


Potential distribution of a climate sensitive species, the White-winged Snowfinch *Montifringilla nivalis* in Europe

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Summary

The White-winged Snowfinch *Montifringilla nivalis nivalis* is assumed to be highly threatened by climate change, but this high elevation species has been little studied and the current breeding distribution is accurately known only for a minor portion of its range. Here, we provide a detailed and spatially explicit identification of the potentially suitable breeding areas for the Snowfinch. We modelled suitable areas in Europe and compared them with the currently known distribution. We built a distribution model using 14,574 records obtained during the breeding period that integrated climatic, topographic and land-cover variables, working at a 2-km spatial resolution with MaxEnt. The model performed well and was very robust; average annual temperature was the most important occurrence predictor (optimum between c.-3°C and 0°; unsuitable conditions below -10° and above 5°). The current European breeding range estimated by BirdLife International was almost three times greater than that classified as potentially suitable by our model. Discrepancies between our model and the distribution estimated by BirdLife International were particularly evident in eastern Europe, where the species is poorly monitored. Southern populations are likely more isolated and at major risk because of global warming. These differences have important implications for the supposed national responsibility for conservation of the species and highlight the need for new investigations on the species in the eastern part of its European range.

Keywords: species distribution model, climate change, mountains

Introduction

Detailed knowledge of the occurrence and spatial arrangement of animal species and their preferred habitats is a basic, key requirement for most research applications and for species and habitat conservation. In this climate change era, climate-sensitive species (and environments) are of particular concern, as they are highly threatened by ongoing modifications in climatic parameters and by changes in the habitat prompted by such modifications.

The White-winged Snowfinch *Montifringilla nivalis* (hereafter, Snowfinch) in Europe (which includes the whole range of the nominate subspecies *M. n. nivalis*) inhabits a range restricted to higher elevations of central and southern mountain massifs, and is among the species most threatened by climate change on the continent (Brambilla *et al.* 2017a). Predicted changes in distribution and connectivity among suitable sites (Brambilla *et al.* 2017a), potential increase in the impact of human alteration to alpine habitats (Brambilla *et al.* 2016), modifications in snow-cover and snow-melt date (Brambilla *et al.* 2018a, Resano-Mayor *et al.* 2019) and in foraging habitat (Brambilla *et al.* 2018b) suggest a very concerning status of the species, at least in the Alps, where the species has been most studied.

Despite the dramatic future prospects for the species, the Snowfinch had been little investigated until a few years ago and was classified as 'Least Concern' in the last Red List of European Birds (BirdLife International 2015). In recent years, evidence for range contractions and/or population declines has accumulated (Knaus *et al.* 2018, Scridel *et al.* 2017) and the Snowfinch is now regarded as a flagship species for high-elevation taxa and habitats threatened by climate change. The current breeding distribution of the species is reasonably well known only for the Alps, the Pyrenees and the Cantabrian Mountains, whereas for the Italian pre-Alps and Apennines, available data are less complete and accurate; in the Balkans, data are even scarcer and knowledge of the species' occurrence is still poor in several areas (see www.snowfinch.eu).

We aimed to identify those mountain areas in Europe with suitable habitats and climates for breeding Snowfinches and to compare them with the current estimated geographical range during the breeding season, in order to: i) provide a better understanding of the breeding distribution over the continent; ii) identify gaps in current knowledge by comparing our output with the currently

known distribution as defined by BirdLife International (BirdLife International and Handbook of the Birds of the World, 2018; hereafter, “BirdLife distribution”) and thus to pinpoint new, potentially important, areas to be explored (Bourg *et al.* 2005, Brambilla *et al.* 2009); and, iii) compare the potential importance for conservation of the different countries hosting Snowfinch populations according to current knowledge (BirdLife distribution relative to the breeding season) and to our model, respectively. All this information is essential to better understand distribution patterns for conservation-related purposes, such as identifying key areas and national responsibilities for Snowfinch conservation.

