

Euclidean distances versus effective pathways: a GIS approach for testing the effects of dispersal constraints on spatial genetic structure of Black grouse (Tetrao tetrix Aves, Tetraoninae) populations

Julie Pierson, Ariane Bernard Laurent, Yann Magnani, Stéphane Marin, Sonia

Said, Alain Caizergues

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7	Julie Pierson ¹ , Ariane Bernard-Laurent ¹ , Yann Magnani ¹ , Stéphane Marin ¹ , Sonia Saïd ²
8	and Alain Caizergues ³
9	
10	1 ONCFS, Direction des études et de la recherche, CNERA Faune de Montagne, soit
11	90, impasse "Les Daudes", BP 41, 74 320 Sévrier, France
12	2 ONCFS, Direction des études et de la recherche, CNERA Cervidés Sanglier, 1 place
13	Exelmans 55000 Bar Le Duc, France
14	3 ONCFS, Direction des études et de la recherche, CNERA Avifaune Migratrice, 53,
15	Rue Russeil, 44000 Nantes, France
16	
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23	4200 words, 5 figure and 3 tables
24	Corresponding author: Alain Caizergues (a.caizergues@oncfs.gouv.fr)
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25 Abstract.

Radio-tracking of juveniles and genetic studies have previously suggested that mountain 26 27 ridges may represent barrier to dispersal for the Black grouse. In the present study, we 28 assessed the potential of a more straighforward approach to test the effect of these 29 topological features on Black grouse dispersal. We simply compared correlations 30 between genetic distances and geographic distances assuming that if mountain ridges 31 truly represent barriers to dispersal for the Black grouse, correlation coefficients should 32 improve when using pairwise distances taking into account dispersal barriers (or 33 effective pathways) instead of straight-line (Euclidean) distances. We computed 34 effective pathways using the Geographic Information System (GIS) package 35 ARCVIEW 3.2 and its extension PATHMATRIX especially designed to determine 36 least-cost pathways from a landscape matrix in which all habitat types are given a 37 specific cost. Whatever the type of distance used (Euclidean distances versus effective 38 pathways) a highly significant isolation-by-distance effect was observed. The 39 correlation between geographic and genetic distances slightly improved when the 40 effective pathways instead of Euclidean distances were used. However, these 41 improvements were very small suggesting either that mountain ridges are "permeable" 42 barriers to dispersal or that the prerequisite for using the GIS approach efficiently were 43 not fully satisfied.

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48 Keywords: Aves, dispersal constraints, genetic differentiation, GIS, least cost analysis,
49 microsatellites.

50 Introduction

51 Describing population spatial structure and understanding its determinants is a key issue 52 for conservation/management purposes (Frankham et al 2002). In this perspective, the 53 challenge it is not only to assess habitat requirements, but also to identify landscape 54 components that effectively impede movements between populations (Coulon et al 55 2006; Vos et al 2001).

56 The Black grouse (*Tetrao tetrix*) is a galliform bird living in forested areas of northern and central Europe, but also in open moors of British Isles as well as in the 57 58 Alps. Over the last 50 years, human activity has greatly diminished in numerous areas 59 formerly occupied by the Black grouse, resulting in the extirpation of the species owing 60 to the closure of forest canopy. In the Alps, the distribution of Black grouse has 61 remained stable over the 25 years, but populations have declined because of changes in 62 land use practices (spreading of ski resorts, land abandonment) (Arlettaz et al 2007; Bernard-Laurent 1994). Legally speaking, the Black grouse is now considered as a 63 64 major conservation concern (Storch 2000), although the species is not threatened but 65 included in Annex I of the EU Birds Directive (79/409/EEC) as a species for which special habitat conservation measures are required. 66

67 Numerous studies aiming to better understand the causes of the decline of Black 68 grouse populations in the Alps have been carried out since 20 years (Bernard 1981; 69 Caizergues and Ellison 1997; Caizergues and Ellison 2000; Caizergues and Ellison 70 2002; Ellison et al 1988; Magnani 1988). A major aim of these studies was to collect 71 data on dispersal movements in order to better understand the spatial organisation of 72 Black grouse populations. In this context, radio-tracking studies showed that dispersing 73 juveniles were reluctant to cross high mountain ridges culminating above their natural 74 habitat (i.e., above 2300 m a.s.l.) and more generally that movements between

