Identifying drivers of spatio-temporal variation in survival in four blue tit populations

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Abstract

In a context of rapid climate change, the influence of large-scale and local climate on population

demography is increasingly scrutinized, yet studies are usually focused on one population.

Demographic parameters, including survival, can vary significantly across populations of a given

species, depending on global or local climatic fluctuations but also on many other population-specific

parameters such as breeding density, habitat naturalness, predation or parasitism. Such ecological

differences between populations could lead to different paces-of-life (POL), whereby populations

where reproductive investment is higher and behaviours are bolder display lower survival. We use

here long-term (19 to 38 years) monitoring datasets from four Mediterranean populations of blue

tits (*Cyanistes caeruleus*) to investigate the effects of sex, age class, large-scale and local climate

temporal variation and population breeding density, on adult survival, using Capture-Mark-Recapture

modelling. Environment heterogeneity in these four populations (two in evergreen and two in

deciduous forests) has been linked to strong multi-trait phenotypic variation, suggesting blue tits in

deciduous forests display faster POL compared to their conspecifics in evergreen habitats. The

present results show heterogeneity in average survival probabilities across the four populations,

with, as predicted, lower survival in the 'fast' blue tits occupying deciduous habitats. Interestingly,

the year-to-year variation in survival probabilities was synchronous among populations. This suggests

that regional environmental conditions could drive survival fluctuations across populations. However,

breeding densities were not correlated across populations, and we found no evidence that adult

survival is correlated with either large-scale or local, climate temporal variation in these four blue tit

populations. Finally, two of the focal populations displayed a linear temporal decrease in adult

survival over the study period. Overall, this multi-site study shows that blue tit parental survival from

one spring to the next can vary substantially across years, in a synchronous way across populations,

yet the climate indices we used are not correlated with the temporal variation. This calls for further

investigations in other potential drivers such as resource (in particular insect) abundance, predation

or parasitism.

Introduction

Many temporal variations in ecological systems can be decomposed in cycles (e.g. daily, seasonal, multi-annual), in longer term trends and in remaining 'noise' (e.g. year-to-year variation) (Wolkovich et al., 2014). These variations often result from abiotic environmental changes over time, such as climate or local weather variations, which in turn result in biotic responses to these changes, e.g. morphological, behavioural, physiological, phenological and/or demographic variations at the population scale. Multiple studies have shown that climate can influence numerous biological processes and biodiversity patterns, with ecological and evolutionary consequences (e.g. Norberg et al., 2012, Woodbridge et al., 2021). In particular, the recent ecological literature informs us that meteorological year-to-year variation can influence population demography in plants (Chang-Yang et al., 2016, Dalgleish et al., 2011) and animals (Selonen et al., 2016, Wood et al., 2016), while recent trends in climate change cause temporal trends in demographic components and their variance across many different taxa of plants (Williams et al., 2015) and animals (Massardier-Galata et al., 2017), including birds (Alves et al., 2019, Gamelon et al., 2017). For example, in polar bears, survival of cubs to recruitment is highly dependent on their mother's body condition in autumn, which itself depends on weather conditions, while the population demography of this species is highly impacted by the increasing reduction of sea ice availability, resulting in strong conservation concerns (Laidre et al., 2020). However, the demographic consequences of climate variation, and the link between climate, ecological factors, and demographic effects is still insufficiently explored (see reviews Chevin et al., 2013, Visser & Gienapp, 2019).

Recent studies have shown that population density can play a major role in the impact of climate (and of climate-induced changes in traits) on population dynamics. In particular, McLean and colleagues have argued that species or populations with strong density-dependent effects on population dynamics will have more robust demographic rates (*i.e.* survival and fecundity) when facing strong climate fluctuations because density-dependent processes can buffer negative demographic consequences of climate change (McLean et al., 2016). An elegant example of such buffering effect of density comes from a study of Dutch great tits *Parus major,* where warmer springs result in a detrimental mismatch between the bird breeding phenology and their main prey seasonal peak (Visser et al., 2006). While this phenology mismatch has fitness consequences for the birds, such that spring warming translates into stronger selection for earlier breeding, an increased mismatch does not result in decreased population growth (Reed et al., 2013) because of density-dependent regulation. During warm springs, great tits have a reduced breeding success, yet their fledglings show increased survival due to relaxed competition. Such examples highlight the

importance of considering density-dependent effects when exploring demographic consequences of
climate change.

In a context of large-scale rapid climate change, it is important to understand whether the
meteorological and climatic variations with which populations are (and will be) confronted are likely
to have a similar impact on their demography, depending on their location. The spatial synchrony of
demographic parameters and local population dynamics (Robert, 2009), or on the contrary their
divergence (Cuervo & Moller, 2013), is a key element of species dynamics on a large spatial and
temporal scale (Siriwardena et al., 1998). Demographic parameters, including survival, can vary
significantly across populations of a given species, depending for instance on large scale climatic
fluctuations (Post & Stenseth, 1998, Mazerolle et al., 2005), local climatic conditions and resource
availability (Winkler et al., 2014, Senner et al., 2017), predation and parasitism (Watson, 2013,
DeCesare et al., 2014), interspecific competition (Gustafsson, 1987), and many factors related to
human activities (Hõrak & Lebreton, 1998, Porneluzi & Faaborg, 1999, Cartwright et al., 2014).
Variation in survival across populations can also arise because of differences in population age-
structure or sex ratio (Loison et al., 1999, Clutton-Brock & Isvaran, 2007).

Here, using a long-term monitoring dataset on four Mediterranean populations studied across 19 to
38 years, we investigated the effects of sex, age class, large-scale and local climate temporal
variation and population breeding density, on the survival of breeding adults in a temperate
passerine bird, the Blue tit (*Cyanistes caeruleus*). Two of these populations are located in habitats
dominated by the evergreen holm oak (*Quercus ilex)* and the two others in habitats dominated by
the deciduous downy oak (*Quercus humilis*). This environmental heterogeneity has been linked to
strong phenotypic variation in blue tits, whereby birds from the two habitats differ in their
morphology, behaviour, colour ornamentation, physiology and life histories (Blondel et al., 2006,
Charmantier et al., 2016). Overall, the phenotypic divergence between the two types of habitats is
consistent with a divergence in pace-of-life syndrome (Réale et al., 2010), with individuals in the
deciduous habitat displaying a faster pace-of-life (e.g. bolder personalities and larger clutches hence
higher reproductive investment) and individuals in evergreen populations a slower pace-of-life
(Charmantier et al., 2016, Dubuc-Messier et al., 2017). Preliminary results suggest that survival
probabilities are concordant with a faster pace-of-life syndrome in deciduous habitats, with lower
adult survival (see Table 1 in Dubuc-Messier et al., 2017).