Methods

Data were collected as widely as possible throughout European countries, in the framework of different studies carried out by the authors (Brambilla *et al.* 2017b, 2018b, Resano-Mayor *et al.* 2017, 2019, Strinella *et al.* 2007) and by national parks and local institutions. All these data were collected in the form of spatially georeferenced observations. In addition, citizen science data (i.e. collected by the public) were gathered through online databases (www.ornitho.at, www.ornitho.ch, www.ornitho.it), after official requests specific to the purposes of the project (data downloaded and received in the period December 2018–April 2019). Data were therefore occurrence-only.

We collated all data for the period 1979–2018 and removed all records not satisfying the temporal and spatial requirements. We only kept records of breeding individuals (atlas code suggesting breeding or ‘territorial’ behaviour), or observed during the species’ breeding season (15 May–31 July, i.e. the ‘core period’ of the breeding season, during which most individuals are likely to be involved in reproduction). All data recorded at a resolution coarser than 1-km were discarded.

Overall, 14,574 records met the above criteria. Twenty-four observations were from elevations between 750 and 1,500 m asl, and 96 observations were recorded above 3,000 m (but below 3,700 m). Most data (14,454) were collected between 1,500 and 3,000 m (mean $2,247 \pm 316$ SD). Note that the recorded elevation (assigned based on the digital elevation model used for model building; see below) could be not exact, because of the resolution of the records (from one or a few meters, up to 1 km); even an approximation of a very few hundreds of meters on mountain slopes may result in elevation differences of up to hundreds of metres because of the steep terrain.

To develop the distribution model, we considered climatic, topographic and land-use/land-cover variables. Climatic data were gathered by the CHELSA database for the period 1979–2013 (Karger *et al.* 2017) at a 30 arc seconds (~ 1 km) resolution. Topographic variables were derived from a 30-m digital elevation model in GRASS GIS (Neteler *et al.* 2012). Land-cover data were derived from Corine Land Cover 2012 (European Environment Agency 2016). All environmental data were expressed at the scale of 2 x 2 km cells, taking the average values for climatic predictors, slope and solar radiation, and the proportional cover for land-use/land-cover categories.

The same grid was used to process Snowfinch records and to create background points. All cells with one or more Snowfinch records were considered as occupied cells and used as occurrence cells for modelling (thus avoiding duplicates and reducing the number of records from the most intensively sampled areas). We thus obtained 2,473 independent 2 x 2 km cells occupied by the species. The distribution of occurrence data and hence of occupied cells was not uniform over the European range. The Alps hosted the largest amount of data, followed by the Cantabrian Mountains. There were fewer data from the Pyrenees and the Apennines, but they were nonetheless adequately sampled (Figure S5 in the online supplementary material). Notably, environmental conditions in the Apennines were representative of Mediterranean mountains and this area may thus provide a test-site to check the model’s ability to predict species distribution over the poorly sampled Mediterranean mountains.

The distribution model was developed using MaxEnt (hence, a presence-background method not requiring absence data), under the package ENMeval (Muscarella *et al.* 2014) in R (R Development

Table 1. Variables selected in the final MaxEnt model and their relative effect, and other variables tested but not included. Irrespective of their inclusion, variables are subdivided into two groups (potentially important predictors vs. other variables). The number before the land-cover variables represents the CORINE category. A short description of the potential importance of the variables presumed to be potentially important for the species is also provided in the last column.

Variable	Permutation importance	Effect	Potential importance
<i>Potentially important predictors of environmental suitability for Snowfinches</i>			
23 – broad-leaved forest	0.29	negative	unsuitable habitat locally reaching high elevation
24 – coniferous forest	1.06	negative	unsuitable habitat reaching high elevation
26 – natural grassland	0.16	positive	positive effect – important foraging habitat
31 – bare rocks	0.25	positive	positive effect – foraging and nesting habitat
32 – sparsely vegetated areas	0.09	positive	positive effect – potential foraging habitat
annual average temperature	94.27	quadratic (optimum between -3° and 0°C)	important driver of species occurrence
precipitation of the warmest quarter	1.88	positive	potential effect because of impact on vegetation (e.g. seed production)
slope	2.01	quadratic (optimum at c. 20°)	generally associated to slopes
average solar radiation			important for microclimate
precipitation of the coldest quarter			potential positive effect (snow-cover in spring is crucial for foraging)
<i>Other variables presumably less important - tested but not included in the models</i>			
2 – discontinuous urban fabric, 3 – industrial or commercial units, 12 – non-irrigated arable land, 13 – permanently irrigated land, 15 – vineyards, 16 – fruit trees and berry plantations, 17 – olive groves, 18 – pastures, 20 – complex cultivation patterns, 21 – land principally occupied by agriculture with significant areas of natural vegetation, 22 – agro-forestry areas, 25 – mixed forest, 27 – moors and heathland, 28 – sclerophyllous vegetation, 29 – transitional woodland-shrub, 34 – glaciers and perpetual snow, 35 – inland marshes, 40 – water courses, 41 – water bodies			

