1	Survival estimates strongly depend on capture-recapture designs in a
2	changing environment inducing dispersal
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24 Abstract:

25 Capture-Recapture (CR) approaches are extensively used to estimate demographic parameters. 26 Their robustness relies on the selection of suitable statistical models, but also on the sampling 27 design and effort deployed in the field. In colonial or territorial species faithful to their breeding 28 site, some recurrent local perturbations such as predation-induced breeding failure may lead 29 individuals to disperse locally, potentially outside the study area. This might induce 30 heterogeneity in CR histories and biases in demographic parameter estimates. Here, we assessed 31 the effects of buffers areas and multi-site designs on dataset homogeneity and survival estimate 32 accuracy. First, from a local multi-site long-term monitoring survey carried out in a colony of 33 black-legged kittiwakes, we tested the homogeneity of individual CR histories and compared 34 survival estimates from three datasets including one or several cliffs with or without buffer 35 areas. Then, using simulated data with or without a multi-site design and buffer areas, we 36 compared survival estimates for different scenarios involving contrasted local and regional 37 dispersal. For field-based CR data, implementing a multi-site design and adding buffer areas 38 did not improve dataset homogeneity, as all three datasets suffered from strong trap-dependence 39 possibly due to individual breeding success conditioning site fidelity. Nevertheless, it 40 significantly improved the accuracy of annual survival estimates. Results from simulations 41 confirmed that a multi-site survey provided estimates that matched true survival probabilities 42 with reduced confidence intervals while a single-site design consistently led to underestimated 43 survival probabilities or greater confidence intervals. Similarly, a multi-site survey allowed 44 detecting significant temporal trends while a single-site survey did not. Adapting the spatial 45 scale of sampling design to the local environment and species behaviour is essential to robustly

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- 46 provide parameters of key relevance for the monitoring and management of populations.
- 47 Implementing buffer areas or multi-site design may be especially beneficial for long-lived

48 species facing regular local perturbation events.

- 49
- 50 Keywords: Demography, environmental change, individual behaviour, population dynamics,

51 prospecting movements, transience, trap-dependence.

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4

53 Introduction

54 Monitoring and understanding how natural populations vary in space and time is a 55 crucial issue in ecology and conservation. Demographic parameters such as annual adult 56 survival rate, breeding probabilities or emigration rates, strongly influence population 57 trajectories but are challenging to estimate directly in the field. Capture-recapture (CR) 58 approaches have been developed to analyze incomplete data of individual life histories obtained 59 from the monitoring of marked individuals (Lebreton et al. 1992, Williams et al. 2002). They 60 are based on two crucial steps: the implementation of a study design in the field to mark and recapture/resight individuals and the application of statistical estimation models (Kendall et al. 61 62 2009).

63 One difficulty with CR data is to properly deal with the imperfect detection of marked individuals in the field, which can bias demographic parameters and reduce their accuracy, 64 65 especially if there is strong heterogeneity in individual detection probabilities (Archaux et al. 2012, Fletcher et al. 2012, Tyre et al. 2003). Variation in detection probabilities can be induced 66 67 by intrinsic factors associated with individual characteristics and behaviour (age, sex, breeding or social status, etc...), but also by extrinsic factors induced by the implementation of the 68 sampling design itself (Crespin et al. 2008, Sanz-Aguilar et al. 2010). In particular, the spatial 69 scale at which the sampling design is implemented may play a role in inducing heterogeneity 70 71 in individual capture histories because dynamic environments or recurrent perturbation events 72 acting at local scales may affect individual behaviour and movements. For example, predation-73 induced breeding failure in colonial or territorial species can lead individuals to disperse over a 74 few hundred meters (e.g., Catlin et al. 2005, Danchin et al. 1998, Danchin and Monnat 1992,

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Forero et al. 1999, Öst et al. 2011), potentially outside the monitored study area. These 75 76 movements may induce heterogeneity in CR histories through temporary or permanent 77 emigration and may ultimately lead to underestimated survival probabilities, as survival is 78 confounded with site fidelity and mortality, with dispersal (Cilimburg et al. 2002, Ergon and 79 Gardner 2014, Kendall et al. 2013, Marshall et al. 2004, Peñaloza et al. 2014, Sanz-Aguilar et 80 al. 2016, Taylor et al. 2015). The potential effect of environmental perturbations responsible 81 for local dispersal in long-term surveys is thus particularly important to consider, as it may directly affect the quality of CR datasets and thus the statistical methodology used and 82 83 subsequent results and inferences.

84 As adult annual survival primarily drives population dynamics and trajectories in long-85 lived species (Sæther and Bakke 2000, Stearns 1976), its estimation and accuracy are crucial, 86 especially when subsequently used in population dynamics models (Fletcher et al. 2012, 87 Ponchon et al. 2015b). Numerous complex statistical methods have been developed to account 88 for different types of individual heterogeneities and improve the reliability and accuracy of 89 demographic parameter estimates. (e.g. Choquet et al. 2013, Gimenez and Choquet 2010, 90 Pledger and Phillpot 2008, Pledger et al. 2003).. They usually validate or relax general model 91 assumptions such as homogeneity of detection probabilities but suffer from issues related to 92 parameter identifiability. They also often multiply the number of estimated parameters which 93 may sometimes be difficult to interpret in terms of biological significance (Lindberg 2012, 94 Schaub et al. 2004). Likewise, robust-design models account for temporary emigration and 95 provide relatively unbiased estimates (Kendall et al. 1997). However, they require a substantial 96 field effort to resight individuals, because they rely on primary occasions between which the

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97 population is open to immigration and emigration, and secondary occasions, during which the 98 population is closed (Williams et al. 2002). Therefore, even though complex CR models allow 99 estimating survival probabilities closer to true survival with greater accuracy, their use is not 100 always appropriate or straightforward (Lindberg 2012). Moreover, sophisticated models often 101 require solid knowledge of the underlying mathematical processes, which may prevent non-102 specialist statisticians or conservation practitioners from easily using them.

