989 Caribou introductions in eastern North America. *Wildl.*
547   *Soc. Bull.* **17**, 111–120.

548   29.   Lesmerises F, Johnson CJ, St-Laurent M-H. 2018 Landscape knowledge is an important
549   driver of the fission dynamics of an alpine ungulate. *Anim. Behav.* **140**, 39–47.
550   (doi:10.1016/j.anbehav.2018.03.014)

551   30.   Bergerud AT. 1974 Relative abundance of food in winter for Newfoundland caribou.
552   *Oikos* **25**, 379–387.

553   31.   Peignier M, Webber QMR, Koen EL, Laforge MP, Robitaille AL, Vander Wal E. 2019
554   Space use and social association in a gregarious ungulate: Testing the conspecific
555   attraction and resource dispersion hypotheses. *Ecol. Evol.* **9**, 5133–5145.
556   (doi:10.1002/ece3.5071)

557   32.   Boertje RD. 1985 An Energy Model for Adult Female Caribou of the Denali Herd,
558   Alaska. *J. Range Manag.* **38**, 468–473. (doi:10.2307/3899725)

559   33.   Duquette LS, Klein DR. 1987 Activity budgets and group size of caribou during spring
560   migration. *Can. Jounal Zool.* **65**.

34.   Fortin D, Courtois R, Etcheverry P, Dussault C, Gingras A. 2008 Winter selection of
landscapes by woodland caribou: Behavioural response to geographical gradients in
habitat attributes. *J. Appl. Ecol.* **45**, 1392–1400. (doi:10.1111/j.1365-2664.2008.01542.x)
35.   Integrated-Informatics. 2014 Sustainable Development and Strategic Science Branch Land
Cover Classification. **5**, 3–19.
36.   Bjørneraas K, Van Moorter B, Rolandsen CM, Herfindal I. 2010 Screening Global
Positioning System Location Data for Errors Using Animal Movement Characteristics. *J.*
Wildl. Manage. **74**, 1361–1366. (doi:10.2193/2009-405)
37.   Avgar T, Potts JR, Lewis MA, Boyce MS. 2016 Integrated step selection analysis:
Bridging the gap between resource selection and animal movement. *Methods Ecol. Evol.*
7, 619–630. (doi:10.1111/2041-210X.12528)
38.   Duchesne T, Fortin D, Rivest LP. 2015 Equivalence between step selection functions and
biased correlated random walks for statistical inference on animal movement. *PLoS One*
10, 1–12. (doi:10.1371/journal.pone.0122947)
39.   Basille M, Fortin D, Dussault C, Bastille-Rousseau G, Ouellet J-P, Courtois R. 2015
Plastic response of fearful prey to the spatio-temporal dynamics of predator distribution.
Ecology **96**, 150511123704000. (doi:10.1890/14-1706.1)
40.   Fortin D, Beyer HL, Boyce MS, Smith DW, Duchesne T, Mao JS. 2005 Wolves influence
elk movements: Behavior shapes a trophic cascade in Yellowstone National Park. *Ecology*
86, 1320–1330.
41.   Prokopenko CM, Boyce MS, Avgar T. 2017 Characterizing wildlife behavioural
responses to roads using integrated step selection analysis. *J. Appl. Ecol.* **54**, 470–479.
(doi:10.1111/1365-2664.12768)
42.   Signer J, Fieberg J, Avgar T. 2019 Animal movement tools (amt): R package for
managing tracking data and conducting habitat selection analyses. *Ecol. Evol.* **9**, 880–890.
(doi:10.1002/ece3.4823)
43.   R Core Team. 2019 R: A language and environment for statistical computing. ,
https://www.r-project.org/.
44.   Robitaille AL, Webber QMR, Vander Wal E. 2019 Conducting social network analysis
with animal telemetry data: applications and methods using spatsoc. *Methods Ecol. Evol.*
10, 1203–1211. (doi:10.1111/2041-210X.13215)
45.   Csárdi G, Nepusz T. 2006 The igraph software package for complex network research.
InterJournal Complex Syst. **1695**, 1–9.
46.   Kasozi H, Montgomery RA. 2020 Variability in the estimation of ungulate group sizes
complicates ecological inference. *Ecol. Evol.* , 1–9. (doi:10.1002/ece3.6463)
47.   Cairns SJ, Schwager SJ. 1987 A comparison of association indices. *Anim. Behav.* **35**,
1454–1469.
48.   Farine DR, Whitehead H. 2015 Constructing, conducting and interpreting animal social
network analysis. *J. Anim. Ecol.* **84**, 1144–1163. (doi:10.1111/1365-2656.12418)
49.   Newman MEJ. 2006 Modularity and community structure in networks. *Proc. Natl. Acad.*
Sci. **103**, 8577–8582.
50.   Farine DR. 2014 Measuring phenotypic assortment in animal social networks: Weighted
associations are more robust than binary edges. *Anim. Behav.* **89**, 141–153.
(doi:10.1016/j.anbehav.2014.01.001)
51.   Shizuka D, Farine DR. 2016 Measuring the robustness of network community structure
using assortativity. *Anim. Behav.* **112**, 237–246.