Core Team 2016). 150,000 background cells were randomly identified within the most intensively sampled countries in the study region (irrespective of Snowfinch records), i.e. Portugal, Spain, France, Italy, Switzerland, Austria and Slovenia. All those countries harbour Snowfinch populations or are close to existing ones (Portugal) and hence potentially reachable by the species.

We used only the variables which did not lead to multicollinearity issues for modelling, on the basis of the generalized Variance-Inflation Factor (gVIF; Zuur *et al.* 2009), removing variables with values >5 . Variables tested included those selected in the final model shown in Table 1, in addition to other climatic and land cover variables. We tested climatic variables potentially relevant for Snowfinches as those related to temperature and snowfall regimes, and all land cover variables occurring within the study area, in three different combinations: i) only the variables of most importance for the species according to current knowledge, ii) the latter plus those more represented in mountain environments (i.e. the cover of pastures, mixed forest, moors and heathland, glaciers and permanent snow, water courses and water bodies), iii) all variables listed in Table 1. Table S1 in the online supplementary material contains further details about the potential effect of environmental variables.

We split occurrence data into four bins (using the function 'checkerboard 2'), i.e. in four partitions of spatially independent occurrence records, which were used for model training and validation over independent datasets. AUC (Area Under the Curve of the receiver operating characteristic -ROC- plot) and omission rates on test data (Muscarella *et al.* 2014) were considered (Table S1). Eight different values of the regularization multiplier were tested (from 0.5 to 4 with increments of 0.5), and the one leading to the model with the lowest AIC was selected. Then, the variables with the weakest effects (with permutation importance and percentage contribution both lower than 1) were removed from the model, which was trained again with the eight different regularization multiplier values. This process was repeated until we obtained a final best-supported model. The logistic model output was reclassified into three different suitability categories to facilitate interpretation: unsuitable (lower than maximum training sensitivity plus specificity threshold), partly suitable (between maximum training sensitivity plus specificity threshold and 10th percentile), and definitely suitable (higher than 10th percentile). These two thresholds are those generally adopted for binary reclassification of models produced using MaxEnt (Liu *et al.* 2005, 2013, Engler *et al.* 2014).

To refine the predictions at a finer scale in order to provide outputs as precise as possible, we limited the potential occurrence of suitable sites to the region-specific elevation belt inhabited by the species, by filtering out as unsuitable all areas below 1,500 m (this being a conservative value, selected to exclude areas at elevations where Snowfinches had never been found in recent years in Europe). The changes due to this post-modelling correction were almost imperceptible at the working scale (see Figure S1).

Finally, we performed a country-based comparison of the suitable breeding area as predicted by our model, with that estimated to be occupied by breeding Snowfinches according to the BirdLife distribution, the most widely used and comprehensive estimation of the species' geographic range currently available. Similarly, we compared the estimated percentage of the species' European range within each country based on our models and the BirdLife distribution. These assessments are particularly relevant because the percentage of a species' population within a country (likely correlated with the percentage of range within it) is among the criteria used to define a country's responsibility for the conservation of a given species (BirdLife International 2017). This analysis was performed i) using raster data and the relative approximation (resulting in an irrelevant difference over such a broad scale), and ii) without any correction for elevation. In addition, the analysis was restricted to the countries hosting the species as a breeder according to the BirdLife distribution, and to those closely neighbouring Snowfinch populations in other countries (Andorra and Bulgaria).