A previous study by Grosbois and colleagues analyzed blue tit survival in three of our four study
populations over a more limited period of time (8 to 16 years, datasets running until 2000) (Grosbois
et al., 2006). Results from this study indicated that adult survival differed considerably both among
years and among populations and that the pattern of interannual variation in survival was similar

among populations, suggesting that adult survival in these blue tit populations was influenced by

environmental factors, such as climate, that operate at a relatively large spatial scale. In particular,

adult survival was correlated with both local-scale weather conditions and a large-scale tropical index

in early summer: rainfall in the Sahel.

Based on this context of a divergence in pace-of-life across habitats and on past investigations, we

expected to find a difference in adult survival between habitat types (with lower survival

probabilities in deciduous habitats), as well as between age classes (with lower survival in older

individuals, Bouwhuis et al., 2012) and sexes (although the sex advantage in survival may vary

annually, see for ex. Clobert et al., 1988). We also expected year-to-year variation in adult survival to

be correlated among populations as well as with global and local climatic indices (Grosbois et al.,

2006). Finally, population density during the breeding season is also expected to have a negative

impact on subsequent adult survival (e.g. Frederiksen & Bregnballe, 2000).

124   Methods

1. Monitored populations

Data was collected in four Mediterranean wild populations of blue tits *Cyanistes caeruleus*, a

European cavity-nesting non-migratory passerine bird, in forest plots equipped with nest boxes and

dominated either by the deciduous downy oak (*Quercus pubescens*, site names starting with D-) or by

the evergreen holm oak (*Quercus ilex*, site names starting with E-). All sites are situated in the

Mediterranean region, three of them (E-Pirio: Lat 42.38; Long: 8.75; D-Muro: 42.55; 8.92; E-Muro:

42.59; 8.96) on the island of Corsica and one (D-Rouviere: Lat 43.66; Long 3.67) on the Mainland in

southern France (Figure 1, Blondel et al., 2006, Charmantier et al., 2016 for details). All four

populations were monitored as part of the same long-term research program, but monitoring did not

begin at the same time in each site (E-Pirio: 38 years of monitoring from 1979 to 2016, D-Rouviere:

26 years (1991-2016), D-Muro: 24 years (1993-2016), E-Muro: 19 years (1998-2016)).

Populations were monitored using nest boxes, which blue tits readily use for breeding and generally

prefer to natural cavities (Newton, 1994). The monitoring of almost all breeding individuals in the

focal populations was ensured by using a high density of nest boxes compared to the abundance of

natural cavities in the various sites (neighbouring boxes are 50m apart). Every year, nest boxes were

checked at least weekly during the blue tit breeding season (March to June). Adult breeders were

captured in the boxes when the nestlings were 10-15 days old. The Centre de Recherches sur la

Biologie des Populations d'Oiseaux (CRBPO) provided the permits under which capture and handling

of birds were conducted, as well as unique numbered metal rings that were used to identify every

(nestling and breeding) bird captured in the four populations. Plumage patterns were used to

determine sex and age class at first capture (1 year breeders vs breeders of 2 or more years) for

parents. Individuals whose minimum age at first capture had not been assessed were removed (3%,

n=170 of parents). In total, 5499 reproducing individuals were considered in this study (E-Pirio: 1562,

D-Rouviere: 1947, D-Muro: 1408, E-Muro: 582).

Figure 1: Location of the four focal populations of blue tits. The three Corsican sites (E-Pirio, D-Muro

and E-Muro) are located ~440km away from the mainland site (D-Rouviere). E-Muro is located ~6km

away from D-Muro and ~30km from E-Pirio. Two sites are in a deciduous oak *Quercus pubescens*

forest (D-Rouviere and D-Muro), and two in an evergreen oak *Quercus ilex* forest (E-Pirio and E-

Muro).

2. Large-scale climatic variation

Two large-scale climate indices were considered in this study as potentially influencing adult blue tit

survival in our populations:

-   The Mediterranean Oscillation Index (hereafter MOI, defined by Conte et al., 1989 and,

Palutikof et al., 1996 as the normalized pressure difference between Algiers (36.4°N, 3.1°E)

and Cairo (30.1°N, 31.4°E)). The MOI is a large scale climate index correlated to the North

Atlantic Oscillation index (NAO) (Lionello et al., 2006). It is linked to rainfall and climate in the

Mediterranean basin such that MOI is negatively correlated with precipitation and positively

correlated with winter and spring temperature (Sangüesa-Barreda et al., 2019). Daily MOI

data were obtained from https://crudata.uea.ac.uk/cru/data/moi/.

-    The Sahel Rainfall index (monthly data obtained from

167   http://research.jisao.washington.edu/data/sahel/), a tropical climate index.

The MOI was aggregated in two separate variables: winter MOI (from December 1st year t to

February 28th in year t+1) and summer MOI (from June 15th to September 15th in year t). A yearly

aggregation (from March 2nd year t to March 1st year t+1) was rejected as it was strongly correlated

to winter MOI (Pearson's correlation coefficient = 0.857). The Sahel Rainfall index was aggregated in

an early summer Sahel rainfall index (hereafter SRF) by adding the monthly values of June and July of

each year, as has been done by Grosbois and colleagues on three of these blue tit populations

(Grosbois et al., 2006).

3. Local meteorological data

Four local climatic variables were used in this study to test whether meteorological conditions drive

differences in survival probability variability in each population:

-    Spring and summer rainfall (hereafter SpringRF: aggregated rainfall between March 2nd and

September 15th of year t)

-    Autumn and winter rainfall (hereafter AutumnRF: aggregated rainfall between September

16th of year t to March 1st of year t+1)

-    Extreme heat events during the summer (hereafter EHE, number of extremely hot days

(average daily temperature in the 5% hottest of average daily temperatures in the studied

summers on this site) between June 15th and September 15th of year t), which indicates if

extreme climatic situations have been encountered during the considered summer

-    Hottest summer temperatures (hereafter HST, average mean daily temperature during the

10 hottest days between June 15th and September 15th of year t), which represents the global

harshness of the considered summer

Rainfall data were obtained by measurements from the meteorological station of Saint-Martin de

Londres (Lat: 43.78 Long: 3.73, approximately 14 km away from the D-Rouviere site) for the

mainland, and Calvi (Lat: 42.52; Long: 8.79, approximately 16 km away from E-Pirio, 11 km from D-

Muro and 16 km from E-Muro) for the Corsican sites. Temperature data at the Corsican sites were

obtained by regressing daily temperature measurements in the different sites with temperature data

from the meteorological station of Calvi over 4 years, and inferring the temperature in the sites over

the remaining years (see Table S1 in Appendix 1). At the D-Rouviere mainland site, temperature data

was obtained from the meteorological station of Saint-Martin de Londres.