103 Another way to mitigate individual heterogeneity is to implement a sampling design 104 adapted to the biology and behaviour of the monitored species, as outlined by Kendall et al. 105 (2009) for seabirds. The inclusion of buffer areas (BA), i.e. areas surrounding the monitored 106 plots (Kendall et al. 2009), has been suggested as an efficient way to avoid edge effects of local 107 dispersal movements (Cilimburg et al. 2002, Horton and Letcher 2008, Marshall et al. 2004). 108 Accordingly, implementing a sampling design with BA that would allow accounting for the 109 potential indirect effects of local environmental perturbation events such as predation should 110 be beneficial for reducing heterogeneity and improving the accuracy of demographic parameter 111 estimates. Using such an approach is nevertheless not necessarily obvious as monitoring 112 surveys are often conducted on single plots that are assumed to hold a sufficient number of 113 faithful individuals undergoing the same environmental conditions as the whole population.

In this context, using field-based and simulated data, we explored the effects of the spatial scale of sampling design on the heterogeneity of individual capture histories and the estimation and accuracy of adult survival estimates in a commonly surveyed and widely distributed abundant seabird species, the black-legged kittiwake *Rissa tridactyla* (Coulson 2011, Frederiksen et al. 2005a). This species is particularly suitable for monitoring surveys

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because breeding individuals can be readily detected once marked and many populations are
monitored across the species' range, notably to detect potential effects of changes in
environmental conditions on the temporal dynamics of adult survival probabilities (Frederiksen
et al. 2005b, Oro and Furness 2002, Reiertsen et al. 2014, Sandvik et al. 2005).

123 In the first part of our study, we used CR data from a long-term monitoring program 124 carried out on the island of Hornøya (Norway), where the kittiwake colony numbers thousands 125 of breeding pairs (Krasnov et al. 2007). We constituted three datasets corresponding to three 126 different spatial scales covered in the survey to compare survival estimates, associated 127 confidence intervals and detection probabilities obtained from simple CR models dealing with 128 individual heterogeneity. The first dataset, based on a single-site design, included a single cliff 129 of the kittiwake colony (Fig. 1), where heavy predation on eggs and/or chicks by ravens Corvus 130 corax, has been regularly recorded over the years (rarely on adults). The second and third 131 datasets, based on a local multi-site design respectively without and with BA, included the cliff 132 of the first dataset and a series of other nesting cliffs (Fig. 1). In a second part, using simulated 133 data from a single-site design with or without buffer areas and a three-site design (Fig. 2), we 134 tested the capacity of statistical models to match true survival for different survival and 135 dispersal values. We also tested the effects of simultaneous local and regional dispersal on the 136 estimates of survival and detection probabilities. Finally, we tested whether negative temporal 137 trends in survival probabilities could be detected within a single-site or a two-site design. We 138 hypothesized that by increasing the spatial scale of sampling, i.e. by adding BA and by 139 extending the number of monitored plots, we would better account for local individual 140 movements and thus be more likely to consider a locally-closed population system. As a result,

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141 we expected survival estimates closer to true survival and lower confidence intervals for the 142 multi-site datasets in the field-based and simulated data analyses (Cilimburg et al. 2002, Horton 143 and Letcher 2008, Marshall et al. 2004). Moreover, we expected the bias in survival estimates 144 to be more important for high dispersal rates. Finally, we expected a better detection of temporal 145 trends for a multi-site design compared to a single-site one in the simulated data analysis 146 (Zimmerman et al. 2007).

147

148 Materials and methods

149 1) FIELD-BASED DATA

150 STUDY SITE AND SAMPLING DESIGN

151 The study colony is situated on the island of Hornøya (70°23'N, 31°09'E), Eastern 152 Finnmark, Norway, where more than 10 000 pairs of kittiwakes nest each year (Krasnov et al. 153 2007). The long-term monitoring survey started in 1998, with the capture of breeding adults 154 which were marked with individual colour ring combinations. Each year, 18 plots, divided into 155 three sectors and situated in different parts of the island (Fig. 1), have been monitored once 156 every three days from late April, before kittiwakes start building their nest, until late July, at 157 fledging. For each ringed individual resighted, the (i) plot name and nest site number, (ii) presence of ringed or non-ringed mate and (iii) individual activity (i.e. standing on a 158 159 constructed/non constructed nest, squatting, incubating, brooding chicks...) have been 160 recorded. Additional resighting efforts have been made outside the main monitored plots, at the 161 beginning and end of the season, when detection probabilities can be high (Chambert et al.

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2012). This notably allowed the resignting of birds in BA, i.e. around the plots, and in denseparts of the colony (Fig. 1).

