

Title	Multi-scale analysis reveals changing distribution patterns and the influence of social structure on the habitat use of an endangered marine predator, the sperm whale Physeter macrocephalus in the Western Mediterranean Sea
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Publication date	2019-11-18
Original Citation	Pirotta, E., Brotons, J. M., Cerdà, M., Bakkers, S. and Rendell, L. E. (2020) 'Multi-scale analysis reveals changing distribution patterns and the influence of social structure on the habitat use of an endangered marine predator, the sperm whale Physeter macrocephalus in the Western Mediterranean Sea', Deep Sea Research Part I: Oceanographic Research Papers, 155, 103169 (10 pp). doi: 10.1016/j.dsr.2019.103169
Type of publication	Article (peer-reviewed)
Link to publisher's version	http://www.sciencedirect.com/science/article/pii/ S0967063719303152 - 10.1016/j.dsr.2019.103169
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Download date	2024-03-28 19:29:25
Item downloaded from	https://hdl.handle.net/10468/9676



Multi-scale analysis reveals changing distribution patterns and the influence of social structure on the habitat use of an endangered marine predator, the sperm whale *Physeter macrocephalus* in the Western Mediterranean Sea

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PII: S0967-0637(19)30315-2

DOI: https://doi.org/10.1016/j.dsr.2019.103169

Reference: DSRI 103169

To appear in: Deep-Sea Research Part I

Received Date: 19 August 2019
Revised Date: 28 October 2019
Accepted Date: 13 November 2019

Please cite this article as: Pirotta, E., Brotons, J.M., Cerdà, M., Bakkers, S., Rendell, L.E., Multi-scale analysis reveals changing distribution patterns and the influence of social structure on the habitat use of an endangered marine predator, the sperm whale *Physeter macrocephalus* in the Western Mediterranean Sea, *Deep-Sea Research Part I*, https://doi.org/10.1016/j.dsr.2019.103169.

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1	Multi-scale analysis reveals changing distribution patterns and the influence
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Abstract

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The habitat use of marine megafauna emerges from the complex interplay between access to 19 patchy and variable food resources and several intrinsic biological factors, such as the interaction 20 with conspecifics and offspring care, resulting in dynamic distribution patterns. Quantifying 21 species' relationships with the underlying environment is further complicated by the scale-22 23 dependent nature of these processes. Multi-scale analyses that incorporate aspects of a species' 24 biology and build on large datasets are therefore required to understand long-term distribution and inform appropriate management measures. In this study, we use monitoring data collected 25 over two study periods (2003-2008 and 2012-2018) to assess the habitat use, trend in local 26 occurrence, and change in distribution of sperm whales, *Physeter macrocephalus*, around the 27 Balearic Islands (Spain), one of the few recognised breeding and feeding grounds for the 28 'Endangered' population in the Mediterranean Sea. Moreover, we investigate the differences in 29 30 the habitat use of single animals and groups, to explore intra-specific niche partitioning in this highly social but behaviourally dimorphic species. Results suggest that overall the occurrence of 31 32 sperm whales in the area has been increasing over time. Animals were found to associate with distinct bathymetric features, but the mechanisms generating these relationships, and the 33 underlying oceanographic processes within this habitat, remained uncertain. Sperm whale 34 distribution also underwent a significant shift between the two study periods, with an increased 35 occurrence in the Mallorca channel and north of Menorca, which further points towards a 36 dynamic use of the broader bathymetric range preferred around the archipelago. Finally, our 37 analyses highlighted that single animals and groups used areas with different characteristics, with 38 groups preferring deeper, warmer waters characterised by lower sea level anomaly, which 39 resulted in some fine-scale spatial segregation. The results of this study shed light on the 40

- 41 mechanisms underpinning the biogeography and complex social system of the species, and
- 42 support the design of targeted conservation measures in this important breeding and feeding
- 43 ground.

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- 45 Keywords: Habitat modeling, Distribution shift, Long-term monitoring, Sperm whale,
- 46 Mediterranean Sea, Balearic archipelago.

1. Introduction

49	Quantifying the relationships between spatio-temporal patterns of animal occurrence and the
50	underlying environment is central to the understanding of a species' ecology (Guisan and
51	Thuiller, 2005). However, habitat modelling generally relies on data that have been collected
52	within a limited time frame and therefore only provides a snapshot of the distribution of a
53	species, unless multiple surveys are combined (Yates et al., 2018). Moreover, both habitats and
54	habitat use can be dynamic, and therefore the importance of different areas can change over time
55	(Roshier and Reid, 2003). This is particularly true for marine megafauna, which can often adjust
56	their habitat use as intrinsic (e.g., population size, habitat knowledge) and extrinsic (e.g.,
57	environmental quality) conditions change (Runge et al., 2014). For these species, data collection
58	tends to be constrained to short spatial and temporal windows by logistic and financial
59	limitations (Yates et al., 2018), which restricts our ability to appropriately describe their
60	distribution and complicates the development of effective conservation measures (Lewison et al.,
61	2015; Wilson, 2016). Understanding the dynamic nature of a species' distribution and the degree
62	of plasticity is particularly important as the marine environment is undergoing unprecedented
63	change as a result of climate change (Hazen et al., 2013).
64	Mechanistically, the availability and abundance of prey resources is one of the major drivers of
65	the movements and distribution of marine predator populations, including marine mammals
66	(Sequeira et al., 2018). Often, prey cannot be sampled at appropriate resolutions (Redfern et al.,
67	2006), but marine mammal occurrence tends to associate with the oceanographic features that
68	affect prey patterns, which can in turn be used in statistical models as proxies of the underlying
69	processes (Elith and Leathwick, 2009). However, identifying the relevant spatio-temporal scale
70	for representing these indirect relationships is challenging (Scales et al., 2017), and the dynamic

