Temporal and spatial variability in availability bias has consequences for

marine bird abundance estimates during the non-breeding season

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Abstract

1. To effectively monitor how marine ecosystems are being reshaped by anthropogenic pressures, we require understanding of species abundances and distributions. Due to their socio-economic and ecological value, predatory species are often at the forefront of survey efforts. However, survey data are only valuable if they can reliably be converted into estimates of underlying distributions.

2. We consider at-sea surveys of marine predators that often inform ecological impact assessments of offshore windfarms. These surveys are subject to a form of detection bias called 'availability bias' whereby individuals which are submerged below the surface are consequently 'unavailable' for detection. Although correction factors are commonly used in these surveys, they are currently based on limited data that may not be species-, time-, or area-specific. Here, we use time-depth-recorder data to investigate variation in marine bird availability bias.

3. We found that the proportion of diving marine birds submerged below the sea surface during daylight hours, and therefore unavailable to be counted during surveys, varied by species, month, and area. For three of our focal species wintering around northwest Europe (Atlantic puffin, common guillemot, razorbill) our results were different to comparable values previously used to correct for the availability bias, whereas no correction factors are regularly used for the fourth species (red-throated diver). We now present availability bias correction factors that are species- and month-specific to the areas the study populations use during their non-breeding seasons: the North Sea, the north and west coasts of the UK, the Baltic Sea, and Icelandic coastal waters.

4. *Synthesis and applications:* Variation in the proportion of daylight hours that marine birds spent submerged lead to differences in availability bias correction factors, thereby impacting estimations of their abundances. We encourage use of correction factors that use data from the species, marine area, and month during which surveys are conducted to provide more accurate abundance estimates. Using more relevant correction factors will result in increasingly accurate abundance and distribution estimates of diving marine birds, with relevance for a range of applications including planning for offshore windfarm developments, the designation and monitoring of protected areas, and understanding environmental change.

KEYWORDS: *Alca torda;* Diving; Environmental impact assessment; *Fratercula arctica; Gavia stellata;* Offshore windfarm; Seabirds; *Uria aalge*

Introduction

Anthropogenic pressures are rapidly reshaping terrestrial, aquatic, and marine ecosystems (Delong et al., 2018; Henson et al., 2017). As we navigate the challenges of the Anthropocene, strive to monitor and manage changes to these different ecosystems effectively, and reduce further harm, an understanding of species abundances and distributions, and how these are changing, is essential (Boivin et al., 2016; Halpern et al., 2015). Examples of this include when researching and documenting the influence of environmental change on species distributions (Jetz et al., 2019), the consideration and monitoring of protected areas (Edgar et al., 2014), and the execution of environmental impact assessments (EIAs) for oil and gas licensing and offshore windfarms (Dierschke et al., 2016). However, if methods used to establish species abundances and distributions are flawed, then this is likely to lead to poor decision making and perhaps even exacerbate anthropogenic environmental problems (Guisan et al., 2013).

When devising management and conservation plans, apex predators tend to be highly regarded due to their social, economic, and ecological value (Sergio et al., 2006). Despite their significance, estimating the abundances and distributions of marine predators is challenging. This is an issue of high importance, since, in an effort to improve energy security and reduce human reliance on fossil fuels, there has been expansion in the development of marine renewable energy technologies that harness natural resources in the habitats occupied by these species (Soares-Ramos et al., 2020). As part of offshore windfarm EIAs, the at-sea abundances and distributions of marine apex predators are typically assessed using either visual observations, from ships or aircrafts, or digital aerial surveys (Buckland et al., 2012; Thaxter & Burton, 2009). Whilst ideally these surveys would capture and record the presence of all individuals, many taxa, including marine reptiles, mammals, and birds, spend extended periods below the water's surface during which they are undetectable by the methods used to survey them. Historically, this detection bias (sometimes referred to as 'visibility bias', but hereafter referred to as 'availability bias') has been corrected for in these taxa by following methods derived to estimate densities of harbour porpoises *Phocoena phocoena* (Barlow et al., 1988), despite the fact that marine birds can engage in additional relevant behaviours, most notably flight. Thus whilst the quantification of at-sea marine bird abundances and distributions is an important issue, it is also imperative that the approaches used to derive correction factors that account for availability bias are appropriate for this taxon (Certain & Bretagnolle, 2008; Winiarski et al., 2014).

Diving marine birds such as Atlantic puffins *Fratercula arctica* (hereafter 'puffins'), common guillemots *Uria aalge* (hereafter 'guillemots'), razorbills *Alca torda*, and red-throated divers *Gavia stellata* are potentially sensitive to industrial activities, including offshore windfarm developments, across their ranges (Dierschke et al., 2016). Accurately surveying the abundances and distributions of these species

is therefore key and historical estimates of at-sea marine birds were likely to have been substantial

underestimations, illustrated via mismatches in at-sea survey estimates and predictions of wintering

population derived from colony counts (Harris & Wanless, 2011). Over the last decade, offshore

windfarm EIAs have attempted to account for availability bias, by considering the time that individuals

from particular locations at specific times of year spend submerged, as a proportion of total time spent

on water. At present, correction factors are applied only to birds visible on the water's surface during

surveys, with birds in flight being added later (e.g., Harker et al., 2022). Often only a single correction

factor is used for each species, this factor frequently having been measured during the breeding

season and then applied over the whole annual cycle. For example, when accounting for availability

bias within monthly at-sea digital aerial surveys of marine birds in waters off south-east Scotland

across an annual cycle, Harker et al. (2022) followed other similar assessments in using values for

puffins derived from those breeding at Petit Manan Island, Maine, USA (Spencer, 2012), for guillemots

and razorbills from those breeding at the Isle of May, Scotland (Thaxter et al., 2010), and for red-

throated divers, no correction factors are regularly used (Irwin et al., 2019). However, the ever

increasing number of biologging studies from across ocean basins (e.g. Lescure et al., 2023), means

that there is no longer a need to rely on these commonly used factors from specific areas. Indeed,

continued advances in biologging provide opportunities to quantify the time that different species of

diving marine birds spend above or below the surface of the sea at different times of year (Duckworth

et al., 2021), enabling the derivation of correction factors for availability bias throughout the annual

cycle. Together, this increased understanding now enables us to fine tune marine bird abundance

estimates, accounting for variation in behaviour in both space and time.

Marine bird diving behaviour varies spatially, temporally, and interspecifically, being influenced by

foraging strategies, energy requirements, life cycle stages, and environmental conditions (Phillips et

al., 2017). Puffins, guillemots, razorbills, and red-throated divers breed once annually during the late

spring – summer at sites around the coastlines of the Atlantic and Pacific oceans, flying and diving

intensively during this period as they forage both to feed themselves and their chicks (Dunn et al.,

2019; Eriksson et al., 1990). These species then spend the rest of the annual cycle (the non-breeding

season) largely at sea, flying less and often experiencing harsher weather conditions and heightened

energetic costs, with consequences for their diving behaviour that vary between species (Duckworth

et al., 2021; Dunn et al., 2019, 2020). Furthermore, whilst these marine birds are highly adapted to

foraging on benthic and pelagic prey (Kleinschmidt et al., 2019; Linnebjerg et al., 2013), they differ in

their ability to forage uniformly throughout the diel cycle, with the diving behaviour of puffins and

red-throated divers, in particular, largely being limited to daylight hours (Duckworth et al., 2020; Shoji

et al., 2015). Throughout the autumn and winter, as daylight hours become increasingly constrained,

varying latitudinally and therefore between non-breeding areas, the proportion of daylight hours that
these birds spend submerged may increase. Variation in behaviour over time and between
populations, caused by these various drivers, will therefore result in differing magnitudes of
availability bias. Failing to account for this variation will lead to inaccuracies in estimates of marine
bird abundance.

Here, we use published biologging data on the dive behaviour of puffins, guillemots, razorbills, and
red-throated divers wintering across northwest Europe to quantify broad scale temporal and spatial
variation in the proportion of time that they spend below the water's surface. From these data, we
calculate new species-, month- and area- specific correction factors for puffins and razorbills in the
North Sea, guillemots in the North Sea or off the west coast of the UK, and red-throated divers in the
North Sea and Baltic Sea, within eastern Icelandic coastal waters, and around the coast of northern
Scotland in the months following their breeding seasons. With these examples, we show the
importance of accounting for temporal and spatial differences in availability bias.

Materials and methods

<u>Biologging deployments</u>

Breeding adult puffins, guillemots, razorbills, and red-throated divers were captured from their
breeding sites at sites in Scotland, Finland, and Iceland (Figure 1) during the breeding season (for
details, see Table 1). All birds were fitted with time-depth-recorders (G5, CEFAS, Lowestoft UK,
31 × 8 mm, 2.7 g in air) attached to Darvic leg-rings. In all cases the attachment process took <10 min
and the weight of the logger was <2% of the total body mass of the bird. Birds were recaptured during
subsequent breeding seasons and the devices were removed. For more details on the puffin and
razorbill deployments see Dunn et al. (2019), for guillemot deployments see Buckingham et al. (2023),
and for red-throated diver deployments see Duckworth et al. (2021). Sampling periods varied (Table
1) due to some TDRs failing during the autumn and winter.

Table 1. Details of the time-depth-recorder loggers deployed on and retrieved from Atlantic puffins, common guillemots,
razorbills, and red-throated divers at breeding sites across northwest Europe. For more details on the puffin and razorbill
deployments see (Dunn et al., 2019), for guillemot deployments see (Buckingham et al., 2023), and for red-throated diver
deployments see (Duckworth et al., 2021).

Species	Winter area	Breeding site	Deployment year	N $_{deployed}$ $_{and\ retrieved}$	Sampling regime	Sampling Period
Atlantic puffin	North Sea	Isle of May, Scotland	2008	30, 12	Either 3 s for a 24 h period every 10 d, or 30 s for a 24 h period every day	July – December
Common guillemot	UK West Coast	Colonsay and Treshnish, Scotland	2019	30, 19	Every 12 s for a 24 h period every 5 days	July – March
	North Sea	Whinnyfold and Isle of May, Scotland	2019	31, 28	Every 12 s for a 24 h period every 5 days	July – March
Razorbill	North Sea	Isle of May, Scotland	2008	24, 12	Either 3 s for a 24 h period every 10 d, or 30 s for a 24 h period every day	June – January
Red-throated diver	North Sea and Baltic Sea	Southern Finland	2018	31, 8	6 s resolution over a 24 h period every 5 days	September – December
	Icelandic coastal waters	Northeast Iceland	2018	12, 8	6 s resolution over a 24 h period every 5 days	September – December
	Northern Scotland	Orkney and Shetland, Scotland	2018	31, 7	6 s resolution over a 24 h period every 5 days	September – December

Data processing

All data processing and analyses were conducted in R version 4.2.2 (R Core Team, 2022). First, we

curtailed our analyses to the non-breeding period by removing data gathered prior to mean

population fledging dates (Duckworth et al., 2022; Dunn et al., 2019; Wanless et al., 2023). We then

extracted all incidences where time-depth-recorder depth values exceeded 1 m. To categorise these

incidences of submergence behaviour as occurring during either daylight, twilight, or nighttime, we

extracted time- and population-specific area estimates. To do this, we obtained centroids of each

population's monthly non-breeding distribution from previous work using light-based geolocator

devices (Buckingham et al., 2022, 2023; Duckworth et al., 2022; Harris et al., 2010) (Supporting

Information S1). We used the '*oce*' R package to extract sun elevation angles at these areas and

categorised each submergence incidence as occurring during daylight (sun was >0 degrees above the

horizon), twilight (sun was 0–15 degrees below the horizon), or nighttime (sun was >15 degrees below

the horizon) accordingly. We used a cut-off of 15 degrees below the horizon to account for uncertainty

in our area estimates. We used the same approach to estimate the duration of daylight hours for each

day. We conducted a sensitivity analysis to investigate the impact of using the population's monthly

centroid on our results using the guillemot data (due to them having the most data available and a

large latitudinal spread). We calculated the mean latitude plus and minus the standard deviation from

all the locations for each population in each month, and this showed that using the lower and upper

standard deviation values (instead of the centroid) led to negligible differences in our results

(Supporting Information S2).

We counted all incidences of submergence that took place during daylight hours, multiplied this by

the sampling interval for that individual, and divided this value by the duration of daylight to calculate

the proportion of daylight hours during which each individual was submerged. We excluded twilight

and nighttime hours from our analyses as visual marine bird surveys occur during daylight hours only

(NatureScot, 2023b). This metric encompasses all behaviour and should be applied to the total count

of birds typically captured during daytime aerial surveys, including both those on the surface and in

flight. By doing this, we assumed that the proportions of time that birds spend engaged in their key

behaviours (i.e., flying, diving and on water) are equivalent to the proportion of individual birds

counted (or not, when they are submerged) engaging in these behaviours during surveys. This

assumption underpins similar corrections of survey data which are made using time-budget data

(Barlow et al., 1988).

<u>Statistical analyses</u>

We composed a set of generalised linear mixed effects models to quantify the proportion of daylight

hours spent submerged by each species based on the explanatory terms available for that species,

using the Stan programming language via the '*brms*' R package (Bürkner, 2017). As we were modelling

proportional data, we used the Beta distribution that provided flexibility but was limited to between

0 and 1 (Heiss, 2021). For the razorbill and puffin models we included an explanatory term of month

only (because data were only available for one area; Tables 1 & 2), whereas for the red-throated diver

and guillemot models we include month, area, and their interaction (Table 2). We included month as

a categorical representation of time within our models, rather than as a continuous covariate, because

marine bird surveys are typically conducted at monthly intervals (NatureScot, 2023a). For all models

we included individual bird ID as a random effect.

We adjusted the priors of the coefficients so that they had a normal distribution centred at 0 with a

standard deviation of 1 and kept the default brms random effect priors. For the model intercepts, we

incorporated prior knowledge in the form of values that are currently used to correct for availability

bias in these species (Harker et al., 2022). We therefore logit-transformed the inverse of the

proportion of time spent at the surface in puffins (Spencer, 2012), guillemots (Thaxter et al., 2010),

razorbills (Thaxter et al., 2010), and red-throated divers (using a value for great northern divers *Gavia*

immer; Winiarski et al., 2014). To eliminate divergent transitions that caused a bias in the posterior

draws of the puffin model, we removed the December data (n_{days} = 1) and one outlier value (<4

minutes submerged). Whilst we initially ran all models with 4 MCMC chains and with 2,000 iterations

per chain, including warmup, we increased the number of iterations of the puffin model to 3,000. We

confirmed model convergence via visual inspection of the chains, posterior predictive check plots, and

by calculating a Gelman-Rubin convergence statistic ($\hat{R}$) of close to 1 (McElreath, 2020). We used the

'*marginaleffects*' R package (Arel-Bundock, 2023) to calculate the marginal effects of each unit change

in the explanatory variables. For puffins, guillemots, and razorbills, we used these values to make

comparisons between our model estimates of the proportions of daylight hours that the species spent

submerged, and the proportion of time spent on the water during which birds were submerged, the

quantity used previously within offshore EIAs (Harker et al., 2022). We note that these values are

therefore not identical in their derivation, due to us not evaluating flight behaviour here. However,

they remain comparable due to the very small proportion of time (typically <2%) that these species

spend engaged in flight during the non-breeding season (Duckworth, 2023; Dunn et al., 2020).

Table 2. Details of the generalised linear models run to model the proportion of daylight hours spent submerged by Atlantic
puffins, common guillemots, razorbills, and red-throated divers from breeding sites across northwest Europe during the non-
breeding season.

Model	Explanatory variables	Categories		Intercept prior
Puffin	Single variable:	Month	Aug, Sep, Oct, Nov	$N(-1.802, 0.1)$
Guillemot	Two-way interaction:	Month	Jul, Aug, Sep, Oct, Nov, Dec, Jan, Feb, Mar	$N(-1.150, 0.1)$
		Area	UK East Coat, UK West Coast	
Razorbill	Single variable:	Month	Jul, Aug, Sep, Oct, Nov, Dec, Jan	$N(-1.504, 0.1)$
Red-throated diver	Two-way interaction:	Month	Sep, Oct, Nov, Dec, Jan, Feb, Mar	$N(-0.847, 0.1)$
		Area	Baltic Sea and North Sea, Icelandic coastal waters, Northern Scotland	

We provide values of $Pr(being\ visible)$, the probability of an animal being at the water's surface or

in flight and therefore available to be recorded on a survey, for the months of the non-breeding season

for puffins in the North Sea, guillemots in the North Sea and the UK west coast, razorbills in the North

Sea, and red-throated divers that move between the Baltic Sea and the North Sea, those within

Icelandic coastal waters, and those off the north coast of Scotland (Table 3). These values were derived

by subtracting the posterior distributions of the models' outputs for the proportion of daylight hours

spent submerged (i.e., the mean estimate ± lower and upper 95% confidence intervals) from 1. Values

of $Pr(being\ visible)$ can be used to correct relative abundance estimates of birds by dividing the

density of birds observed sitting on the sea and in flight by $Pr(being\ visible)$ (Barlow et al., 1988).

Thus, the values generated by this study can be used in offshore windfarm assessments in a similar

way that existing availability bias correction factors are currently used.

Results

There was temporal variation in the proportion of daylight hours that puffins, guillemots, razorbills, and red-throated divers spent submerged over the months that followed the end of the species' breeding seasons. Birds spent an average of 1.5 hours submerged each day, this varying between 1% and 59% of daylight hours spent submerged.

The proportion of daylight hours that puffins in the North Sea spent submerged was lowest in July and August (Fig 2A). In July, the confidence intervals encompassed the proportion of time on water submerged previously used to account for availability bias in this species (Spencer, 2012). Although there were wide confidence intervals surrounding estimates of the proportion of daylight hours that puffins spent submerged, this value increased throughout September, October, and November, again overlapping with the previously used value in October before exceeding it in November (Fig 2A).

The temporal pattern in the proportion of daylight hours submerged was similar in razorbills wintering around the east coast of the UK; they spent little time submerged below the water's surface during the daytime periods that immediately followed the breeding season (July – August), but this increased over the course of the winter, only overlapping with a previously used value in January (Thaxter et al., 2010).

Figure 2. Marginal model predictions of the proportion of daylight hours spent submerged by Atlantic puffins (A) and
razorbills (B) wintering within the North Sea, averaged by month, between August and November and July and January for
puffins and razorbills, respectively. Blue points illustrate mean marginal predictions and shaded lines illustrate 95%
confidence intervals, and the raw data are shown as faint open circles behind. The horizontal dashed lines indicate values of
the proportion of time on water submerged previously used during environmental impact assessments for these species (see
text for details).

The proportion of daylight hours that guillemots spent submerged during their non-breeding season
was similar between birds wintering in both the North Sea and off the UK west coast (Fig 3). Initially,
following their breeding season, birds spent a relatively low proportion of daylight hours submerged
beneath the water's surface. Indeed, values between July – October were lower than values of the
proportion of time on water submerged previously used to correct for availability bias, only exceeding
this value in November for birds wintering in the North Sea, and December for birds wintering off the
UK west coast (Fig 3). The proportion of daylight hours spent submerged then increased for both
groups throughout the sampling period until March (Fig 3).

Figure 3. Marginal model predictions of the proportion of daylight hours spent submerged by common guillemots wintering
(A) in the North Sea and (B) off the UK west coast, averaged by month, between July and March. Blue points illustrate mean
marginal predictions and shaded lines illustrate 95% confidence intervals, and the raw data are shown as faint open circles
behind. The horizontal dashed lines indicate a value of the proportion of time on water submerged previously used during
environmental impact assessments for this species (see text for details).

Red-throated divers wintering in different areas spent differing proportions of daylight hours submerged. Birds that initially wintered within the Baltic Sea, before moving to the southern North Sea during October, spent an increasing amount of time below the water's surface following their breeding seasons (Fig. 4A). Although a similar pattern was observed in red-throated divers that wintered around north Scotland, with the proportion of daylight hours submerged increasing between September and December, there was a slight decrease in time spent submerged in January (Fig. 4C). A similar dip in the proportion of daylight hours submerged in December was observed in red-throated divers that wintered in Icelandic coastal waters, with high values being observed in January and February, before another decrease in March (Fig. 4B).

Figure 4. Marginal model predictions of the proportion of daylight hours spent submerged by red-throated divers wintering (A) in the Baltic Sea during September and October and then in the North Sea during November – January, (B) in Icelandic coastal waters, and (C) off north Scotland, averaged by month, between September and March. Blue points illustrate mean marginal predictions and shaded lines illustrate 95% confidence intervals, and the raw data are shown as faint open circles behind.

Temporal, spatial, and interspecific variation in the proportion of daylight hours spent submerged by the puffins, guillemots, razorbills, and red-throated divers led to variation in values of $Pr(being\ visible)$, the probability of each species being available to be recorded at the water's surface or in flight during surveys (Table 3). These values vary between 0.67 (95% confidence interval = 0.62 – 0.71) for guillemots wintering in the North Sea in March, and 0.95 for razorbills in a similar area in June (95% confidence interval = 0.93 – 0.96).

Table 3. Estimates (and lower and upper 95% confidence intervals in italics) of the probability of being available at the surface or in flight, $Pr(being\ visible)$, for Atlantic puffins, common guillemots, razorbills, and red-throated divers from breeding sites across northwest Europe during the months of the non-breeding season. These correction factors can be applied to survey estimates by dividing estimates of the density or abundance of birds in flight and on water by $Pr(being\ visible)$ to give corrected estimates.

Species	Marine area	$Pr(being\ visible)$									
		Jun	Jul	Aug	Sep	Oct	Nov	Dec	Jan	Feb	Mar
Atlantic puffin	North Sea			0.89 *0.87 − 0.9*	0.84 *0.81 − 0.86*	0.81 *0.72 − 0.86*	0.78 *0.71 − 0.83*				
Razorbill	North Sea		0.92 *0.91 − 0.93*	0.93 *0.92 − 0.93*	0.89 *0.88 − 0.9*	0.88 *0.87 − 0.89*	0.85 *0.84 − 0.86*	0.88 *0.86 − 0.89*	0.81 *0.79 − 0.84*		
Common guillemot	North Sea		0.86 *0.84 − 0.88*	0.88 *0.86 − 0.89*	0.87 *0.85 − 0.89*	0.82 *0.79 − 0.84*	0.76 *0.73 − 0.79*	0.71 *0.68 − 0.74*	0.71 *0.67 − 0.75*	0.69 *0.65 − 0.73*	0.67 *0.62 − 0.71*
	UK west coast		0.87 *0.85 − 0.89*	0.9 *0.89 − 0.92*	0.9 *0.88 − 0.92*	0.85 *0.82 − 0.87*	0.82 *0.79 − 0.85*	0.71 *0.67 − 0.75*	0.7 *0.65 − 0.74*	0.7 *0.65 − 0.74*	0.71 *0.66 − 0.75*
Red-throated diver	Baltic Sea				0.92 *0.88 − 0.94*	0.86 *0.8 − 0.9*					
	North Sea						0.77 *0.68 − 0.83*	0.72 *0.63 − 0.8*	0.72 *0.59 − 0.82*		
	Icelandic coastal waters				0.89 *0.84 − 0.93*	0.82 *0.74 − 0.87*	0.83 *0.75 − 0.88*	0.94 *0.91 − 0.97*	0.81 *0.71 − 0.88*	0.83 *0.71 − 0.9*	0.93 *0.88 − 0.96*
	North Scotland				0.89 *0.86 − 0.92*	0.85 *0.82 − 0.89*	0.83 *0.78 − 0.87*	0.8 *0.74 − 0.84*	0.85 *0.78 − 0.9*		

Discussion

There was substantial variation in the proportion of daylight hours spent submerged by puffins, guillemots, razorbills, and red-throated divers in different marine areas throughout their non-breeding seasons. This variation led to differences in the probability of them being available for detection by commonly used survey methods ('availability bias', defined as '$Pr(being\ visible)$'), ultimately influencing estimates of their absolute densities. Our results suggest that previously used correction factors for availability bias (often derived from individuals of different species, or in different areas, or during different seasons) likely provided incorrect estimates of marine bird abundances, the degree of which would have varied by species, time, and area. Indeed, for red-throated divers where no correction factor is regularly used, surveys are currently likely to often yield underestimates of their non-breeding abundances. We therefore encourage the use of correction factors that are estimated from the species, marine area, and month during which the survey is conducted. Our study highlights the importance of using appropriate input data (i.e., data that is as closely related to the surveyed population as possible) when quantifying species abundances and distributions, evidence that is relevant for various applied science studies. We present new values through which to account for variation in availability bias for our four focal species and their corresponding marine areas (Table 3).

Our increased understanding of marine behaviour outside the breeding season is highlighting a mounting number of differences when compared to behaviour during the breeding season (Drummond et al., 2021). In this study, the proportion of daylight hours that all four species spent submerged below the water's surface varied across the non-breeding season, generally demonstrating an increase over this time (Fig 2-4). This temporal variation is likely to be representative of fluctuations in climatic conditions, and the birds' subsequent energy requirements and activity budgets (Buckingham et al., 2023; Clairbaux et al., 2021; Fort et al., 2009). Differences in dive behaviour can also be compounded by variation in daylight availability impacting the ability of these visual foragers to gain the energy needed to meet their energetic demands, therefore driving changes in their time activity budgets (Duckworth et al., 2021; Dunn et al., 2020). Indeed, the time that marine birds spend submerged varies over the course of the diurnal cycle (Jardine, 2024), and should be considered within future work. Other life history events such as feather moult (Cherel et al., 2016) and migratory travel can also influence marine bird activity budgets, with migratory populations of red-throated divers (Fig 4A) spending a higher proportion of time submerged than more residential populations (Fig 4B-C). This variation in activity is likely due to a requirement to adjust their behaviour as they moved from one wintering area to another in addition to needing to meet the high energetic costs often associated with travel in marine birds (Dunn et al., 2023).

Here we show that accounting for temporal and spatial variation in behaviour is critical to enhancing

our understanding of the consequences of anthropogenic change. In particular, seasonal change in

availability bias would strongly affect estimates of seasonal abundance. For example, for common

guillemots off the west of Scotland, the correction factor was three times higher in February than in

September (Figure 3). It is important that EIA and other assessments do not incorrectly estimate

marine bird abundances, something that can occur if inappropriate availability bias values continue to

be used. Our results suggest that a hypothetical set of sequential surveys might show a decreasing

trend in guillemot abundance if a fixed correction factor was used, but in reality, the trend could be

stable if the appropriate varying proportion of time submerged (and hence correction factor) was

used. This reinforces the recommendation to use correction factors that acknowledge these temporal

trends to garner more accurate estimates of the at-sea abundances of these species outside their

breeding seasons.

EIAs for offshore renewable projects typically derive corrected estimates of marine bird abundances

and distributions using corrections for availability bias, but here we show that current practices may

not be dependable. Currently, raw counts of birds on the water's surface are divided by

$Pr(being\ visible)$, therefore increasing the estimate above the original count, before adding the

number of flying birds. For example, if 3,000 common guillemots were recorded on a digital aerial

survey in the North Sea during September, previously used correction factors would have estimated

that there were 3,947 birds, with no associated availability bias error. Instead, our species-, time-, and

area-specific correction factor (outlined in Table 3) would have produced an estimate of the number

of birds in the area being 3,448 individuals (95% confidence interval = 3,370 – 3,529). Furthermore, if

in January the survey recorded 3,000 common guillemots, 10% of which were in flight, previously used

correction factors would have estimated 3,938 birds, with no associated availability bias error.

Contrastingly, our correction factor (Table 3) would instead suggest that the number of birds likely in

the area was 4,225 individuals (95% confidence interval = 4,000 – 4,478). Furthermore, in red-throated

divers where no correction factors are currently regularly used, population estimates may be missing

almost one third of the red-throated divers in the North Sea during December and January (Figure 4A).

These examples demonstrate that previous estimates of the abundances and distributions of these

marine birds during winter may well have been incorrect, as they fall outside the confidence intervals

generated via this new approach, with implications for assessing the potential consequences of

anthropogenic activities such as offshore windfarm developments.

Our study shows how incorporating our increasing understanding of temporal and spatial variation in

the behaviour of non-breeding marine birds has consequences for the assessment of anthropogenic

activities within the marine environment. The data presented here are most relevant to EIA within the

context of offshore wind development but would also have relevance for protected area designation
and the investigation of long-term change across marine systems. We encourage those making
assessments elsewhere to use and/or collect the most appropriate information to their surveyed areas
or studied populations, as was originally conceived in attempts to correct for availability bias (Barlow
et al., 1988). In doing so, we recommend the use of larger sample sizes, longer term logger
deployments, and improved non-breeding area estimates to enhance future insights, something that
is becoming increasingly tangible with the continued modernisation and miniaturisation of biologging
devices. As these developments continue, it will be important to also consider longer-term interannual
temporal variation, and the movements of juvenile and immature individuals (Fayet et al., 2024;
Merkel & Strøm, 2023). All of this is particularly timely during a period when marine birds are suffering
catastrophic declines (Dias et al., 2019) in tandem with our oceans becoming increasingly intensively
used for renewable energy developments, oil and gas drilling, transport, and sand extraction (Halpern
et al., 2019). Ultimately, when seeking to monitor species abundances and distributions we must
collect, harness, and appropriately interpret relevant datasets, embracing the nuances and seeking to
understand the uncertainty of these data and the underlying ecological processes (Chadwick et al.,
2023).

Authors' contributions

Sue O'Brien, Jonathan Green, and James Duckworth conceived the ideas; Ruth Dunn, James Duckworth and Lila Buckingham managed the data; Ruth Dunn analysed the data; Ruth Dunn led the data visualisation with marine bird illustrations provided by Lila Buckingham; Ruth Dunn led the writing of the manuscript with support from Jonathan Green. All authors contributed critically to the drafts.

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All applicable institutional and/or national guidelines for the care and use of animals were followed.

Conflict of Interest

The authors declare no conflict of interest.

Data availability statement

No new data were collected for the study.

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