164

165 DATA SELECTION

Because experiments involving the manipulation of the breeding success in 1998 and 166 167 1999 were conducted in most of the monitored plots (Boulinier et al. 2008), we focused the analyses on the period from 2003 to 2012. Individuals ringed before 2003 were included only 168 169 if they were resignted at least once as breeders within the monitored plots between 2003 and 170 2011. Moreover, as individuals can be resignted in different locations during the same season, 171 the nest site assigned for a given season corresponded to the site where individuals reproduced 172 or where they were most often seen during the season. Although each plot was visited ~30 times 173 each season, we summarized observations as single yearly occasions to avoid additional intra-174 season temporal heterogeneity in individual detection probabilities potentially related to local 175 breeding performances (Chambert et al. 2012). This means that resighted and non-resighted 176 individuals were respectively coded 1 and 0 for a given year. 177 To constitute the single-site dataset, we chose a cliff where predation on eggs and/or 178 chicks by ravens Corvus corax had been recorded over years. Predation pressure was 179 nevertheless irregular over years, with years of high predation events (i.e. Ponchon et al. 2017b) 180 and other with very few. We only selected birds that had been ringed in this cliff and thereafter 181 we considered only resightings in this cliff. All other observations made outside the cliff were

- 182 thus coded 0, as if the birds had not been seen for that year. Overall, the single- and the multi-
- 183 site datasets included respectively 92 and 779 marked individuals, representing a total of 319

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and 3306 observations over 10 yearly occasions. To constitute the multi-site dataset without
BA, all ringed birds assigned to BA, i.e. observed outside the monitored plots after the first
occasion, were coded 0. Observations in BA represented 9.1% of the total multi-site dataset.

187

188 GOODNESS-OF-FIT ANALYSIS

189 Analyses started with the Cormack-Jolly-Seber (CJS) model assuming time-dependent 190 survival and detection probabilities (Lebreton et al. 1992). This model allows estimating two 191 parameters: the annual apparent survival probability s, i.e, the probability that an individual is 192 alive and available for detection in the monitoring area and detection probability p, i.e. the 193 probability of resighting an individual at the monitored site given that it is alive in the 194 monitoring area.. The inter-individual homogeneity assumption of the CJS model was assessed 195 for each dataset independently using the U-CARE software (Choquet et al. 2009a). With this 196 tool, the causes of some lack of fit can be identified. In particular, test 3.SR tests for the presence 197 of transient individuals by comparing survival probabilities between newly marked individuals 198 and individuals resighted at previous occasions. Test 2.CT examines trap-dependence effects 199 by comparing detection probabilities between individuals already seen and those not seen at the 200 previous occasion. Any significant test (p-value < 0.05) means that the homogeneity 201 assumption of the CJS model (i.e. equal survival and detection probabilities among individuals) 202 is violated.

203

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²⁰⁴ MODEL SELECTION

205

The program E-Surge (Choquet et al. 2009b) was used to build and fit the models. Model 206 selection was performed based on the Quasi-likelihood Akaikes' Information Criterion, 207 corrected for small sample size and overdispersion (QAICc, Burnham and Anderson 2002). The 208 model with the lowest QAICc was considered to be the best model (see Appendix 1 for all 209 models tested).

210

211 2) SIMULATED DATA

212 To highlight further potential biases induced by the spatial scale of the sample design 213 on demographic parameters and subsequent inferences, we conducted a second analysis based 214 on simulated data. We compared survival probabilities estimated from one virtual single-site 215 design where 100 individuals were ringed at first occasion and one two-site design where 50 216 individuals were ringed in each site at first occasion as well (Fig. 2). We did not implement 217 transience and trap-dependence effects in individual capture histories to avoid confounding 218 effects.

219 In a first scenario, we fixed a constant true annual survival probability s of either 0.5, 220 0.7 or 0.9, a local dispersal rate d of 0.1 or 0.3, and a constant detection probability p of 0.7. 221 Dispersing individuals could come back to their initial site. We ran simulations over 8 yearly 222 occasions and we compared the estimates of survival and detection probability obtained from a 223 CJS model and a multistate model with an unobservable state (hereafter multistate model; 224 Schaub et al. 2004) for the single-site design and a multisite model (Lebreton et al. 2003) for 225 the multi-site design.

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In a second scenario, we fixed a constant survival s of 0.9 and we tested the effect of different values for local dispersal L towards buffer areas and regional dispersal r to a second site over 8 yearly occasions. We also tested whether a temporary increase of local dispersal (0.05 from occasion 1 to 3, 0.2 for occasion 4 and 5 and 0.05 for occasion 6 to 8) could bias the estimates of survival and detection probability with the same statistical models used in scenario 1.

In a third scenario, we implemented a fixed dispersal rate d of 0.3 but a time-dependent survival probability, decreasing from 0.9 to 0.75 over 10 yearly occasions. In this case, we tested whether CJS and multistate models were able to detect this negative trend by estimating a time-dependent survival probability depending on study design.

To avoid multiple Monte-Carlo simulations, we built a single three-sitedataset for each combination of dispersal and survival values using a numerical approach based on expected values (see Devineau et al. 2006 for procedure details) using M-SURGE software (Choquet et al. 2004).

- 240
- 241 **Results**
- 242 1) Field-based data

Goodness-of-fit tests did not reveal any transience effect, but a very strong trapdependence effect in the three datasets (Table 1). The multi-site dataset without BA was the one with the strongest over-dispersion, as the two tests of trap-dependence (Test 2.CT and 2.CL) were significant. The multi-site dataset including BA and the single-site dataset were more homogeneous, with only one significant trap-dependent test (Test 2.CT; Table 1).