71	nature of predator and prey distribution may mean that suitable proxies change over space and
72	time. Intrinsic biological drivers also play a role in determining the observed use of space
73	(Cañadas and Hammond, 2008; Guisan and Thuiller, 2005; Palacios et al., 2014). For example,
74	marine mammals may adjust their distribution to ensure safety for their offspring, interact with
75	members of the same social group, or in response to intra-specific competition for food
76	resources. Intrinsic drivers expressed in behaviour therefore add a layer of complexity that
77	further confounds the relationships with concurrent environmental features. The investigation of
78	intra-specific differences in habitat use could shed light on the forces that regulate interactions
79	with conspecifics and, potentially, the evolution of current biogeographical patterns.
80	The sperm whale (<i>Physeter macrocephalus</i>) is a cosmopolitan cetacean species found across the
81	world in deep waters beyond the shelf edge, where bathymetric and oceanographic features
82	interact to promote upwelling, nutrient mixing and, ultimately, secondary productivity and prey
83	concentrations (Whitehead, 2018). As a result, sperm whales are often encountered in association
84	with steep continental slopes, submarine canyons and seamounts, as well as frontal systems and
85	other mesoscale features such as cyclonic eddies (e.g., O'Hern and Biggs, 2009; Skov et al.,
86	2008; Waring et al., 2001; Wong and Whitehead, 2014), but relationships can be obscured by
87	analyses at inappropriate scales that do not match the spatial and temporal scale at which these
88	oceanographic processes occur (Jaquet, 1996). This species is also characterised by a complex,
89	sexually dimorphic social system, whereby females form strong social units that remain at lower
90	latitudes ($<40^{\circ}$), while males become increasingly solitary as they grow older, migrating to
91	higher latitudes to find food resources that can support their larger size (Whitehead 2003). The
92	evolution of sperm whale social structure may have been promoted by intra-specific competition
93	between males and females, as suggested by observations of lower feeding success of males in

areas where they co-occur with females (Whitehead 2003). In turn, strong female social bonds
are believed to be the basis for the development of one of the recognized examples of non-human
culture (Rendell and Whitehead, 2003; Whitehead and Rendell, 2014), which can also affect
patterns of spatial distribution (Eguiguren et al., 2019; Whitehead and Rendell, 2004).
A small, genetically isolated population of sperm whales inhabits the Mediterranean Sea
(Rendell and Frantzis, 2016). This population is subject to intense pressure from the extensive
human activities in the basin, leading to ship strikes, entanglement in driftnets, and ingestion of
plastic, as well as exposure to noise, and chemical pollution. Consideration of both the size of the
population and the threats it faces has prompted its classification as 'Endangered' in the
International Union for Conservation of Nature (IUCN) Red List (Rendell and Frantzis, 2016).
Previous studies have identified a bimodal distribution in the Mediterranean, characterised by the
association with topographic singularities close to the coast, and thermal fronts in offshore areas
(Azzellino et al., 2012; Frantzis et al., 2014; Gannier and Praca, 2007; Pirotta et al., 2011; Praca
et al., 2009; Tepsich et al., 2014). The latitudinal segregation between sexes is much reduced
owing to the limited available latitudinal range in the Mediterranean Sea for the more mobile
males to explore (Drouot-Dulau and Gannier, 2007). The Balearic archipelago (Spain) is one of
the few areas in the Mediterranean Sea where females and calves as well as single males are
regularly observed (Rendell and Frantzis, 2016). This led to its recent identification as an
Important Marine Mammal Area (IMMA; https://www.marinemammalhabitat.org/portfolio-
item/balearic-islands-shelf-slope/; Corrigan et al., 2014), which was informed by some of the
data presented here. In this area, female groups and single males appear to segregate at a fine
spatial scale (Jones et al., 2016).

Pirotta et al. (2011) analysed monitoring and encounter data collected over six consecutive years (2003-2008) around the Balearic archipelago to describe the distribution patterns of the species in this important breeding and feeding ground. In this study, we complement the original dataset with data collected over a subsequent study period of seven years (2012-2018), to evaluate sperm whale medium-term habitat use in the area and assess changes in their occurrence and distribution over time. In addition, we use information on social grouping to investigate the environmental variables associated with the occurrence of groups (mostly females and young individuals) and single animals (likely males), and shed light on the processes that regulate the observed fine-scale segregation. Our results can inform effective management strategies in the region, and support the conservation of the population in the Mediterranean Sea.

2. Material and methods

128 2.1 Data collection

Data collection methods followed the procedures described in Pirotta et al. (2011). Briefly, dedicated research cruises were carried out in the summer months over two study periods, 2003-2008 and 2012-2018 (Table 1 and Fig. A.1), from 11- to 15-m-long motor-sailing yachts travelling at a speed of approximately 6 knots. Sperm whale presence was monitored acoustically every 30 min using, in 2003, a single dipping hydrophone (Sensor Technology of Canada; frequency response 0.1–22 kHz) and, from 2004 onward, a dual-element towed hydrophone (Benthos AQ4; frequency response 0.1–22 kHz) towed at 100 m. Hydrophones deployed at similar speeds and from similar vessels typically settle at around 10 m depth. Transects were not systematically designed, but extensively covered the shelf-break waters around the archipelago

(Fig. 1). Detection range was unknown, but previous work in the Mediterranean has estimated an effective strip half-width of 10 km (Lewis et al., 2007). Whenever sperm whales were heard, dedicated software (RainbowClick by the International Fund for Animal Welfare (IFAW) before 2012; https://www.marineconservationresearch.co.uk/downloads/logger-2000-rainbowclick-software-downloads/; and PAMGuard https://www.pamguard.org/ from 2012 onward) was used to track and, during daylight hours, approach the animals. An encounter was defined as a period of continuous acoustic contact with the animals, from initial detection to a loss of contact greater than 1 h (either inadvertent or deliberate once all desired data were collected). Searching effort was then generally resumed along the previous route. Encounters with single individuals were distinguished from encounters with groups, defined as individuals (often including young animals) engaging in direct interaction or 'moving together in a coordinated fashion over periods of at least hours' (Whitehead, 2003).

Year	Study period		h period ffort)	Searching effort (km)	Encounters (km)	Encounters with groups	Encounters with single animals
2003		03-Aug	26-Aug	2467	141	2	1
2004		11-Jul	05-Aug	2070	310	3	6
2005	1	10-Jul	04-Aug	1992	392	5	7
2006	1	16-Jul	27-Jul	1702	270	2	7
2007		06-Jul	28-Jul	1835	374	0	11
2008		15-Jul	27-Jul	1033	233	6	6
2012		01-Aug	14-Aug	1224	57	1	3
2013		06-Jul	09-Aug	1460	451	7	18
2014		05-Jul	15-Aug	1046	432	16	7
2015	2	18-Jul	06-Aug	1559	407	3	18
2016		05-Jul	26-Aug	2340	601	7	26
2017		01-Jul	24-Sep	2463	592	5	24
2018		15-May	23-Sep	2379	591	12	11

Table 1. Summary of survey effort and encounters per year.