4. Population density index

Breeding density was obtained by measuring the blue tit nest box occupation rates in a defined area

of each study site, where nest boxes were present across all years of study (E-Pirio: n=26 nest boxes

in the first 6 years of monitoring (1979-1984), then 54 to 62 nest boxes, D-Rouviere: n=65 nest boxes,

D-Muro: n=20 nest boxes the first year (1993), 32 to 38 boxes from 1997 to 1999 and 49 to 53 nest

boxes in the remaining years, E-Muro: n=20 to 44 nest boxes during the first 3 years of monitoring

(1998-2000), then 54 to 55 nest boxes).

5. Capture-Mark-Recapture modelling

Individual capture-recapture histories for the 5499 breeding individuals were analysed to provide

robust estimates of survival and recapture probabilities (respectively ϕ and P, Lebreton et al., 1992),

using a logit-link function. All analyses were conducted using the program E-SURGE (Choquet et al.,

2009b). Goodness-of-fit of models to the data was ensured for each dataset using the program U-

CARE (Choquet et al., 2009a), based on the Cormack Jolly Seber model for monostate models.

In several years, experiments were conducted in the population including brood size manipulations

that can highly alter adult survival (Nur, 1984a, Dijkstra et al., 1990). Capture-recapture histories of

the corresponding individuals were censored right, after the first session they experienced such a

fitness-changing experiment. However nestlings born during these experiments and later recaptured

as adults were not removed from the analysis. This necessary censoring resulted in a significant

decrease in observation numbers (-14.95% from 12 131 to 10 316 capture and recapture events in

the dataset).

6. Model selection

The Akaike's Information Criterion corrected for small sample size (AICc) was used for model

selection (Burnham & Anderson, 1998). A low AICc was considered revealing a good compromise

between the fit to the data (likelihood of the model) and the number of parameters used by the

model. The threshold for a significant difference between two models was set at two AICc points. In

case of a lower difference, the model with the lowest number of parameters was selected.

The first models (models integrating the four populations together, without temporal covariate) are

numbered from 1 to 73 (Table 1). Covariate models (models integrating the four populations, with

meteorological and population yearly covariates) are numbered 74 to 79 (Fig. 3B). Finally, additional

models were implemented separately for each population (models 80 to 111, Table 3 and Fig. 4B).

A first model selection was conducted on the four populations together, assessing the effects of

population, year, sex and age on recapture (*P*) and survival (ϕ) probabilities (Table 1). The starting

model (Model 72) included potential effects of population, sex, age and annual variations (hereafter,

year) on survival and potential effects of population, sex and year on recapture probabilities. Only

simple (first order) interactions between variables influencing survival and recapture probabilities

were considered, to enable robust biological interpretation of the results and avoid overcomplexity.

All subsequent models were nested in this one. The model selection started with simplifying

constraints on P. The best model structure for recapture (Model 56) was then retained and simpler

survival models were considered, removing step by step first order interactions between variables

until reaching a model with only 3 interactions (Model 12). All models nested in this one were then

tested. At each step of model selection (selection on P structure, removal of interaction terms in ϕ,

then selection of main effects in ϕ), new models were created and then sorted in descending AICc

order. The model numbers in Table 1 reflect this step-by-step selection protocol.

To assess the potential effect of large-scale climatic fluctuations on survival in our populations, three

potential large-scale climate indices (summer MOI, winter MOI, early summer Sahel Rainfall) were

tested on the four populations together. An analysis of deviance (hereafter *ANODEV*, Skalski et al.,

1993) was conducted for each candidate climate index to test for a linear or quadratic temporal

effect of this index (Figure 3B). It used as general model the best model from the previous selection

(Model 1) including additive effects of age, population and annual variations on survival, and as

reduced model the same without the temporal effect (Model 19). The ANODEV test allows to reject

the null hypothesis H_0, assuming that the considered temporal variable does not explain a significant

proportion of the deviance reduction between the temporally free model and the reduced model.

The null hypothesis was considered rejected when p-ANODEV < 0.05.

An analysis of deviance was also conducted to assess the effect of breeding density in year t on

survival between t and t+1, in the different populations (Figure 4B). In each population separately,

the model including additive effects of age and year on survival was considered as the general model,

while the reduced model included only an effect of age on survival. The effect of breeding density on

survival was considered significant if p-ANODEV < 0.05 or if the difference in AICc between the best

model and the breeding density model was higher than 2. The same analysis was conducted in each

population separately to test for the effect of local climatic variables (Table 3). A linear effect of time

was also tested in this selection, to assess the existence of a potential temporal trend of survival

probabilities in the different populations. All temporal covariates were centered and standardized.

Results

1. Goodness of fit

No significant violation of standard model assumptions was found while performing the goodness-of-fit tests using program U-CARE (Choquet et al., 2009a): for the 4-population monostate dataset $\chi^2(df=602)=366.499$, *p-value* = 1; for each population separately: $\chi^2(df=184)=85.346$, *p-value* = 1 for E-Pirio; $\chi^2(df=140)=117.024$, *p-value* = 0.922 for D-Rouviere; $\chi^2(df=149)=101.227$, *p-value* = 0.999 for D-Muro; $\chi^2(df=129)=62.903$, *p-value* = 1 for E-Muro.

2. Temporal variation in adult survival

Model selection indicated strong annual variation in survival probabilities across populations (Table 1, Figure 2). The effect of year was additive (see Table 1: Models 9, 23, 27, 44 and 51 for a comparison of additive and interactive effects of population and year on survival. ΔAICc(Model 9 – Model 23) = 33.59), which means that survival probabilities covaried across time in our 4 study populations (see Figure 2 for a visual testimony).

Figure 2: Annual survival for 2+ years adults in the four focal blue tit populations between breeding

events in year t and t+1, from 1979 to 2016. 95% confidence intervals are not represented for the

sake of readability. Estimates are from Model 18 (Table 1).

3. Effects of population, age and sex on recapture and survival probabilities

Our model selection indicated that recapture probabilities differed across populations and sexes

(Table 2) ranging from 0.673 (males in E-Pirio) to 0.866 (females in D-Rouviere). Additional model

selection carried out separately for the four populations indicated that the effect of sex on P only

occurred in E-Pirio (see Appendix 2 for additional model selection and Table 2 for estimates on P).

In our best models, age had an additive impact on survival across populations, which means that

while adult survival differed across populations, the effect of age on survival was similar across

populations (Table 1). In particular, 1st year adults had consistently markedly higher survival than

adults aged 2 or more in all populations. In contrast, model comparison suggested that there was no

effect of sex consistent in all populations (see model 1 vs. model 2 in Table 1) and the comparison of

survival estimates for the two sexes with models not constrained to additivity indicated that the

survival differences among sexes was weak and that the sex with the highest survival varied among

populations (Table 2). Males had higher survival than females in deciduous habitats (with survival

probability ranging from 0.419 to 0.479) while the reverse was true in the evergreen populations

(and survival ranged from 0.546 to 0.591).