75 favourable habitat patches were always performed along forest drainages i.e. following 76 the axes of valleys (Caizergues and Ellison 2002). Later, indirect evidence supporting 77 that high ridges impede Black grouse dispersal could be gathered using a genetic approach. More particularly, the genetic differentiation between local populations was 78 79 found to be almost seven times higher in the Alps where Black grouse have to cope with 80 mountains ridges constraining dispersal than in Finland where no such constraints exist 81 (Caizergues et al 2003b). Unfortunately, because the stronger genetic differentiation in 82 the Alps compared to Finland could also be due to a naturally more fragmented habitat 83 in the former, firm conclusions concerning the influence of mountain ridges on the 84 spatial organisation of Black grouse populations could not be derived.

85 In the present paper, we adopted a straightforward approach to assess the effects 86 of mountains ridges on the spatial genetic structure of Black grouse populations in the 87 Alps. Using a least-cost-pathway analysis with a Geographic Information System 88 software, we computed distances incorporating the effects of mountain ridges on Black 89 grouse dispersal (distances hereafter called effective pathways). Then we compared the 90 strength of the correlation between genetic distances and these effective pathways to the 91 correlation obtained between the same genetic distances and (straight-line) Euclidean 92 distances. This kind of approach has previously been successfully applied to several 93 biological models although never in birds (Hamilton et al 2006; Michels et al 2001; 94 Spear et al 2005; Stevens et al 2006; Verbeylen et al 2003). We predicted an 95 improvement of the correlation when using effective pathways instead of Euclidean 96 distances if, as expected, mountains ridges effectively impede dispersal movements of 97 Black grouse.

98 Materials and methods

99 Study area

100 The study site almost covers the whole French Alps as well as portions of Swiss and 101 western Italian Alps, representing an area of about 45,000 km². The topography consists 102 in a succession of valleys, ridges and high mountains with altitudes mainly comprised 103 between 200 and 4800 m a.s.l (Figure 1).

104 Sampling and genetic methods

Tissues (skin, muscles, toes, liver) were collected from Black grouse legally shot by
hunters during the autumns 1997 – 2000. This period does not exceed the average
generation time of the Black grouse which has been estimated to 4-5 years (Caizergues
and Ellison 1997).

Analyses were performed both at the individual and population levels. Four hundred and thirty six samples grouped in 18 entities (populations) were used in the analyses at the population level. Analyses performed at the individual level included a sample of 152 individuals selected to cover as homogeneously as possible the larger area possible (Figure 1). We did not use the whole sample in the individual based approach because it would have implied the calculation of 94830 least-cost paths which represented a computing time exceeding 80 hours.

Each individual retained was characterised with at least 10 of the 14
microsatellites loci whose properties are described in details in Caizergues *et al.*(2003b). DNA extraction were performed using DNeasy Tissue Kits (Qiagen Inc.)
following manufacturer protocols. Amplifications were carried out on PTC 100
thermocycler using Touch Down procedures (Don et al 1991) in volumes of 10 μl
containing 1 μl of extraction product (see Caizergues et al 2003b for details).

122 Statistical analyses of genetic data

Previous analyses of a subsample of the material presented here did not detect significant linkage disequilibrium between pairs of loci (Caizergues et al 2003b). Therefore, only tests of Hardy-Weinberg proportions and the basic statistics averaged over loci are presented here.

127 Departures from Hardy-Weinberg expectations were assessed for each locality 128 using exact tests implemented in GENEPOP (3.3 version, 2001, Raymond & Rousset 129 1995; Rousset 2000).

130 Modelling geographical distances.

131 All geographical distances were computed using the GIS software ARVIEW version 132 3.2. Effective pathways (or least cost distances) measure the shortest distance between 133 two entities obtained by minimising the costs of moving through the various 134 components of the landscape defined by the user. We modelled effective pathways 135 (between individuals and between populations) with the PATHMATRIX extension of 136 ARCVIEW 3.2 (Ray 2004) using a three components landscape matrix derived from the 137 topological model of the study area. These three landscape components were defined as 138 follow:

1391)Natural habitats of Black grouse ranging between 1600 and 2300 m140a.s.l. The species is assumed to move with minimal constraints in his141natural habitats.