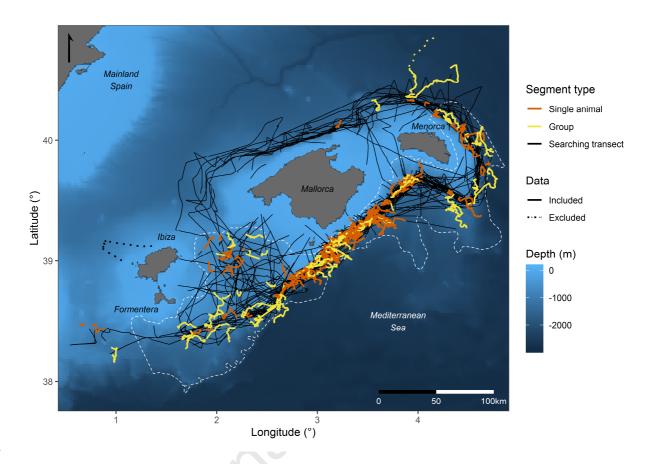


Figure 1. Study area, acoustic searching effort and encounters with group and single sperm whales in the period 2003-2018. Dotted segments indicate data that were excluded from subsequent analyses. The white, dashed line indicates the boundaries of the Important Marine Mammal Area (IUCN-MMPATF, 2017).

2.2 Data processing

For consistency with previous analyses, data were processed following the procedure described in Pirotta et al. (2011). When the hydrophone was not in the water or no systematic acoustic monitoring was conducted, corresponding GPS locations were discarded as off-effort. After preliminary data exploration, we also excluded the limited survey effort north-west of the island of Ibiza and truncated one encounter on 2 August 2004, when whales moved into an area that

163	was never surveyed otherwise, to avoid biasing model results (Fig. 1). On-effort locations were
164	regularised to 20 min intervals using package adehabitatLT (Calenge, 2006) for R (R Core Team,
165	2019). Points were classified as presences when in acoustic contact with the animals, or absences
166	when no animal was heard, and then grouped into either a follow (that is, a series of consecutive
167	presence points corresponding to an encounter with sperm whales) or a searching transect (that
168	is, a series of consecutive absence points between two follows or off-effort intervals).
169	Each point was associated with a set of static and dynamic variables, which were chosen to
170	represent the bathymetric and oceanographic processes that characterise sperm whale habitat,
171	and on the basis of their documented relationship with the species' occurrence. These included
172	depth, slope gradient (hereafter slope), slope aspect (hereafter aspect), standard deviation of
173	depth (hereafter rugosity), topographic position index (the difference between depth in a cell and
174	the mean value of the eight surrounding cells; TPI), sea surface temperature (SST), SST
175	deviation from monthly median (to allow whale presence to respond to relative temperatures,
176	because the median SST varied between years, thus better reflecting the choices available to the
177	animals; hereafter SST deviation), SST slope gradient (that is, the steepness of the SST surface,
178	representing potential frontal systems), standard deviation of SST, sea level anomaly (sea surface
179	height above the mean in a twenty-year reference period; SLA), and absolute dynamic
180	topography (instantaneous sea surface height above the geoid; ADT). Depth, slope and aspect
181	were considered at four spatial scales (30 arc-sec, 2.5 arc-min, 5 arc-min, 10 arc-min), while SST
182	was included at two spatial (4 km and 20 km) and two temporal scales (8-day and monthly
183	composites). A description of the available covariates, units, spatio-temporal scales and
184	corresponding datasets is reported in Table 2. Other commonly used environmental variables,

such as chlorophyll-a surface concentration, were not included because previous work suggested they were not related to sperm whale occurrence at this scale (Pirotta et al., 2011).

The heterogeneous distribution of searching effort around the archipelago and the variability among years could result in different habitats being surveyed with varying intensity over time, which may confound the relationship between sperm whale occurrence and the underlying environment. To account for this heterogeneity, we developed an effort covariate that summarized the amount of time spent in different sections of the study area in each year. Specifically, we counted the total number of regularised on-effort points that occurred within 0.5° x 0.5° grid cells in each year, and associated each presence and absence point with the effort value for the grid cell it was located in.

Covariate	Unit	Description	Origin	Spatial scales	Temporal scales
Depth	m	Depth of the	Obtained from the General	30 arc-sec,	-
		seabed	Bathymetric Chart of the Oceans	2.5 arc-min,	
			dataset (GEBCO;	5 arc-min,	
			http://www.gebco.net), and	10 arc-min	
			aggregated at multiple scales (x1,		
			x5, x10, x20) using package raster		
			for R (Hijmans, 2016)		
Slope	rise	Slope	Calculated from GEBCO dataset	30 arc-sec,	-
	over	gradient,	using package SDMTools for R	2.5 arc-min,	
	run	indicating the	(VanDerWal et al., 2014)	5 arc-min,	
		maximum		10 arc-min	
		rate of			
		change in			
		depth			
Aspect	0	Slope aspect,	Calculated from GEBCO dataset	30 arc-sec,	-
		indicating the	using package SDMTools for R	2.5 arc-min,	

		compass	(VanDerWal et al., 2014)	5 arc-min,	
		orientation of		10 arc-min	
		the slope			
		F			
Rugosity	m	Standard	Calculated from GEBCO dataset	2.5 arc-min	_
		deviation of	using package raster for R		
		depth	(Hijmans, 2016)		
			(11311111111111111111111111111111111111		
TPI	m	Topographic	Calculated from GEBCO dataset	30 arc-sec	_
		position	using package raster for R		
		index, i.e. the	(Hijmans, 2016)		
		difference	(22,22,22,22)		
		between the			
		value of			
		depth in a			
		cell and the			
		mean value	.01		
		of the eight			
		surrounding			
		cells			
SST	°C	Sea surface	Extracted from Moderate	4 km,	Monthly
		temperature	Resolution Imaging	20 km	and 8-day
			Spectroradiometer		composites
			(MODIS) data from NASA's Aqua		1
			satellite, processed by the Ocean		
			Biology Processing Group of the		
			Ocean Ecology Laboratory at		
			NASA Goddard Space Flight		
			Center (available at		
			http://oceancolor.gsfc.nasa.gov/)		
SST	°C	Deviation of	Calculated from MODIS-Aqua data	4 km	Monthly
deviation		the sea	using package raster for R		
		surface	(Hijmans, 2016)		
		temperature			
		in each cell			
		from the			
		monthly			
		median			
		median			

SST slope gradient	rise over run	Rate of maximum change in sea surface temperature	Calculated from MODIS-Aqua data using package SDMTools for R (VanDerWal et al., 2014)	4 km	Monthly and 8-day composites
Standard deviation of SST	°C	Variation in SST across a window of five cells	Calculated from MODIS-Aqua data using package raster for R (Hijmans, 2016)	20 km	Monthly and 8-day composites
SLA	m	Sea level anomaly, i.e. sea surface height above the mean sea surface in a twenty-year reference period (1993- 2012)	Obtained from the Copernicus Climate Data Store (CDS), operated by the European Centre for Medium Range Weather Forecasting on behalf of the European Union (https://cds.climate.copernicus.eu/)	0.125°	Daily
ADT	m	Absolute dynamic topography, i.e. the instantaneous height above the geoid	Obtained from the Copernicus Climate Data Store (CDS), operated by the European Centre for Medium Range Weather Forecasting on behalf of the European Union (https://cds.climate.copernicus.eu/)	0.125°	Daily

Table 2. Description of environmental variables used for the analysis of sperm whale habitat use,

including units, origin of the data, and spatio-temporal scales.