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Consequently, we applied a trap-dependent model (Pradel 1993) for the three datasets and we additionally applied a multistate model (Schaub et al. 2004) on the single-site dataset. This model accounts for temporary emigration and allows individuals to enter an unobservable state when they temporarily or permanently move outside the study area. Additional lack of fit was taken into account by adding corresponding overdispersion coefficients \hat{c} calculated as the sum of chi-squares χ^2 divided by the sum of degrees of freedom *df* (Table 1).

The model selections gave a constant survival for the three datasets (Table 2), but the ones calculated from the single-site was 0.01 to 0.05 lower compared to the ones from the multi-site datasets that were higher, very similar and relatively accurate (0.876 ± 0.011 for the multisite design with BA and 0.872 ± 0.012 for the multisite design without BA; Table 3). The survival estimate obtained with the multistate model was 0.04 higher than the one obtained with the trap dependent model for the single-site dataset but the confidence intervals was as large (Table 3).

261 Detection probabilities were constant for the single-site dataset, both with the trap-262 dependent and multistate model. On the contrary, they were time-dependent for the two other 263 datasets but values were relatively similar (Fig. 3). Detection probabilities for individuals seen 264 the previous occasion were very high, between 0.80 and 0.93 for the two multi-site datasets and 265 fixed at 0.909 ± 0.031 and for the single-site dataset. Those for individuals not seen the previous 266 occasion were consistently lower, fluctuating between 0.18 to 0.58 for the two multi-site 267 datasets and constant at 0.330 ± 0.159 for the single-site dataset with the trap-dependent model. 268 The addition of buffer areas seemed to improve the detection probabilities for individuals not 269 seen the previous occasion (Fig. 3).

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270

271 2) Simulated data

The results from the simulated data showed that as expected, the type of design and model choice strongly influenced both the estimates of annual survival and detection probabilities (Fig. 4).

275 In a first scenario testing different combinations of annual survival and dispersal rates 276 (Fig. 4, left panel), the design including buffer areas led to survival values that matched the true parameter with reduced confidence intervals. The one obtained from the trap-dependent model 277 278 for the single-site design led to constantly lower and less accurate survival values and he bias 279 was exacerbated for high survival and dispersal rates. The multistate model gave survival 280 estimates close to true survival but with large confidence intervals and performed worse when 281 survival and dispersal rates were high. The same pattern was observed for detection 282 probabilities (Fig. 4).

283 In the second scenario (Fig. 4, right panel), the annual survival probabilities were only 284 correctly estimated within a three-sites design or a site including buffer areas when local 285 dispersal L was high (0.3). When local dispersal L was low (0.05), regardless of regional 286 dispersal r, survival probability was 0.03 to 0.15 lower than the true parameter and confidence 287 intervals were larger. Time-dependent local dispersal rates led to slightly lower survival estimates compared to constant ones. It has to be noted that for both high local and regional 288 289 dispersal rates (L0.3/r0.2), the model overestimated survival for the single-site design and 290 strongly underestimated detection probabilities. In case of high local and/or regional dispersal, 291 the single-site design led to strongly biased detection probabilities. The multistate model 292 applied on the single-site dataset gave similar survival estimates as the multisite model applied Ponchon, Aurore; Choquet, Rémi; Tornos, Jéremy; McCoy, Karen D.; Tveraa, Torkild; Boulinier, Thierry. Survival estimates strongly depend on capture-recapture designs in a disturbed environment inducing dispersal. Ecography 2018 ;Volum 41.

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on the two-site design, except when local dispersal was high (L= 0.3). However, estimates had
larger confidence intervals.

295 In the last scenario, which implemented a decrease of annual survival probability from 296 0.9 to 0.75, two models were equivalent for the single-site, both with the CJS and multistate 297 model, as the difference between their QAICc was lower than 2 (Table 4). The best-ranked CJS 298 and multistate model gave a constant survival probability of 0.72 and 0.77 respectively. The 299 second best models gave a time-dependent survival probability (Fig. 5), which temporal trends 300 were not significant, as the confidence interval included 0 (95% confidence interval for the CJS 301 model: -0.16; 0.032; 95% confidence interval for the multistate model: -0.20; 0.04). In the two-302 site design, the survival probability obtained from the best-ranked model was time-dependent 303 and the associated negative trend was significant (95% confidence interval: -0.20; -0.033). 304 Overall, the two-site design led to higher and more accurate survival estimates, closer to true 305 survival compared to the single-site one, regardless of CR models, and allowed detecting 306 temporal decrease in annual survival probabilities.

307

308 Discussion

In this study, we combined field-based and simulated data to test the effects of the spatial scale of sampling design on dataset homogeneity, estimation and accuracy of survival annual and subsequent inferences when individuals disperse locally in buffer areas or at larger scales. We found consistent differences between the results provided by the single-site dataset and the two other multi-site ones in the field based-study and these were confirmed by the simulation study.