- 142 2) Mountain ridges culminating above the altitude of 2300 m a.s.l. which
 143 are assumed to act as total barriers to Black grouse movements.
- 1443)Habitats below 1600 m (valleys), which may be costly to cross but that145do not represent absolute barriers to dispersal.

The delineation of landscape components was achieved on ARCVIEW 3.2 using a raster topological model of the study area with a 200 m resolution grid. Tests performed with 20, 50, 100, 200, 500 m grids showed that a 200 m resolution was the best compromise between computation time and a reliable representation of the putative constraints to dispersal.

Effective pathways were computed using the value 1 as an arbitrary cost of moving across natural habitats. To make mountain ridges act as absolute barriers to movements their cost was set at 10000. Valleys up to 10 km wide do not seem to represent a problem for Black grouse (Caizergues et al 2003b). However, undertaking flights in open spaces might entail higher risks in terms of predation than moving under the sheltering canopy of favorable habitats. For this reason, a cost of 2 was set to habitats below an altitude of 1600 m (valleys).

The Euclidean distances are simply the straight-line distances between two entities. They don't take landscape structure into account. These distances were also computed on PATHMATRIX but the same cost for the three habitats types was set. In the analyses performed at the population level, pairwise geographical distances were calculated between the barycenter of individuals constituting each of the populations.

163 *Correlations between genetic and geographical distances.*

Tests of correlation between genetic and geographical distances were performed in SPAGeDi version 1.2 software (Hardy and Vekemans 2002) with a permutation procedure equivalent to the test of Mantel (Manly 1986; Mantel 1967). SPAGeDi was also employed to compute the various estimates of genetic distances/kinship coefficients. SPAGeDi is of particular interest compared to other genetic packages because it allows importing matrixes and propose a larger collection of genetic distance restimators than most other packages.

171 The statistic Fst/(1-Fst) was used as an estimate of the genetic distances between 172 populations because it is expected to vary linearly with the logarithm of distances and 173 therefore provide unbiased estimates of regression slopes (Rousset 1997). Loiselle's 174 kinship coefficient (Loiselle et al 1995) and the â distance (Rousset 2000) were used to 175 characterize the genetic relationships between individuals. Loiselle's kinship coefficient 176 offers both a low bias and a low sampling variance (Hardy and Vekemans 2002). The â 177 distance suffers from a higher sampling variance but can be compared to analyses 178 performed at the level of populations because it is a generalization of the Fst/(1-Fst) 179 estimator.

180 **<u>Results</u>**

181 Euclidean distances vs. effective pathways

182 The characteristics of both Euclidean distances and effective pathways between both 183 individuals and populations are summarized in table 1. The differences between the two 184 types of distances were highly significant (Paired t, test all P values <0.001). The 185 figures showing the effectives pathways are very interesting because they allow 186 identifying corridors necessary to maintain dispersal between populations under the 187 hypothesis that mountains ridges 2300 m a.s.l. are dispersal barriers. At least five 188 corridors between the West and East side of the Alpine massif are easily visible on both 189 maps (Figure 2).

190 Correlations between genetic and geographical distances

Microsatellite polymorphism ranged from 2 to 17 alleles per locus, with a mean
of 6.5 (Table 2). Among the 18 populations used in the analyses, only population
10displayed any significant departure from Hardy-Weinberg expectations (Table 2).
However, in this population only two 2 loci out of 14 displayed any significant
departure from Hardy-Weinberg expectations.

196 The average F_{st} value between populations was 0.049 (SE=0.0039, range: 0 -197 0.11).

198 Correlations between the genetic distances / kinship coefficient and the 199 logarithms of geographical distances were highly significant either using the individual 200 or the population approach (P<0.001, Tables 3, Figure 3). Regression slopes obtained 201 using Fst/(1-Fst) and the \hat{a} were of the same order of magnitude with confidence 202 intervals largely overlapping (Tables 3). This suggests that the approach based on 203 individuals is as reliable as the approach based on populations despite the high sampling 204 variance of the \hat{a} estimator.