2.3 Statistical analysis

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- We structured the statistical analysis in three parts. First, we investigated the relationships
- 200 between sperm whale overall distribution and available environmental covariates, modelling

sperm whale acoustic presence or absence at each location. Secondly, we assessed whether there was any evidence of a change in distribution between the two study periods using a geographical surface, that is, a two-dimensional smooth of latitude and longitude. While the overall habitat use analysis could also highlight changes in distribution over time, this second model allowed us to explicitly test whether the animals were using different portions of the study area in the two periods. Finally, we evaluated any difference in the habitat used by single animals compared to groups. All analyses were carried out using package MRSea for R (Scott-Hayward et al., 2015), which uses a Spatially Adaptive Local Smoothing Algorithm (SALSA) with cross-validation to fit one-dimensional B-splines and Complex Region Spatial Smoothers (CReSS) (Scott-Hayward et al., 2014; Walker et al., 2011). In all analyses, locations spaced every 20 min were used as the unit of analysis, and the response variable had a binary distribution, which was modelled using a logit link function. Data from 2003-2008, which have already been presented in Pirotta et al. (2011), were reanalysed under the updated modelling framework to ensure that results could be compared between the two study periods.

2.3.1 Overall habitat use

All explanatory variables were standardised to facilitate model convergence. Potential issues of multicollinearity among available covariates were assessed using the variance inflation factor (VIF) and pairwise correlation plots, with values of VIF ≥ 2 and correlation ≥ 0.6 taken to indicate collinearity. Separate models including each of the collinear variables were fitted in MRSea, and the Akaike information criterion (AIC) used to compare model pairs. The same procedure was used for variables available at multiple spatial or temporal scales. The full model then included all non-collinear environmental variables, as well as the effort covariate and year. Aspect was included as a cyclic spline, to reflect the circular nature of this angular measurement.

224	For each one-dimensional smoother, SALSA uses cross-validation to identify the optimal
225	number and location of knots. The algorithm also evaluates whether the model is improved by
226	alternatively including each variable as a linear term (rather than smooth), or by removing it
227	altogether.
228	The autocorrelation function (ACF) plot was used to assess the degree of autocorrelation in
229	model residuals, and the final model was refitted as a working independence model in a
230	Generalised Estimating Equations (GEE) framework (Hardin and Hilbe, 2003), where follows
231	and searching transects represented blocks of correlated data points. Under this framework, a
232	sandwich variance estimator provides robust estimates of precision that account for the observed
233	degree of autocorrelation within each block (Hardin and Hilbe, 2003). The significance of
234	retained smooths was evaluated using Wald's tests based on robust standard errors, and the
235	performance of the final model was assessed using a confusion matrix, comparing predicted to
236	observed sperm whale occurrence at each location. The area under the receiver operating
237	characteristic (ROC) curve (AUC), calculated using package ROCR for R (Sing et al., 2005),
238	offered an additional measure of goodness-of-fit.
239	The estimated relationships between retained explanatory variables and the binary occurrence of
240	sperm whales were visualized using partial residuals plots, where 95% confidence intervals were
241	calculated using a parametric bootstrap of the GEE results (Pirotta et al., 2011; Scott-Hayward et
242	al., 2015); retained covariates were back-transformed to the original scale for ease of
243	interpretation. Model predictions were mapped for each study period using a regular grid of
244	0.01° x 0.01° cells, cropped to the geographical area covered by survey effort in those years.
245	Values of retained explanatory variables were extracted at the centroid of each cell, and
246	standardised according to the mean and standard deviation in the original data. For dynamic

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variables, the mean in the month of July (i.e., the median month in each period, and the most consistently data-rich month across the study period) across the years in each of the two study periods was used for predictions. Uncertainty in model predictions was plotted using 95% confidence intervals obtained from a parametric bootstrap of the GEE results. 2.3.2 Change in distribution in the two study periods The change in distribution of sperm whales between the two periods was investigated using the interaction between a CReSS smooth (that is, a bi-dimensional surface of geographic coordinates) and a categorical variable for study period, effectively fitting two separate spatial surfaces. A range of starting knots for the CReSS surface (4, 8, 12 or 16) was tested, which is advised to avoid the algorithm converging on local minima or maxima (Scott-Hayward et al., 2015). Standardised effort was also included as a one-dimensional smooth term. A similar procedure to the one described for the overall habitat model was followed for model selection, validation and prediction. Due to computing limitations related to the calculation of the distance matrix required by CReSS, spatial predictions were visualised on a coarser grid of 0.1° x 0.1° cells. 2.3.3 Differences between groups and single animals The third part of the analysis focused on presence points only. Encounter locations with groups were classified as 1s, while encounter locations with single animals were classified as 0s. The binary occurrence of groups (versus single animals) was then modelled as a function of environmental covariates following the same analytical procedure described for the analysis of sperm whale overall habitat use.

We then assessed the extent of the spatial overlap between groups and single animals using
Bhattacharyya's affinity (BA), where 0 corresponds to no overlap and 1 to complete overlap
(Bhattacharyya, 1943; Grecian et al., 2018). Following Grecian et al. (2018), the bivariate kernel
utilization distribution of groups and singletons was calculated (package adehabitatHR for R),
with smoothing parameter equal to 10 km and a grid of 1 x 1 km cells. BA for the two utilization
distributions was compared to a null distribution obtained by randomly reassigning encounters to
the two groupings and recalculating the utilization distributions 1000 times. This procedure
generated the expected distribution of BA values in the absence of segregation between groups
generated the expected distribution of BA values in the absence of segregation between groups and singletons.
and singletons.
and singletons. Finally, for each follow, we calculated the bearing (that is, the angle measured from the north)
and singletons. Finally, for each follow, we calculated the bearing (that is, the angle measured from the north) between the location of first acoustic contact and the last location, which provided a proxy for
and singletons. Finally, for each follow, we calculated the bearing (that is, the angle measured from the north) between the location of first acoustic contact and the last location, which provided a proxy for the direction of the whales' movements during that encounter. The resulting angles were plotted
and singletons. Finally, for each follow, we calculated the bearing (that is, the angle measured from the north) between the location of first acoustic contact and the last location, which provided a proxy for the direction of the whales' movements during that encounter. The resulting angles were plotted for all encounters, as well as separately for encounters with single animals and groups, using rose

3. Results

Over the course of 13 research seasons, 23,570 km were covered looking for sperm whales (11,099 km in the first study period, and 12,471 km in the second; Table 1). Sperm whales were encountered acoustically on 214 occasions (56 in the first study period, 158 in the second), for a total of 4,851 km travelled following animals.