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315

316 INDIVIDUAL HETEROGENEITY

317 In the field-based data analysis, all datasets suffered from a strong trap-dependence 318 effect (Table 1). This heterogeneity in detection probabilities may result from the species 319 behaviour and movements related to breeding site selection. Indeed, during the breeding season, 320 kittiwake behaviour may be influenced by individual and conspecific breeding performance 321 (Ponchon et al. 2015a, Ponchon et al. 2017b), which may itself conditions site fidelity the 322 following year (Boulinier et al. 2008, Danchin et al. 1998, Naves et al. 2006, Ponchon et al. 2015a, Ponchon et al. 2017b). If individuals successfully breed within a monitored plot, they 323 324 will be more likely to remain faithful to their breeding site and thus their probability to be 325 resighted the following year may be higher. On the contrary, failed breeding individuals are 326 more likely to disperse to a new breeding site, especially when their neighbours are also in 327 failure, which might decrease their detection probabilities (Boulinier et al. 2008, Danchin et al. 328 1998). Moreover, black-legged kittiwakes frequently display prospecting movements to other 329 breeding areas when they are failed or non-breeders (Boulinier et al. 1996, Cadiou et al. 1994, 330 Danchin et al. 1998, Ponchon et al. 2017a, Ponchon et al. 2013). This means that even if 331 individuals have dispersed out of the monitored area, they may occasionally come back to the monitored area for prospecting, notably if they have failed breeding. This might induce 332 333 temporal heterogeneity in detection probabilities, intra-seasonally (Chambert et al. 2012) but 334 also possibly inter-annually, as kittiwake breeding success can be highly variable in time and 335 space (Paredes et al. 2014, Ponchon et al. 2014). Therefore, in the current case study,

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heterogeneity may originate from temporary emigration but also from individual site fidelityconditioned by individual and conspecific breeding performance,

338

339 ACCURACY AND PRECISION OF SURVIVAL ESTIMATES

340 In the field-based study, the constant survival probability obtained from the single-site 341 dataset was underestimated or less precise compared to the ones estimated from the multi-site 342 datasets (Table 3). The cliff selected to constitute the single-site dataset has been irregularly 343 impacted by predation on eggs and/or chicks (rarely adults) by ravens Corvus corax. This has 344 resulted in regular massive local breeding failures and a progressive desertion of the area, likely 345 involving adult dispersal towards other breeding cliffs (Boulinier et al. 2008, Danchin et al. 346 1998, Ponchon et al. 2017b). In the case of the single-site survey design, the dispersing failed 347 birds may not be detected anymore and may be considered as dead while they are still alive but 348 somewhere else, which may directly decrease apparent survival (Cilimburg et al. 2002, Ergon 349 and Gardner 2014, Horton and Letcher 2008, Marshall et al. 2004, Sanz-Aguilar et al. 2016, 350 Zimmerman et al. 2007). On the contrary, in the multi-site survey, although some monitored 351 cliffs have also been affected by predation, these local effects may have been buffered by the 352 spatial scale considered. Individuals may have potentially dispersed to other monitored cliffs 353 where they could still be detected, which resulted in estimations of higher annual survival 354 probabilities, probably closer to the true survival probability. This interpretation is strengthened 355 by the simulation results, which clearly show that not accounting for local dispersal by 356 monitoring a single site may lead to a strong underestimation of annual survival probability, 357 especially for long-lived species (Fig. 4). Likewise, scenario 3 from the simulation study

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revealed that a multi-site design was essential to detect a slow but significant decrease of annual
survival probability over time (Fig. 5).

Overall, the use of a multistate model accounting for temporary emigration within a 360 361 single-site design provided values close to true survival, as the ones obtained within a multisite design but had larger confidence intervals (Fig.4). Moreover, some models suffered from 362 363 identifiability issues (Appendix 1), leading to less robust results than the ones obtained within 364 a multisite design. A robust design (Kendall et al. 1997) may not have helped obtain better 365 estimated because populations are not closed within secondary occasions due prospecting 366 movements which unevenly occur during the breeding season (Boulinier et al. 1996, Chambert 367 et al. 2012, Ponchon et al. 2015a). Therefore, our resultsemphasize the necessity to match 368 species biology by implementing the right design before developing complicated and 369 potentially ineffective CR models ..

370

371 BENEFITS AND IMPLICATIONS OF BUFFER AREAS OR MULTI-SITE STUDY DESIGNS

372 The survival estimate for the multi-site dataset including BA is only slightly higher 373 (0.876) than the one without BA (0.872), and the confidence interval, slightly lower 374 (respectively 0.854-0.896 and 0.846-0.894). Nevertheless, the dataset with BA is more 375 homogenous, as only one trap-dependence effect is detected (Table 1). Thus, in our field-based 376 CR study, although 9% of observations occurred in BA, the inclusion of these data did not 377 markedly help improve survival estimates and confidence intervals, but efficiently improve 378 dataset homogeneity and detection probabilities for birds not resighted on the previous 379 occasions (Fig. 3). This highlights that in our field study, a multi-site design is more efficient

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than just locally extending one monitored area to estimate annual survival probabilities. Nevertheless, simulations provide contrasted results. Scenario 2, notably testing the effects local and regional dispersal, shows that in case of high local dispersal (L=0.3), a buffer area is sufficient to obtain relatively unbiased survival estimates, even when regional dispersal is high (Fig. 4).