	D-Rouviere	E-Pirio	D-Muro	E-Muro
	26 years of monitoring	38 years of monitoring	24 years of monitoring	19 years of monitoring
	1947 individual histories	1562 individual histories	1408 individual histories	582 individual histories
A) Annual Survival Estimates (φ)				
Adult survival (ϕ)	0.473	0.548	0.424	0.575
(estimates from Model 24)	[0.453 , 0.493]	[0.521 , 0.574]	[0.397 , 0.451]	[0.539 , 0.609]
ϕ (1 year adults)	0.491	0.597	0.449	0.596
(estimates from Model 21)	[0.459 , 0.522]	[0.547 , 0.645]	[0.400 , 0.498]	[0.522 , 0.666]
ϕ (2+ years adults)	0.458	0.526	0.412	0.567
(estimates from Model 21)	[0.430 , 0.486]	[0.494 , 0.558]	[0.380 , 0.445]	[0.525 , 0.608]
ϕ (Males)	0.479	0.546	0.429	0.556
(estimates from Model 33)	[0.450 , 0.509]	[0.508 , 0.583]	[0.390 , 0.467]	[0.505 , 0.606]
ϕ (Females)	0.467	0.549	0.419	0.591
(estimates from Model 33)	[0.439 , 0.495]	[0.513 , 0.585]	[0.382 , 0.457]	[0.543 , 0.638]
B) Recapture Probabilities Estimates (p)				
p (Males)	0.821	0.673	0.749	0.693
(estimates from Model 1)	[0.772 , 0.861]	[0.611 , 0.729]	[0.672 , 0.812]	[0.606 , 0.768]
p (Females)	0.866	0.857	0.701	0.740
(estimates from Model 1)	[0.819 , 0.902]	[0.804 , 0.898]	[0.625 , 0.767]	[0.664 , 0.803]

Table 2: Survival and recapture probability estimates and the influence of age and sex in the four focal populations of blue tits. 95% confidence intervals are enclosed in square brackets.

4. Effect of large-scale climate indices on adult survival

Figure 3A presents temporal variation in the three considered large-scale climate indices (summerMOI,

winterMOI and SRF) during the study period associated with models presented in Table 3. Among these

covariates, only a small positive correlation was found between summerMOI and SRF (Pearson's correlation

coefficient$_{summerMOI-SRF}$ = 0.362; *p-value* = 0.03). No correlation was found between winterMOI and the other

climatic covariates (Pcc$_{summerMOI-winterMOI}$ = 0.094, Pcc$_{winterMOI-SRF}$ = 0.025). No significant linear nor quadratic effect of

any of the tested large-scale climatic indices could be found on survival probability (Table 3).

316 **Figure 3**: Annual variation in the three considered climate indices during the study period.

Model number	Model Description	Number of Parameters	Deviance	ΔAICc	p-ANODEV	R^2
1	**φ (age + pop + year) P (sex.pop)**	**49**	**10320.7472**	**0**		
74	φ (age + pop + q_SRF) P (sex.pop)	15	10423.8673	34.594	0.21	0.084
75	φ (age + pop + summerMOI) P (sex.pop)	14	10427.5694	36.305	0.1797	0.051
	φ (age + pop + q_summerMOI) P (sex.pop)	15	10427.2602	37.9869	0.38	0.054
19	**φ (age + pop) P (sex.pop)**	**13**	**10433.2892**	**40.0183**		
77	φ (age + pop + SRF) P (sex.pop)	14	10432.8102	41.5458	0.7013	0.004
79	φ (age + pop + winterMOI) P (sex.pop)	14	10433.2112	41.9467	0.8771	0.0007

Table 3: Model selection procedure and significance of the effects of the considered climate indices on adult

survival (φ). Starting yearly dependant and constrained model for the ANODEV are highlighted in bold. Quadratic

effects are labelled with a "q_". AICc of best model (Model 1) is 10421.3358. SRF = Early Summer Sahel Rainfall;

MOI= Mediterranean Oscillation Index. The values of R² provided correspond to the proportion of temporal

variance in survival explained by the model covariates, computed through analysis of deviance.

5. Effect of local meteorological variables on adult survival

Despite the apparent similar temporal variation across the four populations (Figure 2), the results of our model

selection suggested important variation in adult survival across years only in the D-Rouviere population: ΔAICc =

47.61 between Model 80 and Model 82 (Figure 4B).

To further explore these annual fluctuations in D-Rouviere and the difference with the other populations, we

tested the effects of local weather variables (*i.e.* spring and summer rainfall, autumn and winter rainfall, extreme

heat events, hottest summer temperatures, see Methods for details), which displayed a difference in variability

between the Corsican (E-Pirio, D-Muro, E-Muro) and the mainland (D-Rouviere) populations (Table 4). The

existence of a linear temporal trend in survival was also tested in all four populations.

While significant effects were found for a negative linear trend in survival in D-Rouviere and E-Muro, no effect of

the tested meteorological variables was detected in any of the populations (Table 4).

6. Effect of breeding density on adult survival

Only one correlation was found between breeding density estimates across the populations: densities in E-Pirio

and D-Muro were negatively correlated (slope = -0.533, *p-value* = 0.007) (Figure 4A). A linear temporal trend was

found for population breeding density in the two most recently studied populations: D-Muro and E-Muro. In both

populations breeding density has increased during the study period. However, further analysis indicated that in all

populations, density estimates showed a significant linear increase over the first 20 years of the local study

program (see Figure 4A).

In only one of the four populations (E-Muro), breeding density significantly decreased the deviance of the model

(Figure 4B, p-ANODEV = 0.042). However, in the E-Pirio population, even though the ANODEV test did not yield

any significant result, the AICc was significantly decreased by the population density covariate. In both cases,

population density had a negative correlation with adult survival: in these two populations, springs with high nest

box occupation were followed by a year of low adult survival (Figure 4C).

A.

B.

Model Number	Model Description	Number of Parameters	Deviance	ΔAICc	p-ANODEV
	D-Rouviere			AICc(Model 80) = 3934.8929	
80	φ (age + year) P (sex)	28	3878.3727	0	
81	φ (age + DPop) P (sex)	5	3969.7184	44.8446	0.2844
82	φ (age) P (sex)	4	3974.4905	47.6103	
	E-Pirio			AICc(Model 83) = 2547.6513	
83	**φ (age + DPop) P (sex)**	**6**	**2535.6175**	**0**	**0.1220**
84	φ (age) P (sex)	5	2540.2929	2.6674	
85	φ (age + year) P (sex)	40	2470.4757	6.1751	
	D-Muro			AICc(Model 86) = 2461.4244	
86	φ (age) P (sex)	4	2453.4049	0	
87	φ (age + DPop) P (sex)	5	2453.1267	1.7316	0.6713
88	φ (age + year) P (sex)	26	2421.5842	12.8528	
	E-Muro			AICc(Model 89) = 1471.9337	
89	φ (age) P (sex)	4	1463.8945	0	
90	**φ (age + DPop) P (sex)**	**5**	**1462.0949**	**0.2201**	**0.0424**
91	φ (age + year) P (sex)	21	1456.1772	27.1656	

C.