3.1 Overall habitat use

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Multicollinearity was identified for slope and rugosity, SLA and ADT, SST slope gradient and the standard deviation of SST, as well as between the same environmental variables at different spatial or temporal scales. Comparison of models including each of the collinear variables in a pair led to an initial full model that included effort, year, depth (30 arc-sec), slope (2.5 arc-min), aspect (5 arc-min), TPI, monthly SST (4 km), SST deviation, SLA, and the standard deviation of monthly SST. The SALSA algorithm retained all variables as smooth terms, with the exception of TPI, which was retained as a linear term. The splines for all other variables were characterised by 4, 2, 2, 6, 5, and 2 internal knots, respectively. The ACF plot suggested that there was large autocorrelation in model residuals, and that the variable identifying searching transects and follows was suitable to separate blocks of correlated residuals. Once accounting for the observed degree of autocorrelation within blocks using a robust sandwich variance estimator, the Wald's tests indicated that only the relationships with effort, depth, aspect and year were significant (p <0.05). The probability of encountering sperm whales initially increased with increasing effort, but then stabilized at larger effort values (Fig. 2a), and there was a general increase in the probability of occurrence over the years (Fig. 2b). Moreover, occurrence increased in deeper waters, with a second peak around 800 m (Fig. 2c). Finally, sperm whales occurred with a lower probability where the slope was directed towards West-Northwest (Fig. 2e). SALSA also estimated that sperm whale probability of occurrence was associated with lower slope gradient, larger TPI, colder (and colder than the median) waters, lower SLA and larger SST variability (Fig. 2), but these relationships were not significant under the Wald's test. The confusion matrix suggested that the final model correctly predicted, on average, 70.8% of presence and absence location. The area under the ROC curve was 0.79, confirming a satisfactory goodness-of-fit.

Model predictions highlighted the areas south, east and north-east of the archipelago as being characterised by a higher probability of encountering the animals in the first period (Fig. 3). In the second period, the probability of sperm whale occurrence was overall greater, illustrated by the effect of year. This intensification also resulted in new areas emerging as relevant for sperm whales, such as the channel between the islands of Mallorca and Ibiza, and areas north of Mallorca and Menorca. For both study periods, the prediction maps showed some edge-effects, predicting high probabilities of sperm whale occurrence in regions of the study area that were poorly surveyed (due to encounters occurring in these low-effort regions, as well as values of the explanatory variables at the extremes of the observed range). Maps of the upper and lower confidence intervals of model predictions derived from the bootstrapping procedure highlighted a moderate degree of uncertainty in the predicted distribution, which reflects the consequences of the large autocorrelation in model residuals (Fig. A.2).

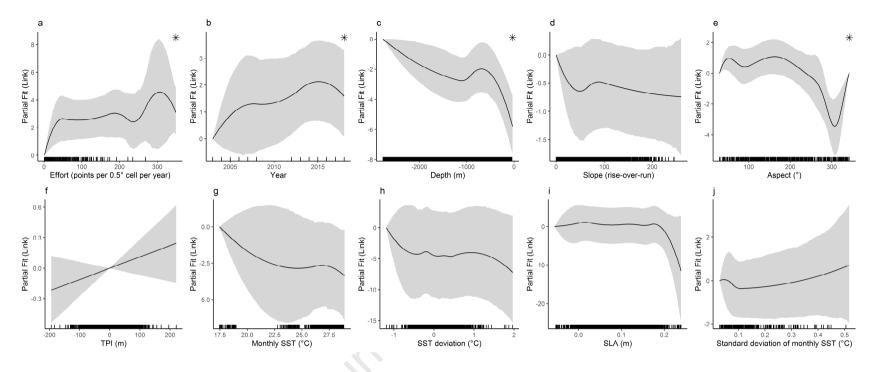


Figure 2. Estimated smooth relationships (on the link scale) between the probability of sperm whale occurrence and survey effort (a), year (b), depth (c), slope (d), aspect (e), topographic position index (TPI; f), monthly sea surface temperature (SST; g), SST deviation from the monthly median (h), sea level anomaly (SLA; i), and the standard deviation of monthly SST (j). Grey shaded areas represent robust, GEE-based 95% confidence intervals. A rug plot of the values of the covariates in the original data is shown at the bottom of each plot. Significant relationships are indicated with the star symbol (*) at the top right of the plot.

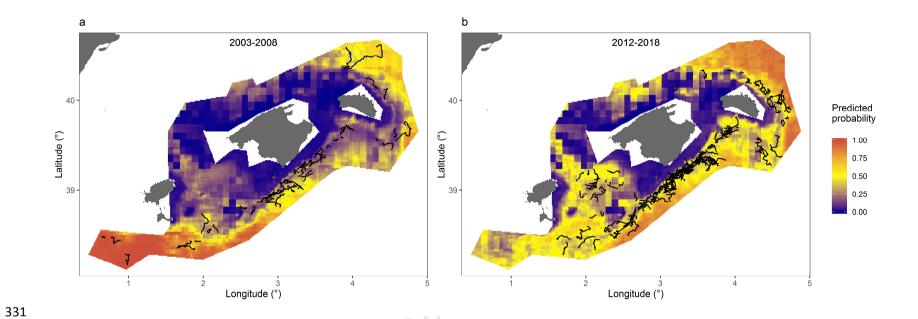


Figure 3. Predicted probability of sperm whale occurrence in the first (2003-2008; a) and second (2012-2018; b) study period, derived from the final model of overall habitat use. In black, the encounters with sperm whales in the corresponding study period.

3.2 Change in distribution in the two study periods

As for the overall habitat use model, the SALSA algorithm retained effort as a smooth term, highlighting an initial increase of the probability of encountering the animals for increasing effort, which stabilised for larger values (Fig. A.4). The CReSS bi-dimensional smooth was estimated to have 20 knots (chosen from a starting value of 12). The ACF plot confirmed the need to correct for autocorrelation within blocks, and the Wald's test suggested that effort, the CReSS surface, and the interaction between the CReSS surface and the study period were all significantly associated with sperm whale occurrence (p < 0.05). Therefore, model results suggested a significant change in geographic distribution between the two study periods. Spatial predictions mimicked the results of the overall habitat use model: in the second study period, animals were repeatedly encountered in the channel between Mallorca and Ibiza, and were also found to the north of the island of Menorca, overall occurring across a wider area (Fig. 4). The model correctly classified 68.9% of sperm whale presences and absences on average, a goodness-of-fit that was confirmed by the area under the ROC curve (0.76). As for the overall habitat use model, there was moderate uncertainty in the predicted distribution (Fig. A.3).