385 In addition to estimating survival probabilities closer to true survival and allowing more 386 robust ecological inferences thanks to reduced confidence intervals, a multi-site sampling 387 design may provide wider perspectives in terms of species management and conservation, but 388 also population ecology. It may allow a better understanding of local movements and thus local 389 population dynamics, by the possibility of estimating dispersal rates using more sophisticated 390 statistical models, such as multi-site (Lebreton et al. 2003), multi-event (Pradel 2005) or 391 spatially explicit CR models accounting for locations associated with individual detection 392 (Borchers and Fewster 2016). If such design could further be implemented in several 393 populations, this would lead to a hierarchical design enhancing efficient conservation and 394 management efforts both at a local and regional scale. This could provide information on the 395 variation of survival estimates (Frederiksen et al. 2005a), revealing specific local demographic 396 and dispersal rates but also regional connectivity (Barlow et al. 2013, Fernández-Chacón et al. 397 2013). More importantly, a multi-site design may anticipate changes in the local environment, 398 which may avoid ending with a completely empty monitored area after years of bad local 399 environmental conditions. This kind of design would therefore be beneficial for species more 400 likely to disperse to other surrounding breeding areas because of recurrent local perturbations 401 such as predation or man-induced disturbance.

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402 Implementing the appropriate sampling design at the beginning of a long-term study 403 primarily depends on the purpose of long-term monitoring surveys, but it also depends on the 404 time to be spent in the field to collect data, sample size, potential individual heterogeneity and 405 the precision needed to make demographic inferences (Field et al. 2005, Lahoz-Monfort et al. 406 2014, Lindberg 2012, Nichols and Williams 2006). Enlarging the monitored area often requires 407 more time spent in the field and may be constrained by logistical or financial issues. However, 408 the number of marked individuals and the proportion of population monitored should not be 409 neglected, because small sample sizes or too small monitored areas lead to underestimated 410 annual survival probabilities and larger confidence intervals (Devineau et al. 2006, Sanz-Aguilar et al. 2016, Zimmerman et al. 2007), which would in the end, weaken ecological 411 412 inferences. The advantage of a multi-site design is that it may account for a larger proportion 413 of the area encompassing the population and it may not necessarily imply more individuals to 414 resight/capture, as the initial number of ringed individuals in one area can be scattered in several 415 ones (Fig. 2). Additionally, identifying the periods when individuals are the most detectable 416 may help optimize a multi-site design, reduce field effort and mitigate individual heterogeneity 417 (Chambert et al. 2012, Field et al. 2005, Lahoz-Monfort et al. 2014). Finally, the duration of the long-term surveys should also be carefully considered, as it may strongly bias survival 418 419 estimates towards the end of the time series (Langtimm 2009, Peñaloza et al. 2014).

420

421 CONCLUSION

In a context of current climate change, CR surveys are frequently used to study theeffects of environmental factors on the survival of potentially sensitive species, but their

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424 robustness is sometimes questioned (Grosbois et al. 2009). The accuracy and reliability of 425 survival estimates based on an appropriate study design and statistical models are crucial to 426 draw reliable demographic inferences, especially when estimated parameters are used to 427 correlate demography and climate variability at large spatial scales (Grosbois et al. 2008, 428 Tavecchia et al. 2016), predict population response to environmental change (Ponchon et al. 429 2015b) or make decisive conservation and management plans (Kendall et al. 2009, Williams et 430 al. 2002, Zimmerman et al. 2007). The lower survival probability obtained from our field-based 431 single-site dataset and the results from the simulation analysis highlight the need to implement 432 adapted study designs to closely match species biology and behaviour and avoid biases due to 433 local environmental factors or perturbation events that might alter individual behaviour and 434 detection at local and regional scales.

In addition to spatial scale, the timing of surveys could also affect heterogeneity through 435 436 strong temporal variation in detection probabilities of individuals (Chambert et al. 2012). An 437 increasing number of studies have shown that species may change their phenology to match 438 new environmental conditions driven by climate change (Walther et al. 2002), breeding earlier 439 (Moyes et al. 2011) or delaying their hibernation emergence (Lane et al. 2012). Consequently, 440 monitoring surveys should thus not only be implemented to match the species' spatial ecology, 441 as highlighted here, but should also possibly be fine-tuned to the local conditions, which may 442 vary at various temporal scales (Chambert et al. 2012, Hurley et al. 2013).

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458 Author contribution statement:

TB, KDM and TT implemented the long-term monitoring survey and associated design. TB, JT, KDM, TT and AP participated to fieldwork to collect the data. AP and JT processed the data. AP and RC ran capture-recapture models from field-based and simulated data. AP led the writing of the manuscript. All authors contributed critically to the drafts of the manuscript and gave final approval for publication.

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627 Significant chi-square statistics (p < 0.05) reflecting model assumption violations are in bold.

628

Single-site dataset			lataset	Multi-site dataset without BA			Multi-site dataset with BA		
Goodness of fit components	χ²	df	р	χ^2	df	р	χ^2	df	р
Test 3.SR	2.6	8	0.95	5.2	8	0.73	10.3	8	0.25
Test 3.SM	1.6	6	0.91	8.1	8	0.43	6.8	8	0.61
Test 2.CT	21.0	6	0.0019	421.9	7	<0.001	278.0	7	<0.001
Test 2.CL	0	4	1	17.5	7	0.014	10.7	6	0.10
\hat{c} with the CJS model		1.05			15.1			10.5	
<i>ĉ</i> for trap- dependent model		1			1.34			1.26	

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<u>Table 2</u>: Most parsimonious models selected based on QAICc for each of the three datasets.

s is the apparent survival probability, T is a transition probability, *p* is the detection probability,

i is a constant, *t* is time-dependence and *a* is a trap-dependant effect.