At the population level, the correlation between genetic and geographic distances slightly improved when using effective pathways instead of Euclidean distances (Table 3). At the level of the individual, this correlation improved only for Loiselle's kinship coefficient (Table 3). The poorer performance of the analyses using the â was probably due to the higher sampling variance of estimator (Vekemans and Hardy 2004).

211 **Discussion**

212 An increasing number of studies explore landscape connectivity using either 213 Geographic Information Systems (e.g., Graham 2001; Larkin et al 2004; Ricketts 2001; 214 Verbeylen et al 2003) or populations genetics approaches (Bouzat 2001; Caizergues et 215 al 2003b; Knutsen et al 2000; Roland et al 2000). In comparison, studies adopting a 216 combination of these two approaches in an integrated manner are scarse (but see 217 (Arnaud 2003; Hamilton et al 2006; Michels et al 2001; Spear et al 2005; Stevens et al 218 2006; Vigneri 2005; Vos et al 2001). Moreover, in most of these studies analyses were 219 carried out at the population rather than the individual level and the integration of

landscape structure almost always involved complex assumptions rarely derived fromdirect observations of the biological model studied.

Here we report the results of an explicit landscape genetics analysis in a bird species, specifically the interaction between landscape features and demography and microevolutionnary processes (Manel et al 2003). We also show, that maintaining gene flow between Italian and Franco-Swiss Black grouse populations will depend upon our capacity to preserve favourable habitats along five mains corridors.

Although most birds usually display good dispersal propensities and therefore little, or no genetic spatial structure at all, we detected not only a highly significant isolation-by-distance effect but also provided support for the negative influence of landscape features such as high ridges on dispersal.

However, we concur that as such, our results do not allow to derive firm conclusion about the exact influence of high ridges on Black grouse dispersal, chiefly because only two tests out of three detected an improvement of the correlation coefficient between genetic and geographic distances when the presence of dispersal barriers was taken into account. Moreover, these improvements were small.

236 Several non exclusive hypotheses might explain why models including effective 237 pathways did not perform much better than model that didn't include them. Firstly, the 238 high variance inherent to all genetic distances estimators (Vekemans and Hardy 2004) 239 would render difficult the detection of small differences between concurrent models. 240 Secondly, barriers occur mainly between the West and East sides of the Alps (Figures 2) 241 while most of our samples come from the West side where numerous passes allow 242 dispersal between neighbouring forest drainages. In other words, our sampling design 243 was probably not optimal to fully exploit the landscape genetics approach simply 244 because the two distances are strongly correlated (Figure 4). Thirdly, owing to the high

245 levels of fragmentation of Black grouse habitats in the Alps, genetic drift could have a 246 more decisive influence on genetic structure of Black grouse populations than 247 previously suspected (Caizergues et al 2003b). This could explain both the higher 248 between-populations genetic differentiation in the Alps than in Finland (Caizergues et al 249 2003b) as well as the relatively poor performance of models taking explicitly dispersal 250 barriers into account. However, in the Alps, Rock ptarmigan populations that 251 experience a level of fragmentation comparable to the Black grouse, did not display a 252 high degree of genetic divergence, with average F_{st} values of 0.01 against 0.049 to 0.07 253 for the Black grouse (present study, Caizergues et al 2003a; Caizergues et al 2003b). 254 Mountain ridges are part of the habitat of Rock ptarmigan meaning that the species, in 255 contrast to Black grouse, doesn't have to cope with barriers to dispersal.

Another explanation to the relative poor performance of models taking dispersal into account could be that the mountains ridges do not represent absolute barriers to dispersal contrary to our initial assumption. Although we cannot total discard this hypothesis, direct observations strongly support these topological features do act as natural barriers to dispersal for the Black grouse. These include radio-tracking data of *ca* 100 individuals (Caizergues 1997; Caizergues and Ellison 2002).

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- **Table 1.** Summary of characteristics of Euclidean and effective pathways between
- 267 populations and individuals (expressed in kilometers).