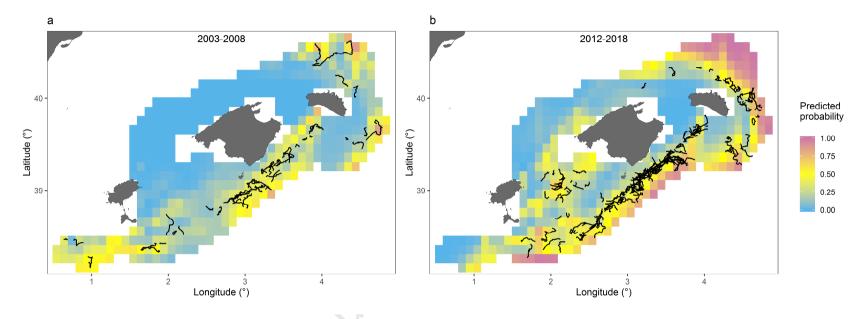


Figure 4. Predicted probability of sperm whale occurrence in the first (2003-2008; a) and second (2012-2018; b) study period, derived from the final model of geographic distribution. In black, the encounters with sperm whales in the corresponding study period.

3.3 Differences between groups and single animals

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The patterns of multicollinearity highlighted by the VIF and pairwise correlations were the same 354 as described for the overall habitat use model. The models for each collinear variable in a pair 355 suggested the full initial model should include effort, year, depth (2.5 arc-min), slope (10 arc-356 min), aspect (5 arc-min), TPI, 8-day SST (20 km), SST deviation, SLA, and the standard 357 358 deviation of 8-day SST. SALSA retained all covariates as smooth terms, selecting 4, 1, 3, 3, 3, 3, 359 3, 5, 1, and 3 internal knots, respectively. The relationships of group occurrence (versus single animals) with effort, depth, slope, year, 8-day SST, SST deviation and SLA were found to be 360 significant under the Wald's test using robust standard errors, corrected for the observed degree 361 of correlation within blocks. Specifically, groups were found in deeper, warmer waters (and 362 warmer than the monthly median), with two peaks in slope gradient, and in association with 363 smaller values of SLA (Fig. 5c, d, g-i). There was also a greater probability of occurrence of 364 365 groups in the first and last years of the survey period, while the relationship with effort showed a peak around high effort values, but was otherwise wiggly (Fig. 5a and b). Some of these 366 relationships had large confidence intervals and should therefore be interpreted with caution. The 367 SALSA algorithm also retained non-significant relationships of group occurrence with greater 368 TPI, slopes oriented towards east and south, and lower variability in SST (Fig. 5). The final 369 model showed high goodness-of-fit (83% of correct grouping classifications, on average, and 370 area under the ROC curve equal to 0.9). 371 372 The BA value between group and single animal encounters was 0.835, suggesting a high degree of spatial overlap. However, the distribution of BA values obtained from randomly reassigning 373 encounters to groups or singletons showed that the observed BA value was substantially lower 374

- than the null expectation (Fig. A.5), which suggests the existence of some geographic
- segregation between the two groupings.

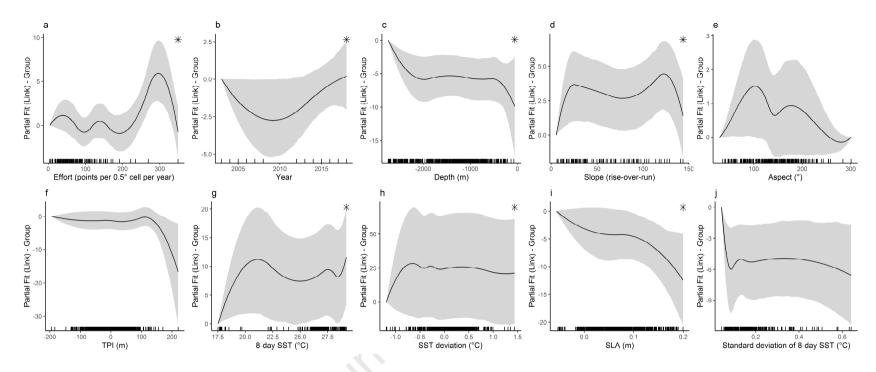
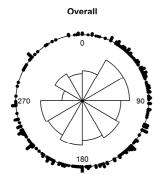
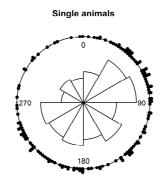


Figure 5. Estimated smooth relationships (on the link scale) between the probability of encountering sperm whale groups (as opposed to single animals) and survey effort (a), year (b), depth (c), slope (d), aspect (e), topographic position index (TPI; f), monthly sea surface temperature (SST; g), SST deviation from the monthly median (h), sea level anomaly (SLA; i), and the standard deviation of monthly SST (j). Grey shaded areas represent robust, GEE-based 95% confidence intervals. A rug plot of the values of the covariates in the original data is shown at the bottom of each plot. Significant relationships are indicated with the star symbol (*) at the top right of the plot.

The rose diagram for the approximate direction of the whales' movements over the course of each encounter showed some tendency to preferentially move towards north-east, south-west and south-east, with a lower occurrence of movements towards north and north-west (Fig. 6). These angles broadly reflect the direction of the continental slope south and east of the islands. The Rayleigh test suggested that observed angles were not uniformly distributed (p < 0.01). There were some subtle differences between groups and single animals (Fig. 6), which may reflect the greater occurrence of groups around the island of Menorca, where the continental slope has a different orientation.





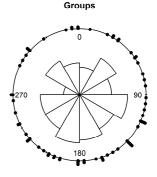


Figure 6. Rose diagrams of the bearing between the first and last location of each follow for all encounters, and separately for encounters with single animals and groups. Black dots on the plot margins represent the actual bearing values.