Model structure	Dataset	Model	Nb of parameters	Deviance	QAIC	QAICc
Multistate model with and unobservable state	Single-site dataset	s(i)T(i)p(i)	4	407.62	415.62	415.75
	Single-site dataset	s(i) p(a)	4	406.60	414.60	414.67
Trap-dependent model	Multi-site dataset without BA	s(i) p(a+t)	12	3925.18	2953.24	2953.30
	Multi-site dataset with BA	s(i) p(a+t)	12	3980.90	3183.45	3183.50

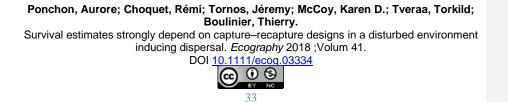
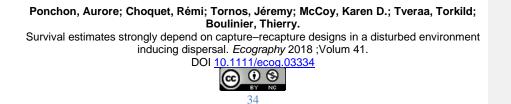


Table 3: Estimates of annual apparent survival probability for the three field-based datasets

637 with a trap-dependant (TD) or a multistate model with an unobservable state (UN).

Field design	Estimate ± SE	95% confidence interval
Single-site (TD)	0.823 ± 0.028	0.762 - 0.871
Single site (UN)	0.861 ± 0.024	0.806 - 0.902
Multi-site (TD)	0.872 ± 0.012	0.846 - 0.894
Multi-site + BA (TD)	0.876 ± 0.011	0.854 - 0.895



640	Table 4: Model selection for the third scenario of simulations, parameterized with a true
641	survival probability starting at 0.9 and decreasing of ca. 0.015 every year and detection
642	probability of 0.7. In the models, survival s , transition T and detection probabilities p are either
643	constant (i), time-dependent (t), or decreasing (slope). The best selected models are in bold. $*$
644	indicates models with unidentifiable parameters.

645

Design	Model	Number of parameters	Deviance	QAIC	QAICc
	s(i)T(i)p(i)	4	812.28	820.28	820.38
Single-site	s(slope)T(i)p(i)	5	810.69	820.69	820.84
(Multistate with an unobservable	s(t)T(i)p(i)	12	810.68	834.68	835.47
state)	s(slope)T(i)p(t)	13	810.68	836.68	837.60
	s(t)T(i)p(t)	19	810.68	848.68	850.63
Single-site	s(i), p(i)	2	828.06	832.06	832.09
C	s(slope), p(i)	3	826.38	832.38	832.44
(CJS)	s(slope), p(t)	11	825.44	847.44	848.11
	s(t), p(t)	17	825.27	859.27	860.84
—	s(slope), p(i)	3	1488.34	1488.34 1494.33 1494	1494.36
Two-sites	s(i), p(i)	2	1495.63	1499.63	1499.65
(CIS)	s(t), p(i)	10	1488.28	1508.28	1508.56
(CJS)	s(slope),p(t)	12	1488.31	1512.31	1512.71
	*s(t)p(t)	18	1488.28	1524.28	1525.16

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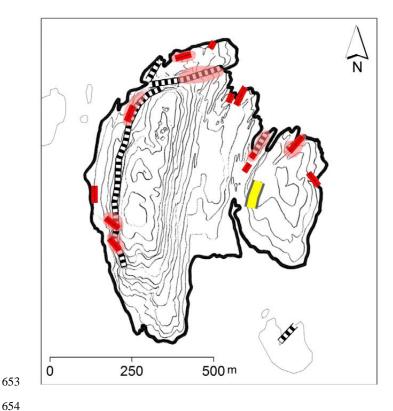
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Figures 648

- 649 Figure 1: Map of the different areas included in the single-site (thick yellow line) and multi-
- site (red lines) designs. All kittiwake nesting cliffs are represented by striped black and white 650
- 651 lines and buffer areas, by light red areas.
- 652



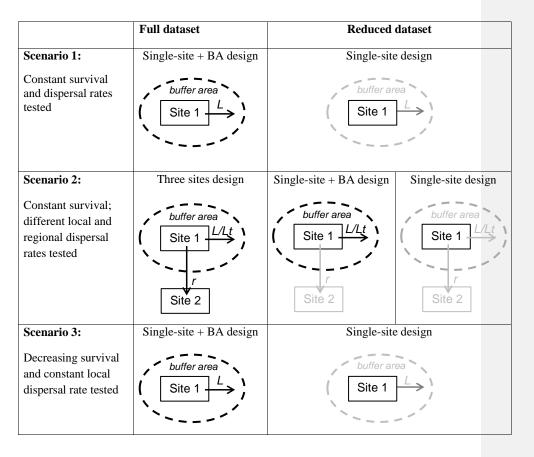


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Figure 2: Field designs tested with simulated datasets. In scenario 1, individuals disperse locally at rate L to buffer areas. They can be monitored (black lines) in buffer areas or not (grey lines) depending on study design. In scenario 2, individuals disperse locally to buffer areas at a constant (L) or time-dependent rate Lt and disperse regionally to another site with a constant rate r. Scenario 3 is similar to scenario 1 but true survival decreases over time.