Results

The output of the distribution model was identical for the three sets of variables tested. The most supported MaxEnt distribution model performed well and was very robust, displaying the same

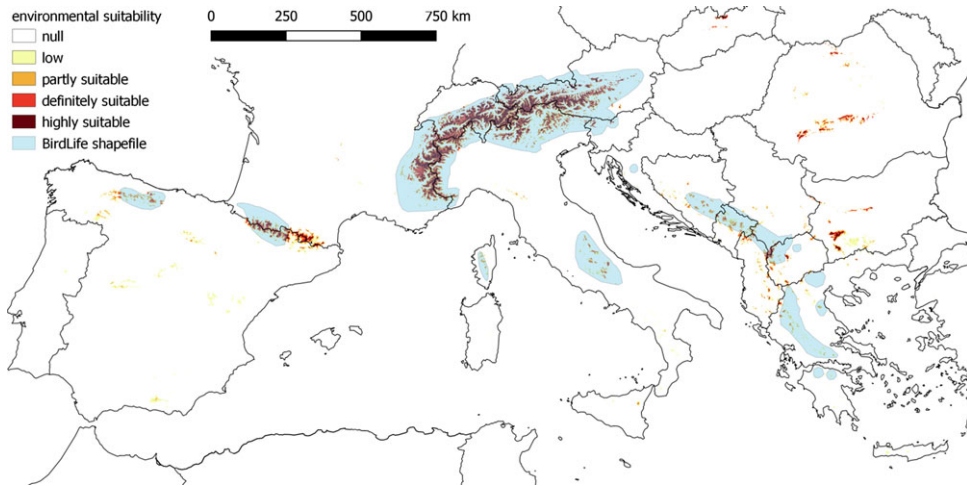


Figure 1. Modelled (our work) and reported (BirdLife) distribution of White-winged Snowfinch in Europe. Partly suitable sites are those with suitability above the maximum training sensitivity plus specificity threshold, definitely suitable sites those higher than the 10th percentile (see text for further details); highly suitable areas are those with suitability close to the maximum value. Areas below 1,500 m were considered as unsuitable (see also Figure S1).

AUC and omission rates very close to the expected values, over all the four independent bins (Table S1).

Average annual temperature (bio1) was by far the most important predictor of species occurrence, having the only notable effect according to permutation importance; it had a quadratic effect, with an optimum for Snowfinch occurrence between c. -3 °C and 0°, and unsuitable conditions below -10° and, especially, above 4°–5°. The other climatic predictor included in the final model was precipitation of the warmest quarter (bio18); this variable was slightly positively correlated with environmental suitability for Snowfinches, as were natural grassland, bare rocks and sparsely vegetated areas. Forest cover (both broadleaved and coniferous) had a negative effect on environmental suitability, whereas a quadratic relationship was identified between suitability and slope, with a peak at intermediate slope values (Table 1). The effect of environmental variables on habitat suitability for Snowfinches is shown in Figure S4.

Modelled and reported (BirdLife) distributions were generally similar (Figure 1), although there were some important discrepancies (Figures S2–S3). Similarly, the estimated suitable area per country displayed important differences; in general, the area inhabited by Snowfinches according to the BirdLife distribution was greater (249,221 km²) than the area suitable or potentially suitable estimated by our model (91,044 km²). The percentages of the European species' range hosted by each country calculated according to the model and the BirdLife distribution were significantly correlated (Spearman's rho 0.71, $P = 0.001$), but revealed some important discrepancies in the relative relevance of each national population (Table 2).

Discussion

For a species so highly threatened by climate change (Brambilla *et al.* 2018b, Resano-Mayor *et al.* 2019), it is essential to accurately define current and potential distribution in order to plan conservation strategies and implement measures in relevant sites; distribution modelling may help considerably in that sense (Engler *et al.* 2017). Our study provides a detailed and spatially explicit identification of the potentially suitable breeding areas of the nominate Snowfinch

