Sea ice loss increases genetic isolation in a high Arctic ungulate metapopulation

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Abstract
Sea ice loss may have dramatic consequences for population connectivity, extinction–colonization dynamics, and even the persistence of Arctic species subject to climate change. This is of particular concern in face of additional anthropogenic stressors, such as overexploitation. In this study, we assess the population-genetic implications of diminishing sea ice cover in the endemic, high Arctic Svalbard reindeer (Rangifer tarandus platyrhynchus) by analyzing the interactive effects of landscape barriers and reintroductions (following harvest-induced extirpations) on their metapopulation genetic structure. We genotyped 411 wild reindeer from 25 sampling sites throughout the entire subspecies’ range at 19 microsatellite loci. Bayesian clustering analysis showed a genetic structure composed of eight populations, of which two were admixed. Overall population genetic differentiation was high (mean $F_{ST} = 0.21$). Genetic diversity was low (allelic richness [AR] = 2.07–2.58; observed heterozygosity = 0.23–0.43) and declined toward the outer distribution range, where populations showed significant levels of inbreeding. Coalescent estimates of effective population sizes and migration rates revealed strong evolutionary source–sink dynamics with the central population as the main source. The population genetic structure was best explained by a landscape genetics model combining strong isolation by glaciers and open water, and high connectivity by dispersal across winter sea ice. However, the observed patterns of natural isolation were strongly modified by the signature of past harvest-induced extirpations, subsequent reintroductions, and recent lack of sea ice. These results suggest that past and current anthropogenic drivers of metapopulation dynamics may have interactive effects on large-scale ecological and evolutionary processes. Continued loss of sea ice as a dispersal corridor within and between island systems is expected to increase the genetic isolation of populations, and thus threaten the evolutionary potential and persistence of Arctic wildlife.

Keywords
Arctic, circuit theory, climate change, extinction risk, harvesting, isolation, landscape genetics, least-cost path, population genetics, sea ice
1 | INTRODUCTION

Many current threats to biodiversity are linked to anthropogenic stressors, which have either direct (e.g., through exploitation) or indirect (e.g., through habitat fragmentation, climate change) effects on populations (Lande, 1998). Direct effects of overharvesting on population demographic can have long-term evolutionary consequences through loss of genetic diversity, changes in population subdivision, or selective genetic changes (Allendorf, England, Luikart, Ritchie, & Ryman, 2008; Pinsky & Palumbi, 2014). Such genetic effects can reduce the evolutionary potential, delay population recovery and increase extinction risk (Frankham, 2005). In the Arctic, mammals such as the polar bear (*Ursus maritimus*), bowhead whale (*Balaena mysticetus*), walrus (*Odobenus rosmarus*), Arctic fox (*Vulpes lagopus*), and caribou and reindeer (*Rangifer tarandus*) have been extensively harvested in the past (Fay, Kelly, & Sease, 1989; Higdon, 2010; Kruse, 2017), but genetic impact studies have so far been rare (Alter et al., 2012; Petersen, Manseau, & Wilson, 2010; Taylor, Jenkins, & Arcese, 2012). Furthermore, global warming strongly influences the Arctic ecosystem, and is expected to increasingly affect future ecological dynamics (Descamps et al., 2017; Post et al., 2009, 2013).

A dominant feature of Arctic climate change is the rapid loss of sea ice, with far-reaching ecological implications (Post et al., 2013; Stroeve, Holland, Meier, Scambos, & Serreze, 2007). Sea ice provides an important corridor for dispersal across water bodies in Arctic fox (Geffen et al., 2007; Norén et al., 2011), polar bear (Laidre et al., 2018), caribou (Jenkins et al., 2016; Leblond, St-Laurent, & Côté, 2016; Miller, Barry, & Calvert, 2005), and even vascular plants (Alsos et al., 2016). Because sea ice loss is associated with increased global temperatures and Arctic amplification (Serreze, Barrett, Stroeve, Kindig, & Holland, 2009), a future increase in the isolation of Arctic island populations is expected. In general, island populations have lower genetic diversity and higher levels of inbreeding than mainland populations (Frankham, 1997, 1998). This makes them more prone to loss of genetic diversity through genetic drift and, ultimately, increased extinction risk following environmental perturbations (Frankham, 1998; Lande, 1998; Saccheri et al., 1998).

Dispersal and recolonization facilitated by sea ice are therefore crucial for the persistence and resilience of many Arctic island populations (Hanski, 1998; Post et al., 2013; Vuilleumier, Bolker, & Levine, 2010).

**Svalbard reindeer (R. t. platyrhynchus)** live in a pristine archipelagic environment with very little human infrastructure and a simple food-web system where predation and interspecific competition are virtually absent (Descamps et al., 2017). Many reindeer populations were extirpated (i.e., locally extinct) due to overharvesting between the 17th and early 20th century (Kruse, 2017; Lene, 1959). Today, Svalbard reindeer have largely recovered their historical range through natural recolonization and human-mediated reintroductions (Le Moulec, Pedersen, Stien, Rosvold, & Hansen, 2019). Their distribution is fragmented into semi-isolated populations by natural landscape barriers such as glaciers, steep mountains, fjords, and open water. Sea ice and fjord ice may provide an important dispersal corridor during winter (Hansen, Aanes, & Saether, 2010; Jenkins et al., 2016; Leblond et al., 2016). However, warm water brought by the North Atlantic Current often restricts sea ice formation in southern and western parts of Svalbard (Nilsen, Skogseth, Vaadal-Lunde, & Inall, 2016), and a strong negative trend in sea ice cover over the past few decades has been reported for northern parts (Onarheim, Smedsrud, Ingvaldsen, & Nilsen, 2014). Overall, spatiotemporal variation in sea ice connectivity is likely an important driver of metapopulation dynamics and genetics in Svalbard reindeer (cf. Peary caribou R. t. pearyi; Jenkins et al., 2016; Mallory & Boyce, 2019), along with dispersal barriers on land and extinction–colonization dynamics linked to past overharvesting and recent reintroductions.