52.   Farine DR. 2013 Animal social network inference and permutations for ecologists in R
608   using asnipe. *Methods Ecol. Evol.* **4**, 1187–1194. (doi:10.1111/2041-210X.12121)
609   53.   Whitehead H. 2008 *Analyzing animal societies: quantitative methods for vertebrate social*
610   *analysis.* University of Chicago Press.
611   54.   Worton BJ. 1989 Kernel methods for estimating the utilization distribution in home-range
612   studies. *Ecology* **70**, 164–168.
613   55.   Fieberg J, Kochanny CO. 2005 Quanitfying home-range overlap: the importance of the
614   utilization distribution. *J. Wildl. Manage.* **69**, 1346–1359. (doi:10.2193/0022-
615   541X(2005)69)
616   56.   Muff S, Signer J, Fieberg J. 2020 Accounting for individual-specific variation in habitat-
617   selection studies: Efficient estimation of mixed-effects models using Bayesian or
618   frequentist computation. *J. Anim. Ecol.* **89**, 80–92. (doi:10.1111/1365-2656.13087)
619   57.   Avgar T, Lele SR, Keim JL, Boyce MS. 2017 Relative Selection Strength: Quantifying
620   effect size in selection inference. *Ecol. Evol.* **7**, 5322–5330. (doi:10.1002/ece3.3122)
621   58.   Sueur C *et al.* 2011 Collective decision-making and fission-fusion dynamics: A
622   conceptual framework. *Oikos* **120**, 1608–1617. (doi:10.1111/j.1600-0706.2011.19685.x)
623   59.   Shizuka D, Chaine AS, Anderson J, Johnson O, Laursen IM, Lyon BE. 2014 Across-year
624   social stability shapes network structure in wintering migrant sparrows. *Ecol. Lett.* **17**,
625   998–1007. (doi:10.1111/ele.12304)
626   60.   Cantor M, Wedekin LL, Guimarães PR, Daura-Jorge FG, Rossi-Santos MR, Simões-
627   Lopes PC. 2012 Disentangling social networks from spatiotemporal dynamics: The
628   temporal structure of a dolphin society. *Anim. Behav.* **84**, 641–651.
629   (doi:10.1016/j.anbehav.2012.06.019)
630   61.   Jacobs A. 2010 Group cohesiveness during collective movements: Travelling apart
631   together. *Behav. Processes* **84**, 678–680. (doi:10.1016/j.beproc.2010.03.004)
632   62.   Krause J, Ruxton GD. 2002 *Living in groups*. Oxford University Press.
633   63.   Merkle JA, Sigaud M, Fortin D. 2015 To follow or not? How animals in fusion-fission
634   societies handle conflicting information during group decision-making. *Ecol. Lett.* **18**,
635   799–806. (doi:10.1111/ele.12457)
636   64.   Sigaud M, Merkle JA, Cherry SG, Fryxell JM, Berdahl A, Fortin D. 2017 Collective
637   decision-making promotes fitness loss in a fusion-fission society. *Ecol. Lett.* **20**, 33–40.
638   (doi:10.1111/ele.12698)
639   65.   Courant S, Fortin D. 2012 Time allocation of bison in meadow patches driven by potential
640   energy gains and group size dynamics. *Oikos* **121**, 1163–1173. (doi:10.1111/j.1600-
641   0706.2011.19994.x)
642   66.   Morrell LJ, Ruxton GD, James R. 2011 Spatial positioning in the selfish herd. *Behav.*
643   *Ecol.* **22**, 16–22. (doi:10.1093/beheco/arq157)
644   67.   Heesen M, Macdonald S, Ostner J, Schülke O. 2015 Ecological and Social determinants
645   of group cohesiveness and within-group spatial position in wild assamese macaques.
646   *Ethology* **121**, 270–283. (doi:10.1111/eth.12336)
647   68.   Barrette C, Vandal D. 1986 Social rank, dominance, antler size, and access to food in
648   snow-bound wild woodland caribou. *Behaviour* **97**, 118–145.
649   69.   Hirotani A. 1990 Social organization of reindeer (*Rangifer tarandus*), with a special
650   reference to relationships among females. *Can. J. Zool.* **68**, 743–749.
651   70.   Hamilton WD. 1971 Geometry for the selfish herd. *J. Theor. Biol.* **31**, 295–311.
652   (doi:10.1016/0022-5193(71)90189-5)

653   71.   Mayor SJ, Schaefer JA, Schneider DC, Mahoney SP. 2009 The spatial structure of habitat
654   selection: A caribou's-eye-view. *Acta Oecologica* **35**, 253–260.
655   (doi:10.1016/j.actao.2008.11.004)

656   72.   Jung TS, Larter NC, Allaire DG. 2019 Social organization of boreal woodland caribou
657   (Rangifer tarandus caribou) in response to decreasing annual snow depth. *Mammal Res.*
658   **64**, 377–385.

659   73.   Schaefer JA, Mahoney SP. 2001 Antlers on female caribou: Biogeography of the bones of
660   contention. *Ecology* **82**, 3556–3560.

661   74.   Croft DP, Madden JR, Franks DW, James R. 2011 Hypothesis testing in animal social
662   networks. *Trends Ecol. Evol.* **26**, 502–507. (doi:10.1016/j.tree.2011.05.012)

663   75.   Nathan R, Getz WM, Revilla E, Holyoak M, Kadmon R, Saltz D, Smouse PE. 2008 A
664   movement ecology paradigm for unifying organismal movement research. *Proc. Natl.*
665   *Acad. Sci.* **105**, 19052–19059. (doi:10.1073/pnas.0800375105)

666   76.   McLoughlin PD, Morris DW, Fortin D, Vander Wal E, Contasti AL. 2010 Considering
667   ecological dynamics in resource selection functions. *J. Anim. Ecol.* **79**, 4–12.
668   (doi:10.1111/j.1365-2656.2009.01613.x)

669   77.   Webber QMR, Vander Wal E. 2018 An evolutionary framework outlining the integration
670   of individual social and spatial ecology. *J. Anim. Ecol.* **87**, 113–127. (doi:10.1111/1365-
671   2656.12773)