Table 2. Comparison between suitable area and relative distribution share per country according to our distribution model and the BirdLife distribution, respectively. Country extent mirrors the raster resolution of the distribution model. List of abbreviations used in the table: pot_suit: area potentially suitable; def_suit: area definitely suitable; tot_suit: sum of potentially and definitely suitable area; area_BLD: extent of BirdLife distribution within the country; %_suit_model: percentage extent of the country potentially or definitely suitable; %_BLD: percentage extent of the country occupied by the species according to BirdLife distribution; %_species_model: percentage of Snowfinch European range within the country according to the potentially or definitely suitable areas (considering only the countries listed in the table); %_species_BLD: percentage of Snowfinch range within the country according to BirdLife distribution

Country	Model				BL distr	National scale		European range	
	country extent	pot_suit	def_suit	tot_suit	area_BLD	%_suit_model	%_BLD	%_species_model	%_species_BLD
France	549264	4892	9860	14752	44073.4	2.7	8.0	16.2	17.7
Spain	498684	3208	2696	5904	12692.2	1.2	2.5	6.5	5.1
Italy	301152	7648	12988	20636	64509.8	6.9	21.4	22.7	25.9
Germany	357578	1084	508	1592	6088.7	0.4	1.7	1.7	2.4
Greece	132184	736	204	940	20977.2	0.7	15.9	1.0	8.4
Bulgaria	111036	1212	1120	2332	0.0	2.1	0.0	2.6	0.0
Serbia	88456	968	380	1348	5864.8	1.5	6.6	1.5	2.4
Austria	83808	7604	12436	20040	45910.0	23.9	54.8	22.0	18.4
Croatia	56488	88	8	96	622.3	0.2	1.1	0.1	0.2
Bosnia and Herzegovina	51196	1104	224	1328	6899.1	2.6	13.5	1.5	2.8
Switzerland	41320	4084	11980	16064	23795.0	38.9	57.6	17.6	9.5
Albania	28488	940	472	1412	1618.4	5.0	5.7	1.6	0.6
Macedonia	25396	924	736	1660	4641.7	6.5	18.3	1.8	1.9
Slovenia	20248	516	364	880	4146.5	4.3	20.5	1.0	1.7
Montenegro	13872	1280	348	1628	7218.0	11.7	52.0	1.8	2.9
Andorra	460	120	264	384	0.0	83.5	0.0	0.4	0.0
Lichtenstein	160	24	24	48	160.0	30.0	100.0	0.1	0.1

subspecies *Montifringilla nivalis nivalis*, which inhabits European mountains. Analyses confirmed the primary importance of climatic (in particular temperature) and topographic (slope) parameters in determining environmental suitability and hence distribution in Snowfinches (see Table 1 and Figure S5), in concordance with assessments carried out over finer spatial scales (Brambilla *et al.* 2016, 2017a). As expected, environmental suitability for Snowfinches increases with the cover of natural grassland and sparsely vegetated areas, which provide key foraging habitats (Brambilla *et al.* 2017b, 2018), and with bare rocks, which provide potential nesting sites.

Snowfinch occurrence data mostly came from the Alps, where the species has a rather broad distribution; however, the model correctly predicted occurrence in other, more isolated mountain chains, such as the Cantabrian Mountains and the Apennines. The latter served as a test-site for model performance in the Mediterranean region; these results were encouraging, as the predicted distribution represented well the actual occurrence of the species, including some isolated habitat patches irregularly occupied by the species which were classified as potentially suitable by the model (E. Strinella and M. Brambilla pers. obs.). Despite this, we cannot exclude the possibility that the potential region-specific variations in habitat associations could to some extent affect model predictions for poorly sampled areas, such as the Balkans. Further occurrence data should be collected in south-eastern Europe to exclude such potential limitations, as well as to improve knowledge and model accuracy for this relatively poorly known portion of the European range of the species. Nevertheless, we believe that the predicted suitability i) makes sense also for south-eastern Europe (see below and supplementary material), and ii) can be used to highlight areas requiring further investigation.

In southern European mountains, the availability of potentially suitable areas according to the model is much more restricted and fragmented than the overall species range suggested by the BirdLife distribution, and Snowfinches likely occupy smaller, and more isolated, suitable patches there (Figure 1), in areas characterised by higher average temperatures. As a consequence, populations breeding in these mountains (Cantabrian Mountains, Apennines, Balkans) appear to be at greater risk because of climate change impacts. Future studies based on the comparison of the species' autecology and demography between large suitable areas (e.g. Alps, Pyrenees) and smaller, more marginal ones (e.g. Cantabrian Mountains, Apennines, Corsica, several areas in the Balkans) may help to understand the potential impacts of climate change by allowing the inclusion of large gradients of climate and isolation.