Figure 4: Effect of population density in year t on annual adult survival from t to t+1, in 4 populations of blue tits. A: Population density across the years of monitoring in the different populations. B: Model selection in each population separately to test for an effect of population density on adult survival. Models 83 (in E-Pirio) and 90 (in E-Muro) are in bold as population density either significantly reduces the AICc or p-ANODEV<0.05. C: Annual adult survival as a function of population density in E-Pirio (orange) and E-Muro (red). Single points represent yearly estimates from Models 85 and 91 (see B) with 95% confidence intervals, lines represent the adult survival – population density relation as estimated by Models 83 and 90 (see B) with 95% confidence intervals in dotted lines.

D-Rouviere — AICc(Model 80) = 3934.8929

Model Number	Model Description	Number of Parameters	Deviance	ΔAICc	p-ANODEV
80	φ (age + year) P (sex)	28	3878.3727	0	
92	**φ (age + Linear_time) P (sex)**	**5**	**3965.4145**	**40.5407**	**0.1351**
93	φ (age + SpringRF) P (sex)	5	3969.1937	44.3199	0.2587
81	φ (age + DPop) P (sex)	5	3969.7184	44.8446	0.2844
94	φ (age + EHE) P (sex)	5	3971.6614	46.7876	0.4122
95	φ (age + AutumnRF) P (sex)	5	3972.4702	47.5964	0.4893
82	φ (age) P (sex)	4	3974.4905	47.6103	
96	φ (age + HST) P (sex)	5	3973.0642	48.1903	0.5619

D-Muro — AICc(Model 102) = 2460.1337

Model Number	Model Description	Number of Parameters	Deviance	ΔAICc	p-ANODEV
102	φ (age + AutumnRF) P (sex)	5	2450.1043	0	0.1340
86	φ (age) P (sex)	4	2453.4049	1.2907	
103	φ (age + EHE) P (sex)	5	2451.7473	1.6429	0.2949
104	φ (age + Linear_time) P (sex)	5	2452.7143	2.6099	0.5024
87	φ (age + DPop) P (sex)	5	2453.1267	3.0223	
105	φ (age + HST) P (sex)	5	2453.2098	3.1054	0.7225
106	φ (age + SpringRF) P (sex)	5	2453.4046	3.3002	0.9891
88	φ (age + year) P (sex)	26	2421.5842	14.1435	

E-Pirio — AICc(Model 83) = 2545.6416

Model Number	Model Description	Number of Parameters	Deviance	ΔAICc	p-ANODEV
83	**φ (age + DPop) P (sex)**	**6**	**2535.6175**	**0**	**0.1220**
84	φ (age) P (sex)	4	2540.2929	2.6658	
97	φ (age + EHE) P (sex)	5	2538.534	2.9166	0.3481
98	φ (age + SpringRF) P (sex)	5	2538.6945	3.0771	0.3714
99	φ (age + HST) P (sex)	5	2539.6612	4.0438	0.5755
100	φ (age + AutumnRF) P (sex)	5	2540.1464	4.529	0.7878
101	φ (age + Linear_time) P (sex)	5	2540.2255	4.608	0.8552
85	φ (age + year) P (sex)	40	2470.4757	6.1751	

E-Muro — AICc(Model 89) = 1471.9337

Model Number	Model Description	Number of Parameters	Deviance	ΔAICc	p-ANODEV
89	φ (age) P (sex)	4	1463.8945	0	
90	**φ (age + DPop) P (sex)**	**5**	**1462.0949**	**0.2201**	**0.0424**
107	**φ (age + Linear_time) P (sex)**	**5**	**1462.1533**	**0.2785**	**0.0464**
108	φ (age + EHE) P (sex)	5	1462.9217	1.047	0.1482
109	φ (age + HST) P (sex)	5	1463.0809	1.2061	0.1886
110	φ (age + SpringRF) P (sex)	5	1463.3806	1.5058	0.3012
111	φ (age + AutumnRF) P (sex)	5	1463.8804	2.0056	0.8663
91	φ (age + year) P (sex)	21	1456.1772	27.1656	

Table 4: Effect of local climatic variables on adult survival in the four focal populations of blue tits. Models in bold display either a covariate that explains a significant part of the temporal variation of survival (p-ANODEV<0.05, models 90 and 107), or a covariate that (1) reduces the AICc compared to the starting model (φ (age) P (s)) and (2) has a ß-estimate with both limits of its 95% confidence interval of the same sign (either both positive or both negative, models 92 and 83). Dpop: breeding population density index, SpringRF : spring and summer rainfall, AutumnRF : autumn and winter rainfall, EHE : summer extreme heat events, HST: hottest summer temperatures, Linear_time: linear effect of time; see methods section for more details on these covariates.

Discussion

The present study provides estimates for adult survival and recapture probabilities in four Mediterranean blue tit populations breeding in nest boxes, and explores factors that explain variation in adult survival across years. The results show that these populations have different average survival probabilities, but that the year-to-year variation in survival probabilities is similar among populations. This suggests that regional environmental conditions (e.g. climate) could drive survival fluctuations across populations. However, we did not find evidence that the atmospheric pressure conditions in the Mediterranean basin (approximated by the Mediterranean Oscillation Index MOI) nor the tropical climate index of rainfall in the Sahel were correlated to adult survival in these four blue tit populations. Estimates of breeding densities were not correlated across populations, but we found that in two of them (E-Pirio and E-Muro) breeding density had a significant negative impact on adult survival during the following year. Finally, two of the focal populations (D-Rouviere and E-Muro) displayed a linear temporal decrease in adult survival over the study period.

1. Recapture probabilities across sex and population

Model selection for the global dataset indicated that sex had an effect on recapture probabilities and that this effect varied among populations. Further results (see Appendix 2 and Table 2) indicated that the effect of sex was significant only in the E-Pirio population, where females had higher recapture probabilities than males. In two other populations, D-Rouviere and E-Muro, recapture estimates were higher for females as well, but female and male confidence intervals largely overlapped (Table 2). Higher recapture probabilities for females may be explained either because they have higher provisional rates than their male partners at the time when we capture parents (i.e. when nestlings are 10-15 days old, Banbura et al., 2001, Garcia-Navas et al., 2012, but see Limbourg et al., 2013, Iserbyt et al., 2019) or because males are more warry of human presence and traps used for nest box captures.

The fact that recapture probabilities are high in the D-Rouviere population (0.821 for males and 0.866 for females) is concordant with the scarcity of suitable cavities for breeding in the forest, as well as of suitable habitats nearby, since the forest is surrounded by vineyards. Conversely, the three Corsican sites are surrounded by adequate forest patches that offer possibilities for blue tit natal or breeding dispersal outside the nest box area. Other factors potentially explaining different recapture probabilities across populations include differences in bird boldness or brood failure.