To better understand the ecological and evolutionary consequences of past and current anthropogenic impacts, and with particular emphasis on the potential consequences of sea ice loss, we adopted a landscape genetics approach using neutral markers in Svalbard reindeer. Landscape genetics have been extensively used to infer knowledge about the effects of landscape barriers and habitat fragmentation (Storfer, Murphy, Spear, Holderegger, & Waits, 2010), including in caribou and reindeer (Jenkins et al., 2016; Yannic et al., 2017). By sampling individuals across the entire distribution range of Svalbard reindeer, we covered a gradient of sea ice cover frequencies and landscape fragmentation through glacial and open water barriers. We were able to assess causes and consequences of extinction–colonization dynamics by including populations that persisted past overexploitation and populations in areas that were recolonized—either through natural expansion or reintroduction by humans—after harvest-induced extirpation. In particular, we expected that sea ice would represent an important dispersal corridor explaining significant parts of the reindeer’s genetic differentiation and structure, potentially modified by harvest-induced extirpations and subsequent reintroduction programs.

2 | MATERIALS AND METHODS

2.1 | Study area and species

The Svalbard archipelago is located in the Arctic Ocean north of Norway (76°–81°N, 10°–35°E; Figure 1a). The largest island is Spitsbergen, followed by Nordaustlandet, Edgeøya, and Barentsøya. The landscape has a highly heterogeneous topography with coastal plains, mountain plateaus, steep slopes, wide valleys, and fjords. About 60% of the land surface is glaciated and only 15% is covered with vegetation, mainly characterized by snowbed communities, *Dryas tundra*, and sparse pioneer vegetation (Johansen, Karlsen, & Tømmervik, 2012). The northward inflow of warm Atlantic water leads to a milder climate and less sea ice along the west coast of Spitsbergen compared to the eastern parts of Svalbard, which are characterized by cold Arctic water. Sea ice concentration usually reaches its maximum in March–April, but has been gradually
The Svalbard archipelago (a) is situated in the Arctic Ocean north of Norway (inset). Svalbard reindeer were nearly extinct due to overharvesting, but four extant populations remained (pink areas indicate known distribution in the 1950s; Lønø, 1959). Reindeer were re-introduced from the population near Longyearbyen (blue circle) to west Spitsbergen near Ny-Ålesund in 1978 (blue square) and Daudmannsøya in 1984–1985 (blue triangle). Today, reindeer are distributed across Svalbard, except in glaciated areas (white). Coastal currents are characterized by southward cold Arctic water in the east (blue arrows) and northward warm Atlantic water in the west (red arrows). (b) Time series of April–May sea ice cover in five coastal areas of Spitsbergen (data retrieved from Prop et al., 2015).

2.2 | Sampling

Samples from 456 reindeer were collected at 25 sites across the entire Svalbard archipelago (Table S1). Dispersal between nearby sampling sites was assumed to be limited by mountains, glaciers, fjords, and/or open sea and was confirmed by genetic differentiation analyses (see below). All samples were obtained between 2014 and 2016, except for nine blood samples from Nordaustlandet that were collected in 1995 and included to increase sample size in this remote location (Kvie et al., 2016).

The collected material included soft tissue \( (n = 104) \), bone \( (n = 66) \), antler \( (n = 69) \), blood \( (n = 9) \); Kvie et al., 2016), or tooth \( (n = 5) \) samples from carcasses, and fresh feces \( (n = 48) \) from live reindeer. Additionally, we obtained hair \( (n = 85) \) or soft tissue \( (n = 70) \) obtained from ear punches during ear-tagging from live reindeer included in mark-recapture programs (Albon et al., 2017). Mark-recapture studies followed ethical requirements approved by the Norwegian Food Safety Authority. Hair, bone, antler, and tooth samples were dry-stored, fecal samples were directly stored in 96% ethanol or frozen, while soft tissue and blood samples were frozen after collection. Coordinates were recorded for all individuals at the time of sampling, except for those from Wijdefjorden and the nine individuals from Nordaustlandet sampled in 1995. For those, approximate coordinates were obtained based on sampling location descriptions.

After excluding samples with a low DNA amplification success (see below), the total dataset comprised 411 genotyped individuals covering the full geographic range of Svalbard reindeer, with sample sizes ranging from 20 to 52 individuals in eight sites, 6 to 19 individuals in eleven sites, and 1 to 4 individuals in six sites (Table S1).
2.3 | DNA extraction and microsatellite genotyping

DNA was extracted using the Chelex 100 method for hair samples (Walsh, Metzger, & Higuchi, 1991) and the spin column method (QIAGEN) for all other sample types (see Methods S1 for detailed procedures). Individuals were genotyped at 19 polymorphic microsatellite loci (Table S2) using standard procedures for PCR, electrophoresis, and quality control. For 45 samples (i.e., 6% of soft tissue samples, 1% of hair samples, 19% of feces samples, 29% of bone samples, 7% of antler samples, and 20% of tooth samples), less than 10 loci could be amplified, probably due to PCR inhibitors or low reindeer-specific DNA yield. These samples were therefore excluded. Overall, of the 411 genotyped individuals, 81% were genotyped at 19 microsatellite loci, 8% at 18 loci, 4% at 17 loci, 2% at 16 loci, and the remaining 5% at 10–15 loci. The final dataset of 411 multilocus genotypes reached a data completeness of more than 96%.