4. Discussion

We used acoustic monitoring data collected over 13 research seasons to characterise the summer habitat of sperm whales around the Balearic archipelago and its variation over the medium term. Even though estimated relationships with environmental proxy variables showed moderate levels

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of uncertainty and complexity, they offer the opportunity to generate hypotheses about the potential mechanisms that determine sperm whale distribution. In line with results from previous studies in this area, the Mediterranean Sea and globally, sperm whales were found to associate with bathymetric features, such as depth and continental slopes with specific orientation (Azzellino et al., 2012; Pirotta et al., 2011; Roberts et al., 2016; Rogan et al., 2017; Skov et al., 2008; Tepsich et al., 2014; Virgili et al., 2019; Waring et al., 2001; Whitehead, 2003). The coupling of these features with water circulation is known to promote local upwelling and increased productivity, which ultimately creates predictable feeding opportunities for top predators like sperm whales (Moors-Murphy, 2014). The preference for the habitat associated with the continental slope south and east of the archipelago was also supported by the broad orientation of the encounters highlighted by the rose diagrams, which indicated that whales tended to move along the direction of the slope. In addition, sperm whales were frequently encountered in the channel between Mallorca and Ibiza during the second study period, an area characterised by the presence of three seamounts (Aguilar et al., 2010). The importance of these submarine structures for sperm whales has been previously documented in other regions (Hann et al., 2016; Wong and Whitehead, 2014). In contrast, the characterisation of the dynamic processes that underpin sperm whale distribution in this area remains incomplete, although the final model provided some indication that colder waters, with lower sea level anomaly and larger temperature variability were preferred (Davis et al., 2002; Gannier and Praca, 2007; Virgili et al., 2019). These relationships were highly uncertain (and non-significant once accounting for residual autocorrelation), which could indicate a relative flexibility in habitat use within the broader bathymetric range that the animals appeared to select in this region. Our analyses also confirm the importance of a multi-scale approach for the evaluation of sperm whale habitat

(Jaquet, 1996), with different environmental characteristics being related to animal occurrence at
different spatial and temporal scales (e.g., Cotté et al., 2009; Jaquet and Whitehead, 1996; Pirotta
et al., 2014). The scale at which oceanographic processes operate and the temporal lags between
these processes and the concentration of sperm whale prey thus present an additional
complication to the functional description of the habitat of this species (Guisan and Thuiller,
2005), especially when the survey effort concentrates where expected encounter probability is
high, like in this case.
Even after accounting for the increasing effort over the research period, sperm whale occurrence
in the waters around the Balearic Islands was found to be overall increasing over time. This
finding supports the critical role of the area as a breeding and feeding ground for the Endangered
Mediterranean population (Rendell and Frantzis, 2016), and reaffirms its identification as an
Important Marine Mammal Area under the IUCN IMMA initiative (Corrigan et al., 2014). The
abundance and trend of the genetically isolated Mediterranean population is unknown, but
numbers are believed to be low (Rendell et al., 2014; Rendell and Frantzis, 2016). A local
increase in occurrence does not necessarily imply that the population as a whole is increasing,
but reinforces the need to protect sperm whales from recognised threats in this region where they
predictably and increasingly occur during the summer months, such as collisions with boats and
entanglement in drift nets, and to extend the research and monitoring effort to other periods of
the year. The frequent occurrence of calves within groups encountered around the islands (J.M.
Brotons, pers. obs.) highlights the importance of these conservation requirements.
Our analyses also highlighted a significant change in the overall geographic distribution of the
animals around the islands between the two study periods. Habitat characteristics and increased
occurrence over time partly explain this change in spatial distribution, as reflected by the ability

447	of the overall habitat model to capture the emergence of new areas of intense use (e.g. the
448	channel between Mallorca and Ibiza). Further investigation of the oceanography of the area
449	across the study period could elucidate some of these trends. However, this change, together with
450	the lack of clear relationships with dynamic, oceanographic variables, could be an additional
451	indication of the flexible nature of the distribution of the animals within the preferred
452	bathymetric range around the islands, particularly in a phase where local density is increasing.
453	This dynamism may imply that food resources are available across the whole area, and that
454	individuals can plastically adjust their habitat use in response to other intrinsic drivers. Marine
455	mammals have been shown to dynamically alter their habitat use over time, for example in
456	response to changes in density and as a result of social dynamics (Arso Civil et al., 2019; Cantor
457	et al., 2016; Carroll et al., 2014; Mobley et al., 1999).
458	In highly social species, such as the sperm whale, intrinsic biological factors strongly influence
459	the distribution of individuals in space and time (Cañadas and Hammond, 2008; Guisan and
460	Thuiller, 2005; Palacios et al., 2014). Interactions with members of the same group, sexual
461	segregation and the need to care for the young are expected to contribute to movement decisions.
462	Here, we highlighted that the habitat used by single animals (likely males) and groups (mostly
463	females and young) differed. Encounters with groups occurred in deeper waters, which is
464	consistent with existing evidence from other areas (Gregr and Trites, 2011), characterised by
465	specific slope gradients. Groups were also associated with warmer locations (and warmer than
466	the monthly median), which is in contrast with results of previous analyses on a subset of these
467	data (Pirotta et al., 2011). This contradiction may be partly reconciled by the fact that surface
468	temperature is highly correlated with month (see the corresponding rug plot in Fig. 2g and 5g),
469	and earlier or later months were only surveyed in more recent years; the relationship with

temperature (but not with SST deviation) may therefore mask a seasonal trend in the relative
occurrence of the two groupings. Finally, group occurrence was related to smaller values of
SLA, which could be associated with cyclonic circulation or confluence zones and higher
productivity (Davis et al., 2002), supporting the hypothesis that groups exploit better foraging
patches (Whitehead, 2003). These differences resulted in some degree of fine-scale spatial
segregation, as indicated by Bhattacharyya's affinity, reinforcing the findings of previous work
on the differential distribution of single males and social units in this and other areas (Gregr and
Trites, 2011; Jones et al., 2016; O'Hern and Biggs, 2009; Whitehead, 2003). Further clarifying
patterns of habitat segregation will allow quantifying differences in exposure rate and
susceptibility to anthropogenic stressors in the region. Being one of the few recognised breeding
ground for the population (Rendell and Frantzis, 2016), the risk of calves being struck by vessels
or separated from the females should be minimised, for example by delineating and protecting
areas specifically selected by groups (e.g., via restrictions on vessel speeds).
In general, the dynamic nature of sperm whale distribution in the area complicates management
efforts, because the whole bathymetric range may require some form of protection, while the
exact location of these mobile, social animals may be hard to identify at any moment in time.
The increasing use of the Mallorca channel is particularly worrying, as it exposes individuals to
the high levels of maritime traffic occurring between the islands (e.g. Fig. A.6). Quantifying the
distribution overlap and individual encounter rate with various human activities operating in the
region (e.g. via dedicated tagging studies) will therefore also be crucial (e.g., Pirotta et al., 2018)
Passive acoustics is an effective tool for the assessment of sperm whale distribution. While
detection range could vary depending on ambient noise levels and environmental conditions, we
do not expect systematic biases in certain areas or times. Animals could have also been missed

along the line if they were not vocalizing, although sperm whales spend 80% of their time foraging (Watwood et al., 2006), and in a joint visual and acoustic survey sperm whales were always detected acoustically first (Barlow and Taylor, 2005). Future research should continue monitoring the presence and habitat use of sperm whales in this important area. In light of the dynamic distribution highlighted by our results, other areas of the archipelago to which limited effort has been dedicated so far should be targeted by future surveys (e.g., the north of Ibiza and waters further to the north and south of the islands), which would also address some of the edge effects emerging in model predictions. Extending the effort to a wider region will also support additional multi-scale analyses of sperm whale habitat use, which could help clarify the underlying ecological processes. Moreover, given the differences between social groupings, the area offers the unique opportunity to investigate the mechanisms that underpin the social system of the species and how this influences distribution patterns; further studies of sex-specific diet and group-specific habitat use could provide additional evidence towards such understanding. More broadly, the trends in encounter rate and change in distribution in this region should be considered in the context of local variation in other key areas of the basin, to understand the wider dynamics of the population and design effective, integrated measures that can support its conservation.