2. Survival probabilities across sex and population

The estimated survival probabilities in our different populations were comparable, yet on the upper range, of those previously published for blue tits in the rest of the species distribution (see Table 5 for published estimates of blue tit adult survival).

Our results indicate that survival is higher in evergreen forest sites (E-Pirio and E-Muro) than in deciduous sites (D-Rouviere and D-Muro). One of reasons behind this difference between types of forest might be that the permanent leaves of the evergreen forest better protect birds from predators. This result adds to other facets of the phenotypic variation among these two habitats (morphology, behaviour, colour ornamentation, physiology and life history, Charmantier et al., 2016). Overall, the multivariate phenotypic differentiation documented between blue tits from deciduous and evergreen habitats is consistent with a divergence in pace-of-life syndrome (Réale et al., 2010) whereby individuals in the deciduous habitat display a faster pace-of-life (in particular larger clutches and higher aggression) compared to individuals in evergreen populations (Dubuc-Messier et al., 2017, Charmantier et al., 2016). Our adult survival estimates are hence in agreement with this expectation (Table 2). A common garden experiment revealed that differences in behaviour (exploration speed and handling aggression) as well as physiology (heart rate) between D-Muro and E-Pirio were maintained when nestlings from the two areas were raised in aviaries and kept for 5 years (Dubuc-Messier et al., 2018). This result suggests a genetic origin to the divergence in morphology and behaviour, yet survival in wild conditions could not be explored in this common garden experiment. The lower survival of blue tits in deciduous compared to evergreen habitats revealed in this study could be of multiple origins (e.g. difference in predator or parasite rates) and might not be of genetic origin.

Blue tit adult survival estimates	Method	Location	Reference
Males: 0.17 - 0.31 *Females:* 0.19 - 0.41	Not corrected by recapture probability: survival probably underestimated	Wytham Woods near Oxford, UK	(Nur, 1984b)
Males: 0.23 - 0.419 *Females:* 0.25 - 0.539	Not corrected by recapture probability: survival probably underestimated	Four study plots near Antwerp, Belgium	(Dhondt, 1987)
1st year adults: *Males:* 0.82 ± 0.07 *Females:* 0.65 ± 0.09 *2+ year adults:* *Males:* 0.68 ± 0.05 *Females:* 0.47 ± 0.05	Capture-mark-recapture (CMR) modelling	Ventoux forest, south of France	(Blondel et al., 1992)
1st year adults (both sexes): 0.65 ± 0.05 *2+ year adults (both sexes):* 0.56 ± 0.03	CMR modelling with data from 1979-1989	E-Pirio, Corsica, France	(Blondel et al., 1992)
Males: 0.30 - 0.63 *Females:* 0.26 - 0.59	CMR modelling	Mixed boreal forest near Tammisaari, Finland	(Class et al., 2014)
Males (unfaithful/faithful): 0.31 / 0.51 *Females (unfaithful/faithful):* 0.26 / 0.48	CMR modelling	Wytham Woods near Oxford, UK	(Culina et al., 2015)

Table 5: A non-exhaustive list of estimates of blue tit annual adult survival previously published

The CMR models also reveal an important effect of age, with adults of 2 years or more having a lower

survival than 1st year breeders (Table 4). This result is in line with previous findings (see Table 5) and

might reveal actuarial senescence in blue tits. In a study of the closely related great tit over half a

century in Wytham Woods (UK), survival probabilities were shown to rapidly decrease with age after

two years (see Figure 1A in Bouwhuis et al., 2012). If age-specific survival follows a similar pattern in

blue tits, aggregating survival estimates of all adults of 2 years or more would result in lower average

survival for birds of 2 years or more than for one-year old birds.

3. Population density and survival

Breeding density was highly variable within populations with up to 30% increase or decrease between years in all populations (Figure 4A). A negative effect of population density on subsequent adult survival was expected as a consequence of the competition induced by high breeding density (Fay et al., 2015, Le Coeur et al., 2016), however it was only detected in the two evergreen populations (E-Pirio and E-Muro, Figure 4B&C).

Comparing breeding density across populations of blue tits requires great caution. The density indices all describe year-to-year variation of nest box occupation in a portion of each focal population where nest-box density is similar and stable over time. However, these occupations are most probably not driven by the same factors across populations, and in particular might be differentially related to intra- and inter-specific competition. As the number of nest boxes, their relative position, and the number of natural cavities vary between populations, a direct comparison is not easy to interpret. This limitation is a shortcoming of studying survival and recapture probabilities in a nest-box population. The type of data we used could not have been obtained without nest boxes, since they allow easy nest localisation, easy access to nestlings, and facilitate the capture and recapture of parents. However, nest boxes also influence parent survival (e.g. by altering predation and parasitism rates (Burke et al., 2004), and by allowing for higher breeding densities).

4. Temporal variation in survival and the role of climate

Our results confirm previous findings based on shorter time scales (*i.e.* 10 annual estimations in E-Pirio in Blondel et al. (1992) and 6 to 14 annual estimates in Grosbois et al. (2006) for D-Muro, D-Rouviere and E-Pirio) whereby blue tit survival varies substantially across time and across populations. Exploring the drivers of temporal fluctuations however, gave quite contrasting results compared to previous findings. Indeed, contrarily to Grosbois et al. (2006), our longer-term analysis did not reveal any correlation between parental survival and large-scale climate indices, nor between survival and local meteorological variables. This major difference may be a consequence of several variations between the two studies: in particular Grosbois et al. worked on one less population (E-Muro) and fewer years of monitoring (they focused on the 1985-2000 period) with a different data filtering (they did not right-censor adults that had undergone fitness-altering experiments such as brood-size manipulations). Moreover, their model selection resulted in recapture probabilities differently constrained: contrarily to us, they allowed annual variations of recapture probabilities, which may have consequences on the estimated variation in survival. In order to understand the

reasons for the discrepancy between our results and those of Grosbois et al. (2006), we conducted

some additional analysis on a restricted dataset (only including the populations and periods studied

by Grosbois et al., see Table S3 in Appendix 3). The results indicated that as in the study by Grosbois

et al., and contrary to our analysis of the full dataset, variation in survival was correlated with large-

scale climatic variation. However, in our case, it was essentially the Mediterranean Oscillation Index

(calculated over the summer period) that was correlated with adult tit survival, and not the Early

Summer Sahel Rainfall. This suggests that the discrepancy in the results obtained between our study

and that of Grosbois et al. (2006) concerning the link between survival and large-scale climatic

variations is largely due to the dataset considered : our total sample covers 5499 life histories over

periods of 19 to 38 years, while the reduced sample covers 1566 life histories over periods of 8 to 16

years. However, beyond the spatial and temporal sampling, other differences in approaches may

explain the divergent conclusions, particularly in the modeling and in the treatment of individuals

who have undergone experiments during their lifetime.