2.4 | Population genetic structure

We were interested in the population genetic structure of Svalbard reindeer at two hierarchical levels. First, we quantified genetic differentiation by calculating the global and pairwise $F_{ST}$ (Weir & Cockerham, 1984) between all sampling sites with >5 sampled individuals ($n = 18$ sites; Table S3; note that one site with eight samples was excluded due to mixed cluster results; see below). $F_{ST}$ values were tested for significance based on 10,000 permutations among samples using the R-package `strataG` (Archer, Adams, & Schneiders, 2017) in R version 3.5.1 (R Core Team, 2016). Second, to reveal large-scale patterns of genetic diversity, isolation, and connectivity, we identified populations based on a higher-order (hierarchical) genetic structure. For this, we used the individual-based Bayesian clustering analyses as implemented in `STRUCTURE` 2.3 (Pritchard, Stephens, & Donnelly, 2000) and `TESS` 2.3 (Chen, Durand, Forbes, & Francois, 2007; see Methods S2 for technical details). `STRUCTURE` performs well for inferring the number of clusters without prior knowledge of sampling locations (Latch, Dharmarajan, Glaubitz, & Rhodes, 2006). However, the joint analysis of genetic and geographic structure, as implemented in `TESS`, is expected to yield more realistic cluster assignments for individuals from weakly differentiated but distant populations (Chen et al., 2007). We then assigned individuals to their most likely cluster based on a proportional assignment threshold of ≥0.75 (see Yannic et al., 2016, for a similar approach) and, for each sampling site, calculated the proportion of individuals assigned to each cluster (Table S4). Populations could thus be delineated based on both genetic structure and geographic segregation of sampling sites.

As a conservative measure, sites with ≤5 samples as well as sites with ≤10 samples showing strong genetic admixture were hereafter only included in individual-based analyses. Genetic differentiation between populations was estimated using pairwise $F_{ST}$ values as for sampling sites. To account for consequences of recent reintroductions on the genetic structure, we conducted a post hoc analysis in `STRUCTURE` excluding individuals from reintroduced sites (Figure S2). Finally, the geographic range of genetically distinct populations is of practical interest for the future management of Svalbard reindeer. To determine these ranges, we first plotted individual coordinates on the map and interpolated the individual clustering assignments (as identified with `TESS`) using thin plate spline regressions in the R-package `fields` (Nychka, Furrer, Paige, & Sain, 2017). The geographic range of each population was then delineated using minimum convex polygons in the R-package `adehabitatHR` (Calenge, 2006).

2.5 | Genetic diversity analyses

Analyses of genetic diversity were performed at the sampling site and population level, but only reported for populations in the main text (see Table S1 for genetic diversity estimates of sampling sites). Genetic diversity was assessed by measures of AR and heterozygosity. AR and private AR (i.e., alleles unique to the population; AR_{P}) were calculated using a rarefaction approach on a minimum sample size of 34 genes (i.e., 17 diploid individuals) implemented in `HP-RARE` 1.1 (Kalinowski, 2005). Observed and unbiased expected heterozygosity ($H_{O}$ and $\mu H_{E}$) were calculated and averaged over loci using the R-package `hierfstat` (Goudet, 2005). Differences in observed and expected heterozygosity between populations at the locus level were tested using two-tailed, paired sample t tests after arcsine square root transformation (e.g., Côté et al., 2002). Similarly, a non-parametric Mann–Whitney U test was used to test for differences in AR. Inbreeding coefficients ($F_{is}$) were calculated in `hierfstat` and bootstrapped using 10,000 iterations to obtain 95% confidence intervals.

Individual heterozygosity was estimated as the proportion of heterozygous loci for all 411 genotyped individuals (Coulon, 2010), which was then spatially interpolated using a thin plate spline regression in the R-package `fields` (Nychka et al., 2017). We tested for an expected decrease in individual heterozygosity toward the range margin (Eckert, Samis, & Lougheed, 2008) using a generalized linear regression model (binomial family with logit link function) with the Euclidian distance between individual coordinates and the reindeer density-weighted range center (Le Moulec et al., 2019) as a predictor. We also used a generalized mixed regression model to correct for non-independence among individuals from the same sampling site ($n = 25$).

2.6 | Effective population sizes and gene flow

To investigate metapopulation dynamics over an evolutionary timescale, we estimated long-term mutation-scaled effective population sizes ($\theta$) and directional gene flow between populations using a coalescent approach in `Migrate-n` 3.6.11 (Beerli, 2006; Beerli & Felsenstein, 2001; Methods S3; Table S5). Here, individuals from reintroduced sites and admixed populations were excluded to avoid spurious estimates of coalescent effective population size and gene flow.
Contemporary effective population sizes (Ne) were estimated using the linkage disequilibrium method in NeEstimator 2.0 (Do et al., 2014). As bias and precision of this method depend on the criterion of screening out rare alleles (Waples & Do, 2010), we estimated Ne using alleles with frequencies > 0.02 and > 0.01. To estimate ratios of Ne/Nc, we obtained estimates of census population size (Nc) and density (Nd) through distance sampling monitoring surveys conducted across Svalbard from 2013 to 2016 (Le Moullec et al., 2019). In further details, reindeer abundance was predicted at the pixel level (250 × 250 m) from a density surface model (Miller, Burt, Rexstad, & Thomas, 2013) where vegetation productivity index (maximum Normalized Difference Vegetation Index) was the main driver. Comparison of Ne estimates and Nc/Nd ratios clearly indicated that the point estimate of Ne in one population was inflated when using allele frequencies > 0.02 (Table S6). We therefore only report Ne based on allele frequencies > 0.01 in the main text. Directional estimates of recent gene flow (i.e., over the last few generations) were estimated using BayesAss 3.0.4 (Wilson & Rannala, 2003) for all populations (Methods S3; Table S7).