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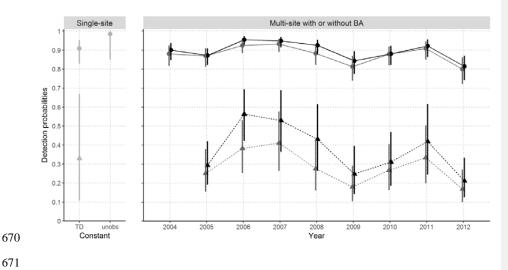


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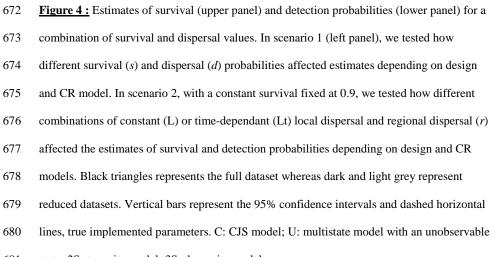


Figure 3: Detection probabilities and associated 95% confidence intervals estimated from the 662 663 best selected models for the single-site field-based dataset analyzed with a trap-dependent 664 model (TD) and a multistate model with an unobservable state (unobs) and the two multi-site 665 field-based datasets analyzed with a trap-dependent model. Dark grey represents the multi-site dataset without BA and black, the multi-site dataset with BA. Points and triangles represent 666 667 respectively the group of individuals seen the previous occasion and that of individuals not seen 668 the previous occasion. Note that the multistate model with an unobservable state only has one group of detection probabilities. 669

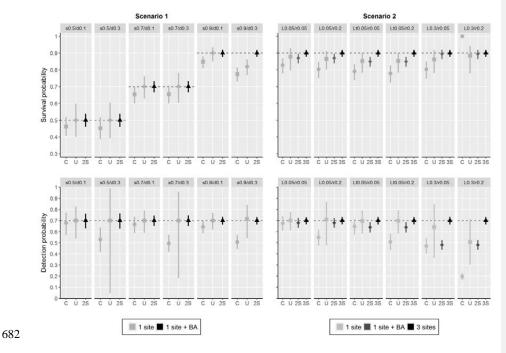


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state; 2S: two-site model; 3S: three-site model.



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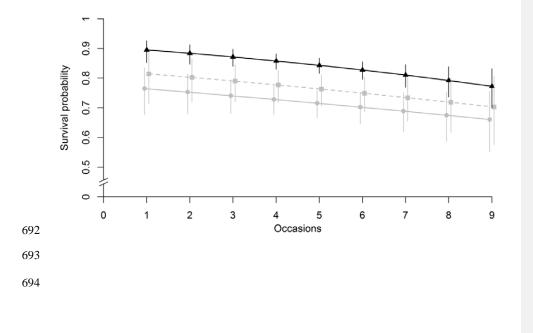
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Figure 5: Annual survival probabilities estimated from the best ranked-model for Scenario 3, implemented with a fixed dispersal rate of 0.3 and a time-dependent survival probability, decreasing from 0.9 to 0.75. Light grey points and squares respectively represent estimates obtained with a CJS and multistate model with a unobservable state applied on a single-site dataset. Black triangles represent estimates from a CJS model applied on a two-site design dataset. Vertical bars indicate 95% confidence intervals.



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695 Supporting information:

Model selection for the analysis of the field-based CR data. s is the annual apparent survival probability, T is a transition probability, p is the detection probability, i is a constant, t is timedependence and a is a trap-dependent effect. Selected models for each dataset are in bold. *

699 indicates models with unidentifiable parameters.

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Dataset and model applied	Model	Number of parameters	Deviance	QAIC	QAICc
	s(i)p(a+t)	12	3980.90	3183.44	3183.50
	s(i)p(a.t)	19	3964.10	3184.11	3184.24
Multi-site+BA	s(t)p(a+t)	19	3968.76	3187.81	3187.94
	s(t)p(a.t)	25	3956.04	3189.72	3189.94
Trap-dependent	s(t)p(a)	12	3999.98	3198.58	3198.64
model	s(i)p(a)	4	4033.04	3208.82	3208.84
	s(t)p(i)	11	4177.67	3337.61	3337.66
	s(i)p(i)	3	4208.40	3346.00	3346.00
	s(i)p(a+t)	12	3925.18	2953.24	2953.30
	s(i)p(a.t)	19	3908.96	2955.14	2955.28
Multi-site without	s(t)p(a)	12	3938.13	2962.90	2962.96
BA	s(t)p(a+t)	19	3920.83	2964.00	2964.13
Tran danandant	s(t)p(a.t)	25	3905.04	2964.21	2964.45
Trap-dependent model	s(i)p(a)	4	3962.56	2965.14	2965.14
mouer	s(i)p(i)	3	4236.99	3167.93	3167.93
	s(t)p(i)	11	4219.17	3170.63	3170.68
	s(i)p(a)	4	406.60	414.60	414.67
Single-site	s(i)p(a+t)	12	397.61	421.61	422.18
Single-site	s(t)p(a)	12	400.73	424.73	425.30
Trap-dependent	s(i)p(a.t)	19	391.85	429.85	431.25
model	s(t)p(a+t)	19	394.27	432.27	433.67
	s(i)p(i)	3	432.26	438.26	438.30
	s(t)p(a.t)	25	388.57	438.56	440.99
	s(i)T(i)p(i)	4	407.62	415.62	415.75
Single-site	s(i)T(i)p(t)	12	400.60	424.60	425.62
	s(t)T(i)p(i)	12	401.55	425.55	426.57
	s(t)T(t)p(i)	20	391.34	431.34	434.16

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Multistate model	*s(t)T(i)p(t)	19	394.66	432.66	435.20
with an	s(i)T(t)p(t)	20	392.78	432.78	435.60
unobservable state	s(t)T(t)p(t)	26	387.66	439.66	444.47
	s(i)T(t)p(i)	12	457.13	481.13	482.14

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