We found that two populations (D-Rouviere and E-Muro) display a declining linear temporal trend in

adult survival across 26 and 19 years of study respectively. Our analysis on small and large-scale

climate variation suggests that this decline is not linked to direct (temperature, rainfall)

meteorological causes, even though a significant warming has been reported in our study areas over

the last three decades (Warming of 0.61-0.66 °C per decade in spring across the different

populations, Bonamour et al., 2019). It might however be caused by more indirect effects that were

difficult to capture statistically, such as an influence of food availability (warming spring

temperatures may cause a phenological mismatch between the blue tits and their prey, Visser &

Gienapp, 2019, Visser et al., 1998), or of earlier blue tit phenology. In E-Pirio springs of early breeding

are followed by a year of low adult survival (Bastianelli et al., submitted work) hence it would be

interesting to extend the meteorological analysis to conditions in spring at year t-1 influencing

survival between year t and t+1. Overall, the present study does not provide evidence that the recent

rapid climate change influences blue tit adult survival.

Although local and regional climate conditions were not identified as drivers in the temporal

fluctuations in survival, an interesting result was that survival probabilities covaried across time in

the 4 study populations (Model 1 in Table 1), consistently with Grosbois et al. (2006), even though

temporal variability was of higher magnitude in D-Rouviere (Figure 2 and Figure 3B). The estimation

of the inter-annual variance in survival depends on both sampling and process variances, which

magnitudes are influenced by several parameters (e.g., sample size and proximity of mean survival to

the 0 and 1 bounds). It is therefore complicated to formally compare the inter-annual variances

obtained here with those from previous work. Nevertheless, year-to-year variation in survival of the

order of 20-30% as observed here (Figure 2), is consistent with other studies in small passerines (see

e.g. Siriwardena et al., 1999, Hõrak & Lebreton, 1998, Perdeck et al., 2000).

If climate is not the main driver in the variation of parental survival from one spring to the next, the

three main mortality causes that come to mind are resource (in particular insect) abundance,

predation and parasitism. Our long-term study does not provide data that would allow to test

whether fluctuations in these three environmental features explain synchronous survival fluctuations

across the four focal populations. We hope that this comparison across four populations that are

relatively close (south of France) will inspire further comparisons at a larger scale (across the species

distribution), including perhaps datasets where more biotic and abiotic environmental features have

been monitored over time.

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Data accessibility: Two data files (CMR histories and covariates) and all models run in E-SURGE are

shared as supplemental files in the bioRxiv submission of the manuscript.

Table 1: General model selection for models on pooled capture-recapture histories of the four focal blue tit populations, to test for the importance of age, sex,

population (pop) and annual variation (year) on survival (φ) and recapture (P) probabilities. When a first order interaction between two variables is indicated,

both non-interactive terms are present in the model even if not explicitly mentioned (e.g. Model 23: φ (pop.year) P (sex.pop) is equivalent to φ (pop + year +

pop.year) P (sex + pop + sex.pop). Best model is Model 1 with AICc = 10421.3358.

Model number	Model Description φ	P	Number of Parameters	Deviance	ΔAICc
1	φ (age + pop + year)	P (sex.pop)	50	10320.747	0
2	φ (sex + age + pop + year)	"	51	10320.645	1.9211
3	φ (sex.age + pop + year)	"	52	10319.664	2.9648
4	φ (age.pop + year)	"	53	10319.735	5.0604
5	φ (sex.pop + age + year)	"	54	10318.756	6.1061
6	φ (age.pop + sex + year)	"	54	10319.581	6.931
7	φ (sex.age + sex.pop + year)	"	55	10318.06	7.4357
8	φ (sex.age + age.pop + year)	"	55	10318.494	7.8697
9	φ (pop + year)	"	49	10333.044	10.2735
10	φ (sex.pop + age.pop + year)	"	57	10317.734	11.1619
11	φ (sex + pop + year)	"	50	10333.036	12.289
13	φ (sex.pop + year)	"	53	10331.4	16.7247
18	φ (age + pop.year)	"	112	10227.817	33.424
19	φ (age + pop)	"	14	10433.289	40.0017
20	φ (sex + age + pop)	"	15	10433.169	41.8879
21	φ (age.pop)	"	16	10432.083	42.8095
22	φ (sex.age + pop)	"	16	10432.468	43.1946

23	φ (pop.year)	"	111	10240.316	43.8704
24	φ (pop)	"	12	10441.723	44.4227
25	φ (age.pop)	"	17	10432.083	44.8173
26	φ (sex.pop + age)	"	18	10431.529	46.2721
27	**φ (pop)**	"	**13**	**10441.723**	**46.4286**
28	φ (age.pop + sex)	"	18	10431.906	46.6491
29	φ (sex.age + sex.pop)	"	19	10431.048	47.7993
30	φ (sex.age + age.pop)	"	19	10431.107	47.8589
32	φ (sex + pop)	"	14	10441.703	48.4157
33	φ (sex.pop)	"	16	10440.244	50.9703
34	φ (sex.pop + age.pop)	"	21	10430.317	51.0871
35	φ (sex.age + sex.pop + age.pop)	"	22	10429.717	52.498
36	φ (sex.pop)	"	17	10440.244	52.9781
37	φ (age + year)	"	47	10387.125	60.3094
40	φ (sex + age + year)	"	48	10387.09	62.2971
41	φ (sex.age + year)	"	49	10386.246	63.4753
44	**φ (year)**	"	**46**	**10396.344**	**67.5072**
45	φ (sex + year)	"	47	10396.344	69.5291
48	φ (age)	"	11	10501.131	101.8258
49	φ (sex + age)	"	12	10501.083	103.7827
50	φ (sex.age)	"	13	10500.657	105.3633
51	**φ (.)**	"	**10**	**10507.342**	**106.0316**
54	φ (sex)	"	11	10507.339	108.0339

12	**φ (sex.age + sex.pop + age.pop + year)**	**P (sex.pop)**	**58**	**10316.908**	**12.3625**
14	φ (sex.age + sex.pop + age.year)	"	89	10268.19	26.7113
15	φ (sex.age + age.pop + age.year)	"	89	10268.778	27.2995
16	φ (sex.pop + age.pop + age.year)	"	91	10267.741	30.347
17	**φ (sex.age + sex.pop + age.pop + age.year)**	**P (sex.pop)**	**92**	**10267.109**	**31.758**
31	φ (sex.age + sex.pop + age.pop + pop.year)	"	121	10224.033	48.1327
38	φ (sex.age + sex.pop + age.year + pop.year)	"	152	10172.66	60.756
39	φ (sex.age + age.pop + age.year + pop.year)	"	152	10172.665	60.7612
42	φ (sex.pop + age.pop + age.year + pop.year)	"	154	10171.829	64.0702
43	**φ (sex.age + sex.pop + age.pop + age.year + pop.year)**	**P (sex.pop)**	**155**	**10171.214**	**65.5283**
46	φ (sex.age + sex.pop + sex.year + age.pop + age.year)	"	128	10236.912	75.4222
47	φ (sex.age + sex.pop + sex.year + age.pop + pop.year)	"	157	10193.051	91.5132
52	φ (sex.age + sex.pop + sex.year + age.year + pop.year)	"	188	10143.481	106.4804
53	φ (sex.age + sex.year + age.pop + age.year + pop.year)	"	188	10144.029	107.0279
55	φ (sex.pop + sex.year + age.pop + age.year + pop.year)	"	190	10142.757	109.9362
56	**φ (sex.age + sex.pop + sex.year + age.pop + age.year + pop.year)**	**P (sex.pop)**	**191**	**10142.009**	**111.2785**
57	"	P (sex + pop)	188	10156.101	119.1006
58	"	P (sex.pop + year)	226	10084.811	127.5635
59	"	P (pop)	187	10168.876	129.7857
60	"	P (sex)	185	10177.07	133.8033
61	"	P (sex + year + pop)	223	10098.434	134.8637