2.7 | Landscape genetics analysis

To better understand how gene flow in this fragmented metapopulation was affected by landscape barriers (glaciers, open sea, and mountains) and a seasonally dynamic corridor (winter sea ice), we investigated patterns of isolation-by-distance (IBD) and isolation-by-resistance (IBR). Landscape genetics were primarily analyzed at the sampling site level due to the discontinuous distribution and sedentary behavior of Svalbard reindeer (Côté et al., 2002) and the strong genetic differentiation observed among sampling sites (Table S3). Genetic distances among sites were estimated as FST/ (1 − FST; Rousset, 1997). Landscape distances were obtained by averaging distances among individuals between pairs of sampling sites, which were calculated as Euclidian distances (IBR; see below). Genetic distances among individuals were estimated as Rousset's δ (Rousset, 2000) using the program SPAGeDi 1.5 (Hardy & Vekemans, 2002) to analyze landscape genetics at the individual level.

Ecological distances (IBR) were estimated using both least-cost path and circuit theoretic models on resistance surfaces (McRae, 2006; McRae & Beier, 2007) in the R-package gdistance (Van Etten, 2017; see Methods S4 for technical details). Least-cost paths estimate the optimal route between positions, assuming perfect knowledge of the landscape, whereas circuit theory simultaneously considers all possible pathways connecting pairs of individuals/sites (McRae, 2006; McRae & Beier, 2007). Conductance values (0 < c < 1) were used to assign low to high probabilities of dispersal across different types of landscape surface (slope and elevation, glaciers, open water, and sea ice). We first investigated the effects of glaciers and open water on genetic distances (IBR scenario 1), while accounting for the effects of slope and elevation on movement cost. We then expanded the landscape model of IBR scenario 1 to include the effect of sea ice (IBR scenario 2) by increasing the conductance of sea surfaces with various thresholds of sea ice frequency (minimum of 50%–99%) during March 1986–2015. A threshold of approximately 100% sea ice frequency corresponded to a landscape model including connectivity across ice only in marine areas where ice was present during March in all years between 1986 and 2015. Parameter selection of conductance values and the optimal threshold of sea ice frequency were informed by Mantel correlations (Mantel, 1967) between genetic and ecological distances among natural sites (i.e., excluding sampling sites of reintroduced origin; Figures S3 and S4).

To understand the effects of reintroductions on landscape genetics, we separately analyzed IBD and IBR for genetic distances among natural sites (n = 13 sites) and genetic distances between reintroduced sites (n = 5 sites) and natural sites of either source (n = 4 sites) or non-source (n = 9 sites) origin of reintroductions (Table S1). However, due to the high computational workload to estimate circuit theory distances (see Methods S4), we only investigated the relationship between genetic distance and circuit theory distance at the natural site level. Landscape genetics at the individual level were analyzed for IBD and least-cost path IBR using the same conductance values and threshold of winter sea ice frequency as obtained from landscape genetics at the sampling site level.

Mantel tests between genetic and landscape distances were performed using the R-package vegan (Oksanen et al., 2018), but IBD versus IBR was tested using maximum-likelihood population effects models (MLPE) for final inference (Clarke, Rothery, & Raybould, 2002). The mixed effects parameterization in MLPE accounted for non-independence among the pairwise data. We implemented MLPE using the R-package ResistanceGA (Peterman, 2018), but only for analyses among natural sites as MLPE could not handle the incomplete distance matrices of pairwise combinations between natural and reintroduced sites. Marginal and conditional R² (i.e., variance explained by fixed effects only and both fixed and random effects, respectively) were estimated using the R-package piecewiseSEM (Lefcheck, 2016). Model selection was informed by Akaike’s information criterion corrected for sample size (AICc).

3 | RESULTS

3.1 | Population genetic structure

STRUCTURE revealed a strong genetic structure characterized by three major clusters and a further subdivision into six clusters (Figure 2a; Figure S1). The analysis in TESS confirmed the maximal number of six clusters (Figure 2; Figure S1). A clear composition of sampling sites from the same region was evident for each genetic cluster, with minimum 60%, but typically > 80% of individuals in each site assigned to one cluster (Table S4). Furthermore, the strong genetic structure of both the individual-based analyses (Figure 2) and pairwise genetic differentiation among sampling sites (Table S3) strongly reflected the geographic segregation of
sampling sites. However, two sampling sites with ≥10 sampled individuals, Wijdefjorden and North Isfjorden, showed a high degree of admixture (<60% of individuals assigned to one cluster).

![Diagram](image)

**FIGURE 2** Bayesian clustering analyses revealed three major clusters (K = 3) and a further subdivision into six clusters (K = 6). (a) Individual cluster assignments (q [0, 1]) for 411 Svalbard reindeer in STRUCTURE (K = 3 and 6) and TESS (K-max = 6), ordered in a south- and eastward direction by population. (b) Genetic structure at 25 sampling sites based on K = 6 in TESS (pie charts increasing in size relative to number of sampled individuals n). Cluster ranges were estimated using minimum convex polygons after spatial interpolation of individual clustering assignments.

Individuals and sampling sites were accordingly delineated into eight populations, that is, six separate populations for each of the six genetic clusters (North Spitsbergen, West Spitsbergen, Central Spitsbergen, South Spitsbergen, East Svalbard, and Nordaustlandet) and two populations showing mixed genetic cluster origins (Wijdefjorden and North Isfjorden). Note that a large proportion of individuals (0.45–0.50 per sampling site; Table S4) from Central Spitsbergen were admixed between Central Spitsbergen and West Spitsbergen cluster. This suggests a founder bottleneck and genetic drift after individuals were translocated from Central Spitsbergen to West Spitsbergen (Aanes et al., 2000). The post hoc analysis excluding the reintroduced sites supported the delineation of Central Spitsbergen as a distinct (rather than an admixed) population (Figure S2). The recolonized Wijdefjorden population indicated strong admixture with genetic signals from North and West Spitsbergen, while the post hoc analysis also supported the delineation of Wijdefjorden as a separate population (Figure S2). Finally, the mixed genetic signature in North Isfjorden and complementary analysis of recent gene flow (Table S7) suggested a recolonization from Central Spitsbergen and the reintroduced site at Daudmannsøya (Figure 1a).

Global F$_{ST}$ was 0.21 ± 0.09, while pairwise F$_{ST}$ between populations ranged from 0.05 (North Isfjorden–Central Spitsbergen) to 0.37 (North Spitsbergen–Nordaustlandet; Table 1). North Spitsbergen was the most strongly differentiated population (F$_{ST}$ > 0.31, except when paired with Wijdefjorden). The population of reintroduced origin, West Spitsbergen, and its source, Central Spitsbergen, were significantly differentiated from each other. The admixed North Isfjorden population was also significantly differentiated from West and Central Spitsbergen. Finally, pairwise F$_{ST}$ values between sites supported the expectation that gene flow was limited due to landscape barriers (Table S3).

### 3.2 Genetic diversity

We found 73 alleles at the 19 microsatellite loci, with a range of two to seven alleles per locus (Table S2). AR of populations

| Population             | North Spitsbergen | Wijdefjorden | West Spitsbergen | North Isfjorden | Central Spitsbergen | South Spitsbergen | East Svalbard | Nordaustlandet |
|------------------------|-------------------|--------------|------------------|----------------|---------------------|-------------------|--------------|----------------|
| North Spitsbergen      |                   | 0.158        |                  |                |                     |                   |              |                |
| Wijdefjorden           |                   |              | 0.342            |                |                     |                   |              |                |
| West Spitsbergen       |                   |              |                  | 0.370          | 0.194               | 0.084             |              |                |
| North Isfjorden        |                   |              |                  | 0.317          | 0.163               | 0.064             | 0.051        |                |
| Central Spitsbergen    |                   |              |                  | 0.361          | 0.164               | 0.123             | 0.222        | 0.134          |
| South Spitsbergen      |                   |              |                  | 0.361          | 0.164               | 0.123             | 0.222        | 0.134          |
| East Svalbard          |                   |              |                  | 0.370          | 0.277               | 0.197             | 0.188        | 0.158          | 0.262          |
| Nordaustlandet         |                   |              |                  | 0.374          | 0.207               | 0.166             | 0.215        | 0.160          | 0.198          | 0.219          |

Note: All F$_{ST}$ values were significantly different from zero after Bonferroni correction based on 10,000 permutations among samples.
was on average 2.36 ± 0.19 and ranged from 2.07 in North Spitsbergen to 2.58 in Wijdefjorden (Table 2). Private AR (when excluding the two admixed populations) was highest in North Spitsbergen, with approximately 10% of the estimated AR being unique to this population (calculated as the ratio of AR_{P, sub} to AR; Table 2). Similarly, 6%–8% of AR was unique to South Spitsbergen, East Svalbard, and Nordaustlandet. Observed heterozygosity within populations was on average 0.31 ± 0.07 and ranged from 0.23 to 0.43 in Nordaustlandet and Central Spitsbergen, respectively (Table 2). Differences in AR and expected heterozygosity between loci were not statistically significant after Bonferroni correction for multiple comparisons. However, observed heterozygosity was significantly higher in Central Spitsbergen than East Svalbard (paired t_{18} = 3.74; p < .01) and Nordaustlandet (paired t_{18} = 4.30; p < .001), and also significantly higher in Central Spitsbergen than North Spitsbergen (paired t_{18} = 3.64; p < .01) after Bonferroni correction when the two admixed populations were excluded from the multiple comparison. As such, conforming with the central-marginal hypothesis (Eckert et al., 2008), heterozygosity at the individual level was highest in Central Spitsbergen and decreased with increasing distance from the reindeer density-weighted range center (binomial regression, estimates with standard errors (±SE) on logit scale: intercept = −0.16 ± 0.05, slope = −0.52 ± 0.05 per 100 km from the range center; Figure 3). A mixed model correcting for non-independence among individuals from the same sampling site yielded qualitatively the same results (intercept = −0.26 ± 0.10, slope = −0.45 ± 0.08, random variance of intercept = 0.02 for n = 25 groups). Finally, significant levels of inbreeding were found in all peripheral populations, but not in Central Spitsbergen or the admixed populations Wijdefjorden and North Isfjorden (Table 2).

### TABLE 2 Genetic diversity and population size estimates of the eight populations. Wijdefjorden and North Isfjorden were identified as admixed populations

| Population      | n   | AR  | AR_{P} | AR_{P, sub} | H_{O} | uH_{E} | F_{IS} (95% CI) | N_{e} ± SE | N_{d} | N_{e} (95% CI) |
|-----------------|-----|-----|--------|-------------|-------|-------|----------------|------------|-------|---------------|
| North Spitsbergen | 27  | 2.07| 0.04   | 0.20        | 0.24  | 0.30  | 0.19 (0.08; 0.28) | 2,678 ± 400 | 1.84  | 36.2 (11.1; ∞) |
| Wijdefjorden     | 29  | 2.58| 0.01   | —           | 0.37  | 0.39  | 0.03 (-0.08; 0.17) | 572 ± 81   | 1.97  | 219.9 (36.3; ∞) |
| West Spitsbergen | 98  | 2.42| 0.01   | 0.02        | 0.33  | 0.35  | 0.07 (0.02; 0.13)  | 1,887 ± 241 | 2.88  | 100.2 (47.8; 437.9) |
| North Isfjorden  | 19  | 2.45| 0.01   | —           | 0.35  | 0.38  | 0.09 (-0.03; 0.19) | 1,730 ± 184 | 2.87  | 38.5 (13.5; ∞)  |
| Central Spitsbergen | 119 | 2.55| 0.02   | 0.07        | 0.43  | 0.43  | 0.00 (-0.05; 0.05) | 5,581 ± 650 | 2.97  | 221.1 (106.1; 1593.1) |
| South Spitsbergen | 28  | 2.35| 0.09   | 0.13        | 0.27  | 0.34  | 0.20 (0.02; 0.36)  | 648 ± 82   | 2.26  | 21.6 (10.8; 67.2) |
| East Svalbard    | 50  | 2.36| 0.09   | 0.15        | 0.29  | 0.34  | 0.16 (0.07; 0.31)  | 3,289 ± 1,085 | 1.38  | 108.5 (35.7; ∞)  |
| Nordaustlandet   | 23  | 2.08| 0.13   | 0.23        | 0.30  | 0.30  | 0.25 (0.12; 0.37)  | 1,922 ± 710 | 0.77  | 63.3 (14.0; ∞)   |