	Popul	ations	Indiv	iduals	
	Euclidean distances	Effective pathways	Euclidean distance	Effective pathways	
Maximum	284.42	362.59	300.20	371.96	
Minimum	17.62	24.12	0.569	0.70	
Average	132.05	176.32	119.26	160.41	
Median	120.64	171.99	114.40	159.82	
Standard deviation	71.35	88.14	70.73	90.37	
Paired samples t-test	P<0.001		P<0.001		

Table 2. Summary statistics of the variability of the 14 microsatellites loci in each of the 18 sampled populations.

Population	Ν	Mean number of alleles	Minimum number of alleles	Maximum number of alleles	Average Ho	Average He
01 (Veglia/Devero)	11	6.07	3	11	0.71	0.77
02 (Vigezzo)	58	8.57	5	13	0.72	0.77
03 (Valais uptstream)	15	6.93	4	13	0.79	0.78
04 (Valais downstream)	24	8.14	4	13	0.79	0.78
05 (Giffre)	15	7.71	4	13	0.82	0.79
06 (Aravis/beaufortain)	54	8.64	5	14	0.75	0.78
07 (Tarentaise)	13	6.21	3	10	0.68	0.75
08 (Maurienne upstream)	19	6.36	3	10	0.75	0.74
09 (Maurienne downstream)	49	7.29	4	12	0.73	0.75
10 (Queyras)	57	7.14	4	12	0.71	0.75
11 (Champsaur)	10	5.50	3	9	0.80	0.75
12 (Haut-Var)	18	6.07	4	9	0.76	0.76
13 (Haute-Tinée)	24	6.43	4	13	0.75	0.71
14 (Haute Roya)	45	6.79	4	11	0.75	0.73
15 (Anzasca)	6	5.14	3	7	0.78	0.75
16 (Aosta)	5	4.93	2	7	0.83	0.83
17 (Briançonnais))	6	5.00	3	8	0.73	0.77
18 (Ubaye)	7	5.21	3	8	0.66	0.75

Table 3. Correlation tests between the logarithm of geographical distances (Euclidean distances vs effective pathways) and genetic distances / kinship coefficient between individuals (Loiselle's coefficient of kinship and Rousset's â) or between populations (Fst/(1-Fst). r^2 = determination coefficient, a = intercept, b = regression slope value, CI = confidence interval of b and P-value of the correlation test.

	Genetic distance	Geographic distance	r ²	a	b	95%CI-inf	95%CI-sup	P-value
	Loiselle's	Euclidean distances	0.051	0.211	-0.018	-0.020	-0.0172	< 0.001
Individuals	coefficient	Effective pathway	0.052	0.215	-0.018	-0.019	-0.0171	< 0.001
Inuividuais	â	Euclidean distances Effective pathways	0.036 0.036	-0.186 -0.187	0.020 0.019	0.016 0.016	0.0238 0.0231	<0.001 <0.001
Populations	Fst / (1-Fst)	Euclidean distances Effective pathways	0.439 0.452	-0.275 -0.314	0.027 0.030	0.021 0.023	0.0346 0.0377	<0.001 <0.001

279

280 Legend of figures

281	Figure 1. Location of the study area and spatial distribution of individuals (open
282	circles) and populations barycentres (black squares). $(1 = \text{Veglia/Devero}, 2 = \text{Vigezzo}, 3$
283	= Valais upstream, 4 = Valais downstream, 5 = Giffre, 6 = Aravis/Beaufortain, 7 =

- 284 Tarentaise, 8 = Maurienne upstream, 9 = Maurienne downstream, 10 = Queyras, 11 =
- 285 Champsaur, 12 = Haut-Var, 13 = Haute-Tinée, 14 = Haute-Roya, 15 = Anzasca, 16 =
- Aosta, 17 = Briançonnais, 18 = Ubaye)

287

Figure 2. Least-cost paths as estimated by PATHMATRIX based on (a) Euclidean distances between populations barycentres, (b) effective pathways between populations barycentres and (c) effective pathways between individuals.

291

Figure 3. Relationship between Euclidean distances (log transformed) and genetic distances/kinship coefficients, and between effective pathways (log transformed) and genetic distances/kinship coefficients.

295

Figure 4. Relationship between pairwise Euclidean and effective pathways between populations barycentres. r^2 = determination coefficient.









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