62	"	P (.)	184	10189.131	143.7768
63	"	P (pop + year)	222	10111.116	145.4394
64	"	P (sex + year)	220	10119.319	149.4312
65	"	P (year)	219	10127.433	155.4405
66	"	P (sex.pop + sex.year)	260	10053	167.7178
67	"	P (sex.year + pop)	257	10059.606	167.9508
68	"	P (sex.pop + pop.year)	289	10000.485	177.045
69	"	P (sex.year)	254	10076.062	178.0382
70	"	P (pop.year + sex)	286	10011.914	182.0564
71	"	P (pop.year)	285	10021.253	189.2575
72	**φ (sex.age + sex.pop + sex.year + age.pop + age.year + pop.year)**	**P (sex.pop + sex.year + pop.year)**	**323**	**9962.8949**	**212.503**
73	"	P (sex.year + pop.year)	320	9971.2466	214.3854

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Appendix 1

	Intercept	Slope	adjusted r²
D-Muro	**1.85 ± 0.39**	**0.91 ± 0.02**	0.73
E-Muro	0.80 ± 0.46	**0.98 ± 0.02**	0.78
E-Pirio	-0.26 ±0.25	**0.96 ± 0.01**	0.84

730 **Table S1:** Correlation between local average daily temperature (minimum daily temperature +

731 maximum daily temperature / 2) from thermocron i-buttons placed in trees near nest boxes, in the

732 three Corsican sites (E-Pirio, E-Muro and D-Muro) and from measures of the national meteorological

733 station of Calvi from 2013 to 2016. Significant estimates (*p-value* < 0.05) are highlighted in bold.

Appendix 2

In order to test *a posteriori* the effect of sex on the probability of recapture (P) in the study

populations, we compared models with and without the effect of sex on P, considering for each

population the survival structure of the best model (model 80 for D-Rouviere, model 84 for E-Pirio,

model 86 for D-Muro, model 89 for E-Muro, see selection in Table 3 of the main manuscript).

The results presented below in Table S2 indicate that the effect of sex on the probability of recapture

is significant only for the E-Pirio population.

Model description	Number of parameters	Deviance	AICc	ΔAICc
D-Rouviere				
φ(age + year) P(sex)	28	3878.3727	3936.9302	0
φ(age + year) P(.)	27	3880.4642	3936.9844	0.0542
E-Pirio				
φ(age) P(sex)	4	2540.2929	2550.3171	0
φ(age) P(.)	3	2563.5723	2571.5884	21.2713
D-Muro				
φ(age) P(.)	3	2454.505	2462.5246	0
φ(age) P(sex)	4	2453.4049	2463.4342	0.9096
E-Muro				
φ(age) P(.)	3	1464.5992	1472.6385	0
φ(age) P(sex)	4	1463.8945	1473.9534	1.3149

746 **Table S2:** Comparison of models with and without a sex effect on recapture probabilities (P) in the

747 four study populations of blue tits. In each population, model structure for survival (φ) is based on

748 results from Table 3 of the main manuscript.

Appendix 3

In order to understand the reasons for the discrepancy between our results and those of Grosbois et al. (Grosbois et al., 2006), who analyzed blue tit survival in three of our four study populations over a more limited period of time (8 to 16 years, datasets running until 2000), we restricted our dataset to years and populations studied by Grosbois and collaborators: E-Pirio : 1985-2000, D-Muro : 1993-2000 and D-Rouviere : 1991-2000. We then reanalysed the effect of large-scale climatic variation on mean survival observed in the three populations (based on the additive *Year+Pop* survival model).

The removal of the 1979-1984 and 2001-2016 periods and the E-Muro population restricted the dataset in terms of numbers of individuals (see Table S3 below, where our sample sizes are compared with those of Grosbois et al. (2006)).

Study population	Nb. of individuals (restricted dataset, this study)	Nb. of individuals (Grosbois et al., 2006)
D-Rouviere	569	501
E-Pirio	689	692
D-Muro	308	253

Table S3. Sample sizes obtained with our restricted dataset and sample sizes used in Grobois et al. (2006).

To make our models comparable to those used by Grosbois et al. (2006), we used a general model φ (pop + year) P (sex + year.pop) and a reduced model φ (pop) P (sex + year.pop).

We tested the effects of three covariates (summerMOI, winterMOI and SRF) on survival variation through model comparison and ANODEV testing. Only linear effects of covariates were tested. The results are presented in Table S4 below (P-values do not include any correction for multi-testing).

Model description	Nb. parameters	Deviance	ΔAICc	p-ANODEV	R^2
φ(pop + winterMOI) P (sex+pop.year)	37	2374.53	0	0.023	0.284
φ(pop) P (sex+pop.year)	36	2379.73	3.141		
φ(pop + SRF) P (sex+pop.year)	37	2378.18	3.659	0.27	0.084
φ(pop + summerMOI) P (sex+pop.year)	37	2379.71	5.188	0.91	0.0009
φ(pop+year) P(sex+pop.year)	50	2361.40	13.848		
φ(pop.year) P(sex + pop.year)	64	2351.36	33.176		

Table S4: Model selection procedure and significance of the effects of three large-scale climatic indices on adult survival (φ) based on a restricted dataset (three populations over the period 1985-2000, see main text for details). Starting yearly dependant and constrained model for the ANODEV

are highlighted in bold. AICc of the best model (first raw of the table) is 2449.71. SRF = Early Summer

Sahel Rainfall; MOI= Mediterranean Oscillation Index. The values of R^2 provided correspond to the

proportion of temporal variance in survival explained by the model covariate, computed through

analysis of deviance.

The results indicate that as in Grosbois et al. (2006), and contrary to our analysis using a larger

dataset, variation in survival is correlated with large-scale climatic variations. However, in our case, it

was essentially the Mediterranean Oscillation Index (summerMOI, calculated over the summer

period) that was correlated with adult tit survival, and not the Early Summer Sahel Rainfall (SRF). The

discrepancy between these results is explained by the difference in datasets used.