Despite a general concordance with the currently known distribution, modelling also highlighted some important differences with the current range of the species as defined by BirdLife International (BirdLife International and Handbook of the Birds of the World 2018), especially (but not exclusively) for eastern Europe. Generally, the BirdLife distribution includes most of the areas predicted to be suitable by our model (apart for some parts of the Cantabrian Mountains and Pyrenees), but also includes rather large unsuitable areas compared to our models. Some of these discrepancies are likely due to the different spatial resolution of the two approaches; approximate polygons surrounding occupied areas and not considering elevation are very likely to 'overpredict' occurrence compared to a dedicated species distribution model for such a species strongly linked to high elevation. Nevertheless, some discrepancies definitely require further investigation (see Figures S2–S3 for detailed comments and comparisons). Some of the differences arose because the model identified potentially suitable sites in mountain chains currently believed not to harbour breeding Snowfinches, such as the Carpathians, Rila and Rhodope mountains, and Mount Etna. In some of these sites, further investigations should be carried out (see below and supplementary material for further discussion). Suitable areas in the Balkans show only a moderate concordance with the Snowfinch range according to the BirdLife distribution (Figure S2). Even if this is largely attributable to the relatively low spatial resolution of the BirdLife distribution, which needs to be considered at a larger scale, most areas encompassed by it in Greece are located below 1,500 m, and suitable areas are located largely outside the supposed species range. In Albania, Macedonia and Bulgaria, several suitable patches occur outside the known range of the species. Some of those sites could potentially host important populations that need to be preserved for the conservation of the

species in eastern Europe. According to the BirdLife distribution, Snowfinches do not breed in Bulgaria. However, our model suggested the availability of large suitable patches in the south-western part of the country. Even if the species is currently considered as non-breeding in the country, historical records of breeding pairs in the 1960s were reported (although regarded as not reliable) for Mt. Rila (Ivanov 2011). According to our model, this site has the highest suitability in the country. Further intensive field work is needed in this key unknown area.

These differences have important implications for the supposed national responsibility for the species' conservation, and such discrepancies result also in very large differences in the expected frequency of a species within a country (Table 2). For example, the estimated area occupied by Snowfinch in Greece according to the BirdLife distribution is 22 times higher than the suitable/potentially suitable area estimated by our model. Considering the BirdLife distribution, Greece should host a proportion of the European Snowfinch breeding range that is very close to that harboured by Switzerland, whereas our model suggests that the proportion of the European Snowfinch range in Switzerland is almost 18 times that found in Greece. A further example relevant to those regarding the perceived frequency of the species within each country is provided by Montenegro, where Snowfinches should occur in more than half of the country according to the BirdLife distribution, whereas our model suggests that less than 12% of the country is potentially suitable for the species, and only 2.5% is definitely suitable.

The concerning situation of the species in Europe has triggered several studies on its ecology, distribution and demography in several areas in Europe, namely the Cantabrian Mountains, the Pyrenees, Corsica, the Alps and the Apennines (see ongoing initiatives on www.snowfinch.eu). Unfortunately, we are not aware of any detailed study in the Balkans. The results of our work suggest the need for new investigations on the species in the eastern portion of its European range, where the basic distribution of the species is also poorly known. In addition, such areas need to be investigated in the future, as they could provide key features for planning conservation measures for this species. Within such sites, Snowfinches may experience climatic conditions that are now lacking in other European mountains, but that will be potentially much more widespread in the future. In addition, an increase in the availability of occurrence data from the Balkans could confirm or increase the accuracy of model predictions for this area. Further site-specific studies to better understand the ecology and population dynamics of the species are required across the entire geographic range to better assess Snowfinch conservation status and promote management and other conservation actions for the most emblematic (and highly threatened) alpine passerine of Europe.

Supplementary Materials

To view supplementary material for this article, please visit <http://dx.doi.org/10.1017/S0959270920000027>.

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