Abbreviations: AR, rarefied allelic richness averaged over loci; AR_{P}, rarefied allelic richness of private alleles; AR_{P, sub} same as AR_{P} but estimated without admixed populations; H_{O}, observed heterozygosity; uH_{E}, unbiased expected heterozygosity; F_{IS}, inbreeding coefficient with significant estimates indicated in bold; n, sample size; N_{e}, estimated census population size with standard error (SE); N_{d}, population density (N_{e}/km^2); N_{e}, effective population size based on linkage disequilibrium with 95% jackknifed confidence intervals.

### FIGURE 3 Spatial variation of individual heterozygosity (H_{ind}, i.e., the proportion of heterozygous loci in individuals based on 19 microsatellite loci) in 411 Svalbard reindeer (black dots). Estimates of individual heterozygosity were spatially interpolated using a thin plate spline regression. The density-weighted range center is indicated by the large blue dot.

### 3.3 Effective population sizes and gene flow

Estimates of contemporary N_{e} were lower than 500 in all populations, and lower than 50 in North Spitsbergen, South Spitsbergen, and North Isfjorden (Table 2). The N_{e}/N_{d} ratio was clearly highest in Wijdefjorden.
from South Spitsbergen to East Svalbard and Central Spitsbergen. However, mutation-scaled migration rates were significantly higher from Central Spitsbergen to South Spitsbergen and East Svalbard than in the opposite direction. Patterns of recent gene flow indicated highest migration rates from West Spitsbergen (migration rate $m = 0.09$) and Central Spitsbergen ($m = 0.14$) to North Isfjorden, with very little migration in the opposite direction ($m < 0.02$; Table S7). Other estimates of recent gene flow were generally low ($m < 0.05$) with no clear patterns of directionality, indicating high insularity of populations during the past few generations.

### 3.4 Landscape genetics

Genetic differentiation among natural sites increased with geographic distance (IBD: Table 3; Figure 5a). However, IBR explained twice the amount of variation in genetic distances than IBD when using least-cost path or circuit theory distances accounting for landscape fragmentation with low conductance for glaciers ($c = 0.01$) and open water ($c = 0.05$; IBR scenario 1: Table 3; Figure 5b,c). Models accounting for high connectivity across winter sea ice performed considerably better, especially when using circuit theory distances (IBR scenario 2: Table 3; Figure 5d,e). Mantel correlations in circuit theory and least-cost path analyses were highest for landscape models with areas covered by winter sea ice during, respectively, ≥75% and ≥99% of the time from 1986 to 2015, and using sea ice conductance values near 0.7 and 0.9, respectively (Figures S3 and S4). Therefore, genetic differentiation was not explained by IBD alone, but was strongly influenced by landscape barriers (open water and glaciers) and dispersal across winter sea ice.

Landscape distances among natural sites were moderately correlated between IBD and circuit IBR scenario 2 (Mantel's $r = .44$; $p = .002$) and between least-cost IBR scenario 1 and circuit IBR scenario 2 (Mantel's $r = .48$; $p < .001$). All other landscape distances were strongly correlated (Mantel's $r > .70$; $p < .001$; Table S8). Partial Mantel correlations after correcting for circuit distance of IBR scenario 2 were only significant between genetic distance and least-cost distance of IBR scenario 1 (Mantel's $r = .47$; $p < .001$).

### Table 3 Results from landscape genetics analysis among natural sites of Svalbard reindeer

Mantel correlations are given for the relationship between Rousset’s pairwise differentiation ($F_{ST}/(1 - F_{ST})$) and either Euclidian (IBD), least-cost path or circuit theory distances (IBR). Landscape models of IBR accounted for barrier effects (glaciers, open water, and topography) only (IBR 1), or both barrier effects and connectivity across sea ice (IBR2).

| Landscape distance       | Mantel test | MLPE   |
|--------------------------|-------------|--------|
|                           | $R$         | $\beta \pm SE$ | $R^2$ marginal | $R^2$ conditional | $\Delta AIC_c$ |
| Circuit IBR 2            | 0.78***     | 0.161 ± 0.019 | 0.64       | 0.70     | 0.00 |
| Least-cost IBR 2         | 0.70***     | 0.139 ± 0.019 | 0.53       | 0.69     | 4.24 |
| Least-cost IBR 1         | 0.63***     | 0.119 ± 0.019 | 0.42       | 0.63     | 13.06 |
| Circuit IBR 1            | 0.64***     | 0.116 ± 0.020 | 0.42       | 0.61     | 18.09 |
| Euclidian IBD            | 0.48**      | 0.079 ± 0.018 | 0.20       | 0.52     | 28.59 |

Note: Results from maximum likelihood population effects models (MLPE) show standardized effect size ($\beta$) with standard errors (SE), marginal and conditional $R^2$, and differences in sample-size-corrected Akaike’s information criterion ($\Delta AIC_c$) from the model with lowest $AIC_c$. $p$ Values of Mantel correlations computed based on 10,000 permutations: **$p < .01$; ***$p < .001$.
p = .008). Accordingly, including both landscape distances in the same MLPE model slightly improved the explained variance of genetic distance (standardized effect size with standard errors $\beta \pm SE$: intercept = 0.334 ± 0.019, circuit IBR2 = 0.124 ± 0.017, least-cost IBR1 = 0.066 ± 0.016; $R^2_{\text{marginal}} = .69$; $R^2_{\text{conditional}} = .71$).

At the individual level, genetic distances among pairs of reindeer were positively correlated with geographic distances (IBD: Mantel’s $r = .36$; $p < .001$; MLPE $R^2_{\text{marginal}} = .22$; $R^2_{\text{conditional}} = .44$), but more strongly correlated with least-cost distances (IBR scenario 1: Mantel’s $r = .40$; $p < .001$; MLPE $R^2_{\text{marginal}} = .34$; $R^2_{\text{conditional}} = .53$; and IBR scenario 2: Mantel’s $r = .41$; $p < .001$, MLPE $R^2_{\text{marginal}} = .35$; $R^2_{\text{conditional}} = .53$; Figure S5). Note that the high statistical power arises from the high number of observations ($n = 294$ reindeer; 86,436 pairwise distances; Luximon, Petit, & Broquet, 2014) and that circuit distances were not computed for all pairs of individuals (Material S4).

The importance of sea ice connectivity was also evident from the strong genetic differentiation between sampling sites of North Spitsbergen and nearby reintroduced sites in West Spitsbergen, where reindeer had previously been extirpated due to overharvesting (Tables 1 and S3). When comparing these reintroduced sites with natural, non-source sites, a tendency for a negative relationship between genetic distance and landscape distance was observed (Figure S6). Such an inverse relationship suggests a lack of migration-drift equilibrium among these sites, likely related to the rare sea ice cover along the west coast in the past few decades (Figure 1b; Onarheim et al., 2014).

### Discussion

This study demonstrates how the metapopulation genetics of an island ungulate can be shaped by a combination of landscape barriers and interactive effects of past and current anthropogenic drivers of extinction–recolonization dynamics. Whereas glaciers and open sea were clearly strong barriers to gene flow (Figure 5b,c), connectivity strongly increased in areas with frequent winter sea ice (Figure 5d,e). However, these landscape effects were modified by past conservation actions following harvest-induced extirpations, causing strong differentiation between reintroduced and natural sites over very short geographic distances. Clustering analyses resulted in a delineation of eight populations; six separate populations for each of the
six genetic clusters and two populations showing mixed cluster origins (Figure 2). Peripheral populations had lower genetic diversity, lower effective population sizes, and higher inbreeding coefficients than the central population, Central Spitsbergen (Table 2; Figure 3). Furthermore, estimates of directional gene flow based on coalescence theory suggested evolutionary source-sink dynamics with Central Spitsbergen as the main source (Figure 4). These observed patterns in genetic variation, structure, and connectivity suggest major anthropogenic impacts on the metapopulation dynamics and genetics of Svalbard reindeer.

Like Peary caribou, connectivity between island populations of Svalbard reindeer depends on sea ice corridors (Jenkins et al., 2016; Jenkins, Yannic, Schaefer, Conolly, & Lecomte, 2018). However, genetic differentiation in island populations of Peary caribou was largely explained by IBD as sea ice in the Canadian high Arctic has been relatively continuous in space and throughout the year (Comiso, Meier, & Gersten, 2017; Jenkins et al., 2016). Sea ice occurrence in coastal waters of Svalbard is more heterogeneous in time and space due to the inflow of warm Atlantic water in the west and cold Arctic water in the east. As our range-wide sampling covered this strong gradient in sea ice, we observed a marked IBR between natural sites, which explained more than three times the variation in genetic distance than IBD (Table 3). The strong genetic divergence and patterns of IBR also reflect the sedentary behavior of Svalbard reindeer (Côté et al., 2002; Hansen et al., 2010; Loe et al., 2016; Tyler & Orlitsland, 1989). This is in sharp contrast with the migratory behavior of Peary caribou and other Rangifer (Jenkins et al., 2016; Miller et al., 2005), as well as the high dispersal capability of, for example, Arctic fox across sea ice (Ehrich, Carmichael, & Fuglei, 2012; Geffen et al., 2007).

The observed sea ice effect likely reflects both short- and long-term patterns of divergence, indirectly modified by past harvesting. Genetic differentiation depends on patch area and connectivity as well as the age of the population (Cosentino, Phillips, Schooley, Lowe, & Douglas, 2012) and colonization trajectories (Le Corre & Kremer, 1998). Therefore, genetic differentiation was likely weaker when reindeer could rely on dispersal across sea ice to recolonize patches where reindeer had been extirpated by harvesting, particularly in the east of Svalbard (e.g., from Edgeøya to Barentsøya; Norderhaug, 1970). However, less frequent sea ice and ice-covered fjords have likely led to greater divergence in recolonized locations along the west coast, and we even observed a tendency for an inverse relationship between genetic and geographic/ecological distance when accounting for reintroduction effects (Figure S6). Historical connectivity among remnant populations, particularly between Central Spitsbergen and the islands in East Svalbard, was evident from evolutionary estimates of gene flow (Figure 4). Genetic differentiation among extant populations has thus been influenced by sea ice connectivity and landscape barriers in the past. However, the observed source-sink dynamics are likely influenced by extirpations or quasi-extinctions from overharvesting if populations inhabiting the outer range experienced a slower recovery due to a harsher climate and/or prolonged bottleneck effects, such as increased genetic load (Kirkpatrick & Jarne, 2000). In addition, reduced gene flow as an indirect effect of overharvesting, that is, due to depleted source populations and extirpations of connecting patches, has likely contributed to the distinct genetic clustering and high $F_{ST}$ values among the extant North Spitsbergen, Nordaustlandet, and East Svalbard populations.

Svalbard reindeer were assumed to be extirpated in South Spitsbergen before the 1820s, but this region was recently recolonized by a gradual southward expansion from Central Spitsbergen since the 1960s (Nolderhaug, 1970). However, the low genetic diversity and the strong genetic differentiation between these two populations suggest that South Spitsbergen may have held a small remnant population after harvesting was banned. Nevertheless, continuous range expansion can result in a transient increase in population divergence and gradual loss of genetic diversity due to cumulative founder effects (Le Corre & Kremer, 1998). This is especially the case for expansions resembling a linear stepping-stone model as colonists are sampled from consecutively newly founded populations. The high Arctic muskox (Ovibos moschatus) experienced such a gradual decline in genetic diversity following successive colonizations of high Arctic Canadian islands and Greenland, leading to extremely high genetic differentiation among populations (Hansen et al., 2018). Hence, in Svalbard reindeer, a harvest-induced extirpation followed by successive founder events is the most logical explanation for the genetic signature in South Spitsbergen, especially considering the small available habitat patches and fragmented landscape along the southwest coast. A southward recolonization is also supported by the differential estimates of long-term gene flow (Figure 4) and most likely involved dispersal across ice-covered fjords in winter.

A similar southward expansion has occurred from the strongly divergent, remnant population in North Spitsbergen. However, a general lack of sea ice and ice-covered fjords in recent decades has prevented introgression with the reintroduced population in West Spitsbergen (Figures 1b and 2; Onarheim et al., 2014). Indeed, the naturally recolonized Mitrahalvøya is separated by merely 16 km of open water from the nearest reintroduced site, but $F_{ST}$ values are as high as 0.41 (Table S3). Although reindeer are able swimmers, recolonizations of the peninsulas south of Ny-Ålesund and the easternmost island Prins Karls Forland were most likely mediated by dispersal across sea ice (Hansen et al., 2010). The genetic signature strongly supports the notion that reindeer would still have been extirpated in this region if they had not been reintroduced to Ny-Ålesund in the late 1970s. In comparison, introgression between two (or more) clusters mediated by dispersal across fjord ice was evident in both North Isfjorden and Wijdefjorden. The latter site in particular appears to be an important contact zone with elevated levels of genetic variability owing to strong divergence between the source (North Isfjorden and/or Central Spitsbergen) and recipient (North Spitsbergen) populations (see Maudet et al., 2002). This may explain why our analyses of recent gene flow failed to detect recent immigration to Wijdefjorden, but not to North Isfjorden (Table S7).

Today, Svalbard reindeer have recolonized their historical range, but recently recolonized populations are still recovering
from past overharvesting (Le Moullec et al., 2019). This ongoing extinction–colonization process is supported by the observed patterns of gene flow and strong genetic differentiation across the subspecies’ distribution range. Furthermore, the effect of past harvesting through extirpations and reintroductions on the current metapopulation genetic structure was modified by recent loss of sea ice and reduced connectivity, restricting dispersal and gene flow. The strongest negative trends in Arctic sea ice have been observed around Svalbard and the northern Barents Sea (Comiso et al., 2017; Onarheim et al., 2014; Stroeve et al., 2007). Therefore, our study provides an early warning that continued sea ice decline linked to global warming will lead to increased genetic isolation and population differentiation through genetic drift, not only in Svalbard reindeer but also other Arctic wildlife, such as Peary caribou (Jenkins et al., 2016, 2018; Mallory & Boyce, 2019), Arctic fox (Geffen et al., 2007; Norén et al., 2011), polar bear (Laidre et al., 2018), and muskox (Hansen et al., 2018). Indeed, the populations of Svalbard reindeer that were reintroduced into West Spitsbergen are already facing increased demographic isolation due to sea ice loss (Nilsen et al., 2016; Pedersen et al., 2018), which increases their vulnerability to current climate change and the increased frequency of extreme winter weather events (Hansen, Aanes, Herfindal, Kohler, & Sæther, 2011; Peeters et al., 2019). Particularly rainy winters, causing extensive ice encapsulation of the tundra vegetation, usually occur over large areas and contribute to spatially synchronized population dynamics (Hansen et al., 2019; Peeters et al., 2019). Strong synchronization by climate is expected to increase the extinction risk of a metapopulation (Engen, Lande, & Sæther, 2002; Heino, Kaitala, Ranta, & Lindström, 1997). In Svalbard reindeer, the observed synchrony in short-term abundance fluctuations appears to be buffered by contrasting long-term trends in local abundances, linked to the spatial heterogeneity in climatic change effects (Hansen et al., 2019). Nevertheless, increased levels of inbreeding and the loss of genetic diversity are of great concern in small populations living at the outer limits of the distribution range, making them more vulnerable to extinction (Frankham, 1998; Saccheri et al., 1998). Future source–sink dynamics will likely be more characterized by dispersal across land and increased isolation of populations surrounded by the sea. Contact zones with high genetic mixture, such as Wijdefjorden, are therefore important focal populations for management and conservation. Further studies should implement modern genetic tools, such as SNPs or whole-genome sequencing, to detect local adaptation in small isolated populations and genome level changes under climate change.

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CONFLICT OF INTEREST

We declare no conflicts of interest.

DATA AVAILABILITY STATEMENT

The data that support the findings of this study are available from the corresponding author upon reasonable request.

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