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Acknowledgements

Funding: this work was supported by the National Geographic Society [grant number EC-53116R-18]. One World Wildlife, Whale and Dolphin Conservation, Mallorca Preservation Fund, Marilles Foundation, Obra Social La Caixa, and the Nando Peretti Foundation have also

515	supported the data collection with various grants over the years. We thank all the volunteers that
516	have helped with the data collection, and A. Cañadas (Alnilam, Spain) and C. Mundy (One
517	World Wildlife, UK) for their contribution to the initial design of the monitoring program. We
518	are especially grateful to the Swiss Cetacean Society for their continuous support to the Balearic
519	Sperm Whale Project. We are grateful to L. Scott-Hayward for her advice on the statistical
520	analysis. We thank Emer Rogan and University College Cork for providing desk space to EP.
521	Finally, we thank the Editor-in-Chief Prof. Imants Priede and one anonymous reviewer for their
522	useful comments on the manuscript.
523	
524	Data availability statement
525	The dataset and code to run the analysis are available on the Open Science Framework repository
526	(https://osf.io/x5afs/).
527	
528	Author contributions
529	L.E.R conceived the original study, which was further developed in collaboration with J.M.B.
530	and E.P.; L.E.R, J.M.B., M.C. and E.P. collected the data; E.P. conceived and carried out the
531	data analysis, with the help of S.B.; E.P. wrote the manuscript, and all authors contributed to
532	revisions.
533	
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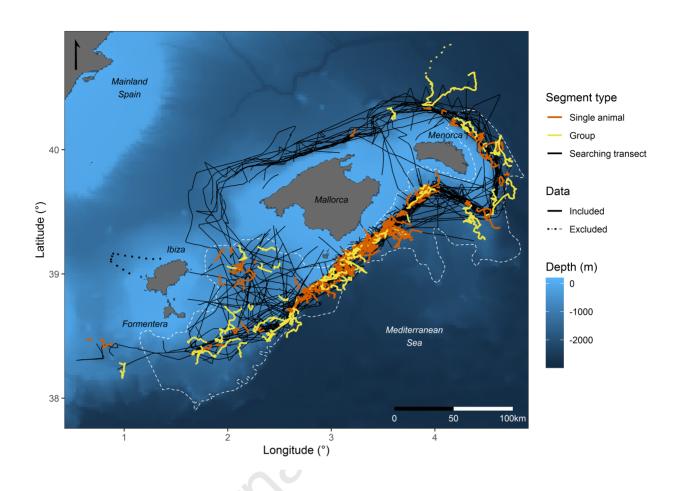
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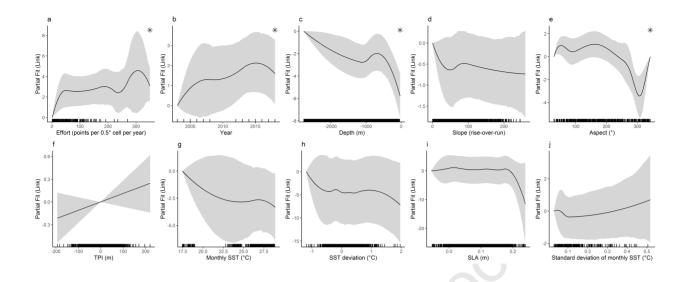
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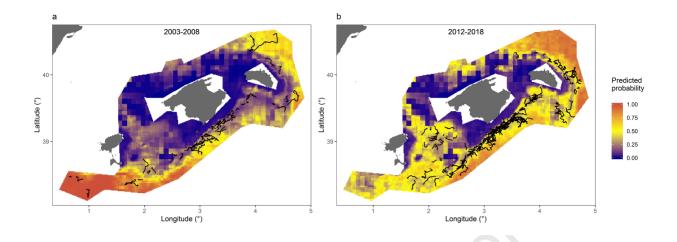
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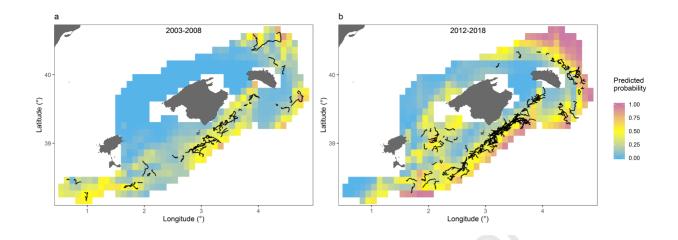
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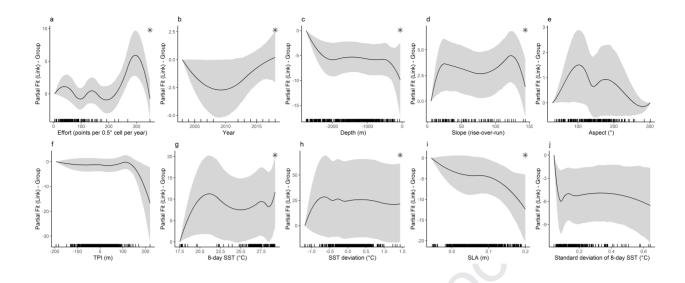
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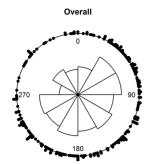


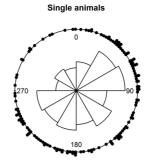


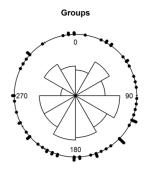












Highlights

- Marine megafauna habitat use is dynamic and affected by intrinsic biological factors
- We assess the habitat use of an endangered population of sperm whales over 15 years
- Local occurrence increased over time and distribution underwent a geographic shift
- Single animals and social groups used habitat with different characteristics
- Results support the design of targeted conservation measures in this important area

Declaration of interests
oxtimes The authors declare that they have no known competing financial interests or personal relationships that could have appeared to influence the work reported in this paper.
☐The authors declare the following financial interests/personal relationships which may be considered as potential competing interests: