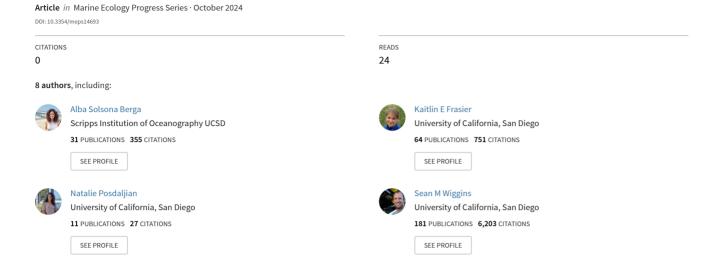
Accounting for sperm whale population demographics in density estimation using passive acoustic monitoring







Accounting for sperm whale population demographics in density estimation using passive acoustic monitoring

Alba Solsona-Berga^{1,*}, Kaitlin E. Frasier¹, Natalie Posdaljian¹, Simone Baumann-Pickering¹, Sean Wiggins¹, Melissa Soldevilla², Lance Garrison², John A. Hildebrand¹

¹Scripps Institution of Oceanography, University of California San Diego, La Jolla, CA 92037, USA ²NOAA Southeast Fisheries Science Center, Miami, FL 33149, USA

ABSTRACT: Sperm whales *Physeter macrocephalus* are highly sexually dimorphic, with adult males having larger bodies, more powerful echolocation clicks, and slower echolocation clicking rates compared to females. This study introduces methods for estimating sperm whale population densities in the Gulf of Mexico (GoMex) by accounting for the population demographics using passive acoustic monitoring and reveals that ignoring the differences between demographic segments can introduce bias in density estimates. Weekly densities were estimated per 3 demographic segments: social groups consisting of adult females and their offspring, mid-size animals, and adult males. Analysis revealed that the GoMex sperm whale population is primarily composed of social groups, which account for 92 to 98% of the overall population. Mid-size animals and adult males made up a small proportion of the population and were only intermittently present. Our 7 yr GoMex density estimates, including the 2010 *Deepwater Horizon* (DWH) oil spill period and subsequent years, revealed demographic-specific trends. Declines found at 2 north-central GoMex sites, coupled with increases at a southeastern site, may indicate population movements and potential impacts from the 2010 DWH oil spill and elevated noise levels from anthropogenic activities in the north-central GoMex.

KEY WORDS: Physeter macrocephalus \cdot Sperm whale \cdot Population demographics \cdot Density estimation \cdot Passive acoustic monitoring

1. INTRODUCTION

Human pressures, including underwater noise, ship collisions, fisheries interactions, pollution, and climate change, hinder the recovery of sperm whale *Physeter macrocephalus* populations (National Academies of Sciences, Engineering, and Medicine 2017, Sousa et al. 2019). Historical commercial whaling severely reduced global abundance by 57% (Whitehead & Shin 2022), leading to endangered status under the US Endangered Species Act, depleted under the US Marine Mammal Protection Act, and Vulnerable under the IUCN Red List of Threatened Species (Taylor et al.

2019). Sperm whales in the northern Gulf of Mexico (GoMex) are managed as a separate stock under the Marine Mammal Protection Act with an estimated population of 1180 animals and an average density of 1.7 animals 1000 km⁻² (2017–2018 period; Garrison et al. 2020, Hayes et al. 2022). While still recovering from historical commercial whaling (Townsend 1935, Reeves et al. 2011), this species faces contemporary challenges related to habitat degradation. The GoMex region experiences exceptionally high levels of noise pollution from seismic surveys for oil and gas exploration and trafficked shipping lanes (Wiggins et al. 2016, Estabrook et al. 2016). Additionally, the lingering ef-

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 $^{{\}tt ^{\star}Corresponding\ author:\ asolsonaberga@ucsd.edu}$

fects of the 2010 *Deepwater Horizon* (DWH) oil spill—the largest in US history (Ramseur 2010, Levy & Gopalakrishnan 2010) — pose a unique challenge to this population. The consequences of these stressors remain poorly understood (Farmer et al. 2018), highlighting the need to monitor changes and inform management and conservation efforts.

Reliable estimates of population abundance or density trends are crucial indicators of species' status in the wild but remain challenging, particularly for pelagic species like sperm whales. Passive acoustic monitoring (PAM) is an attractive approach for remote long-term data collection, population estimation (Frasier et al. 2016, von Benda-Beckmann et al. 2018), and trend analysis (Hildebrand et al. 2015). Assessment of animal density and abundance from autonomous PAM recorders relies on the detection and counting of individual or group vocalizations in a given area and time period (Margues et al. 2013). Animal vocalization rates, source levels, and group sizes are key scalar parameters required for accurate density estimation from such acoustic recorders (Marques et al. 2009, Hildebrand et al. 2015), and to date, these parameters are assumed to apply at the species or population level. These assumptions may introduce bias in density estimates, as sperm whales have complex population demographics including social and sexual maturity segregation across different latitudinal ranges for most populations (Best 1979, Lyrholm et al. 1999), as well as sexual dimorphism in body size, which has been linked to differences in echolocation click characteristics and diving behaviors (Gordon 1991, Watwood et al. 2006, Growcott et al. 2011, Solsona-Berga et al. 2022).

Adult females and immature animals form social groups in low and mid-latitudes (Best 1979, Rice 1989). In contrast, maturing males form bachelor groups of similar-aged animals and become increasingly solitary, moving to higher latitudes as they mature (Best 1979, Whitehead 2003). Males transit to lower-latitude breeding grounds, but the timing remains poorly understood (Rice 1989). The GoMex population is mostly comprised of adult females and immature animals, with smaller body and group sizes than in other ocean basins (Jaquet & Gendron 2009). Adult male movements and breeding times in this regional population are still unknown. The little knowledge we have, derived from studies of genetic diversity and occasional visual observations showing both sexes moving across the GoMex basin, with some males breeding in different ocean basins (Lyrholm et al. 1999, Alexander et al. 2016), is complemented by findings from a recent long-term PAM

study (Solsona-Berga et al. 2022). This study revealed spatial and temporal variability of the northern GoMex population demographics, including seasonal patterns and possible male presence year-round.

Adult males, which have larger bodies, produce more powerful echolocation clicks at a slower rate than adult females and juveniles (Goold & Jones 1995, Solsona-Berga et al. 2022). Larger heads of adult males may allow the buildup and discharge of greater volumes/pressures of air during sound production, resulting in higher source level emissions (Goold & Jones 1995). Differences in clicking rates between sexes may be related to the maximum detection range for prey (Jensen et al. 2018). With a larger echolocation range, adult males may slow down the interval between clicks to wait for more distant echoes. Differences in clicking rates, detectability, and group sizes among sex/age groups (Douglas et al. 2005) may significantly impact acoustic density estimation of sperm whales. Calculating demographic-specific acoustic density estimates could reduce error introduced by averaging parameters that are extremely different across demographic segments, and facilitate analyses of demographic-specific population trends.

Using long-term PAM, we present a framework for estimating demographic-specific sperm whale densities using 2 approaches, cue and group counting, accounting for differing clicking rates, group sizes, and detectability among demographic segments. Our 7 yr GoMex density estimates, including the 2010 DWH oil spill period and subsequent years, revealed demographic-specific trends. Declines found at 2 north-central GoMex sites and increases at a southeastern site could be linked to population movements, potential DWH oil spill impacts, and anthropogenic activities.

2. MATERIALS AND METHODS

Sperm whales were monitored at 3 northern GoMex locations during and following the DWH oil spill (2010–2017). These monitoring sites (Fig. 1) included one near Mississippi Canyon (MC) within 15 km of the DWH wellhead, another near Green Canyon (GC), located outside and northwest of the DWH surface oil footprint, and a third near Dry Tortugas (DT), outside and southeast of the oil footprint (Table S1 in the Supplement at www.int-res.com/articles/suppl/m746p121_supp.pdf).

At each site, a High-frequency Acoustic Recording Package (Wiggins & Hildebrand 2007) recorded

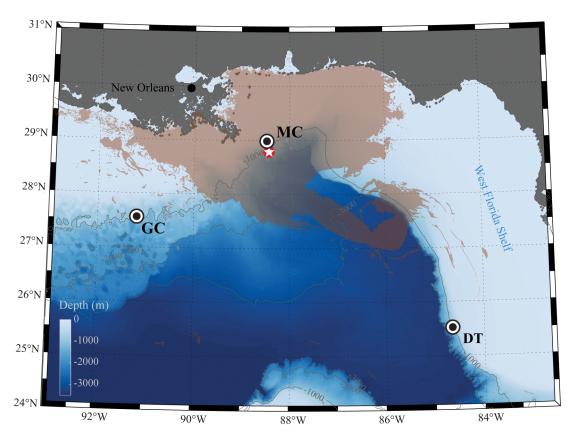


Fig. 1. Acoustic monitoring sites in the northern Gulf of Mexico from 2010 to 2017, named for nearby oceanographic features: Green Canyon (GC), Mississippi Canyon (MC), and Dry Tortugas (DT). *Deepwater Horizon* wellhead is shown by a star, with cumulative surface oil footprint (Kobara 2019) in brown. Bathymetric contours at 1000 m depth increments are illustrated

sound nearly continuously at a sampling rate of 200 kHz. Sperm whale echolocation clicks were automatically detected and classified for use as cues for estimating weekly densities. All analyses, including signal detection, classification, and subsequent density and trend analyses were carried out in MATLAB R2016b (Mathworks). Signal detection and classification was performed using an automated multi-step approach described by Solsona-Berga et al. (2022). This approach implemented a band-pass filter between 5 and 95 kHz to reduce background noise. Detections were filtered out at a peak-to-peak received sound pressure level (RL) of 135 dB $_{pp}$ re: 1 μ Pa to establish a consistent detection threshold. Other marine mammals like beaked whales or delphinids were automatically filtered out using spectral click characteristics such as peak frequency and spectral shape (Solsona-Berga et al. 2022). Times during ship passages were also excluded because sperm whale clicks were often indistinguishable from ship-related noise. An automated vessel detector, Triton Ship-Detector (Supplement of Solsona-Berga et al. 2020),

identified ship passages as times of increased noise in 3 specific frequency bands: 1-5, 5-10, and 10-50 kHz (see detector settings in https://github.com/ MarineBioAcousticsRC/Triton/wiki/Ship-Detector). Reduced recording effort resulting from the removal of these noise-dominated time periods was accounted for in density estimates by adjusting recording effort parameters to reflect the duration of data suitable for sperm whale detection. Sperm whale detections were grouped into encounters and manually validated in DetEdit software (Solsona-Berga et al. 2020) by 2 analysts (A.S.B., N.P.). While this process ensured classification accuracy of encounters, not every individual detection was evaluated. Sperm whale social vocalizations (codas, creaks, and slow clicks) and creaks for foraging were not included in this analysis. The high RL threshold of the detector prevented the detection of codas and creaks, which have lower source levels compared to regular echolocation clicks (usual clicks). Although slow clicks were detectable with these settings, they were not found during the Det-Edit validation process.

2.1. Population demographic classes

Sperm whale demographics were incorporated into density estimates to reduce potential sex/age class bias by considering variability in clicking rates, detectability, and group sizes among size classes. Echolocation click repetition rate, measured as the interclick interval (ICI), served as a proxy for body length to infer demographic structure (Solsona-Berga et al. 2022, Posdaljian et al. 2024, Westell et al. 2024). This method transformed ICIs into time series of modal ICI distributions within 5 min windows, associating dominant ICI patterns to body size classes. In this study, we used the same time series of size classes found for GoMex sperm whales by Solsona-Berga et al. (2022). Sperm whale detections were categorized into 3 age/ sex groups (referred to hereafter as demographic classes) in 5 min bins based on ICI distributions: (1) small animals (<12 m) with ICIs of 0.44-0.70 s, presumed to be social groups of adult females and their offspring; (2) mid-size animals (11-13 m) with ICIs of 0.70-0.85 s, presumed to be adult females or subadult males; and (3) large animals (13-15 m) with ICIs of 0.85-1.0 s, presumed to be adult males (Solsona-Berga et al. 2022).

2.2. Density estimation

Animal densities were estimated using 2 point-transect distance sampling approaches: cue and group counting (Buckland et al. 2001, Buckland 2006). Cue counting used individual click detections as the fundamental unit for density estimation. Group counting used presence or absence of click detections in discrete time windows as the unit of analysis (e.g. Hildebrand et al. 2015). Both approaches require knowledge of detector performance, cue properties, and cue detectability around the sensor (Marques et al. 2009).

Cue counting converted the number of detected clicks, n_{kt} , at site k during week t into animal density \hat{D}_{kt} , based on an estimator of animal density using acoustic cues recorded from fixed single sensors (Marques et al. 2009):

$$\hat{D}_{kt} = \frac{n_{kt} \left(1 - \hat{c}_{pk} \right) \left(1 + \hat{c}_{nk} \right)}{\pi \, w^2 \, \hat{P}_k \, T_{kt} \, \hat{r}_k} \tag{1}$$

where \hat{c}_{pk} is the proportion of false positive detections at site k, \hat{c}_{nk} is the proportion of false negative detections at site k, \hat{P}_k is the probability of detecting a click within a horizontal radius of size w (beyond which no detections are assumed possible) at site k,

 T_{kt} represents the total time monitored at site k during week t, and \hat{r}_k is the estimated click production rate at site k.

Group counting required detection of animal presence within time windows (Hildebrand et al. 2015), along with knowledge of the detectability of clicks produced by the observed group of animals. It relies on knowledge of group size and group vocalization behavior. Using group counting, the estimated density D_{kt} at site k, during week t is:

$$\hat{D}_{kt} = \frac{n_{kt} \left(1 - \hat{c}_{pk} \right) \left(1 + \hat{c}_{nk} \right) \hat{s}}{\pi w^2 \hat{P}_k \hat{P}_v T_{kt}}$$
(2)

where n_{kt} , c_{pk} , c_{nk} , and P_k are similar to those in Eq. (1) but refer here in particular to time bins (5 min) with group detections rather than individual click detections. \hat{s} represents the mean group size, and \hat{P}_v is the probability of a group being vocally active in a bin.

The coefficient of variation of estimated densities, $CV(D_{kt})$, was obtained using the delta method approximation (Seber 1982), to incorporate uncertainty in detection performance, probability of detection, and click rate for cue counting:

$$CV(\hat{D}_{kt}) = \left\{ \hat{D}_{kt}^2 \times \left[CV(\hat{c}_{pk})^2 + CV(\hat{c}_{nk})^2 + CV(\hat{P}_k)^2 + CV(\hat{r}_k)^2 \right] \right\}^{1/2}$$
(3)

and uncertainty in group size and vocalization probabilities for group counting:

$$CV(\hat{D}_{kt}) = \left\{ \hat{D}_{kt}^2 \times \left[CV(\hat{c}_{pk})^2 + CV(\hat{c}_{nk})^2 + CV(\hat{p}_k)^2 + CV(\hat{s})^2 + CV(\hat{p}_v)^2 \right] \right\}^{1/2}$$
(4)

Weekly densities were estimated per demographic class identified in the data set, considering density parameters specific to each class (social groups, mid-size animals, and adult males). Inclusive densities, estimated as the summed densities of the different population segments accounting for demographic differences (e.g. click production rate, group size, probability of detection, probability of vocal activity), were compared with generalized densities, which were estimated without consideration of demographic differences.

2.2.1. Detections and detector performance

To estimate the total number of detected clicks per demographic class each week (n_{kt}) , the class given to

each time bin was applied to all detected clicks within that bin. If a bin was assigned to more than 1 class (n =280 216; 0.3% of the total bins with sperm whale presence), the clicks in that bin were proportionally assigned a class based on the proportion of classes for that week. Some bins were not candidates for class categorization due to low numbers of clicks or lack of neighboring time bins to inform categorization (less than 9% of bins). To estimate the proportion of false positive clicks at each site (\hat{c}_{pk}) , a random subsample of clicks (n = 11000) that was evenly distributed across the entire data set was selected and assessed using DetEdit's evaluation tool (Solsona-Berga et al. 2020). The proportion of false positive 5 min bins (\hat{c}_{nk}) was calculated by evaluating the bin containing the randomly selected clicks. False negatives (\hat{c}_{nk}) were estimated using a simulation method incorporating the distribution of received click amplitudes, including those that did not pass the criteria of the detector spectral click characteristic, to account for all missed detections of clicks with RLs \geq 135 dB_{pp} re: 1 μ Pa (described in Section 2.2.2 below).

2.2.2. Detection probability

A Monte Carlo simulation estimated the probability of detecting an individual click or any clicks from a group of echolocating animals in 5 min bins (\hat{P}_k) over a horizontal distance range around each sensor (Frasier et al. 2016). This simulation aimed to determine received levels of expected clicks, accounting for signal characteristics, diving behavior, acoustic environmental propagation, and detector performance. This simulation method selected random samples (over 500 iterations) from a uniform distribution of animal positions and orientations ($n = 100\,000$ iteration⁻¹) with respect to the hydrophone (Table 1). A mean detection probability and its variance was computed as the ratio of detected clicks or groups to the total number of simulations (Frasier et al. 2016). Unlike previously implemented Monte Carlo simulations from fixed single sensors (Hildebrand et al. 2015, 2019, Frasier et al. 2016), this approach used: (1) a bimodal diving pattern (Fig. S1) rather than one unique diving pattern to resemble tag data diving behavior from GoMex sperm whales, where animals featured search and foraging acoustic behavior in the mid-water column and seafloor transits referred here as benthic dives (Mate et al. 2017), and (2) a 3-dimensional forward-beam pattern which was allowed to vary in the backward direction based on a circular piston model, as shown in sperm whales when echolocating (Zimmer et al. 2005).

To estimate demographic-specific detection probabilities, 6 scenarios were simulated per site (1 for each of the 3 demographic classes for both the cue and group counting methods) to account for differences in echolocation click characteristics and dive behavior based on body size (Gordon 1991, Watwood et al. 2006, Growcott et al. 2011, Solsona-Berga et al. 2022) and the different bathymetric features and depths of each site (Fig. 1; Table S1). Click peak frequencies were obtained from the data for each site and class (Table 1). Frequency-dependent propagation loss volumes at each site at the typical peak frequency per class were simulated using the beamtracing algorithm 'Bellhop' (Porter & Bucker 1987), with site-specific environmental and physical parameters (see settings, databases, and versions in Frasier et al. 2016) drawn from the Oceanographic and Atmospheric Master Library (OAML) using ESME Workbench (Mountain et al. 2012). Simulations used previously published diving behavior parameters of the GoMex population (e.g. start clicking depth and foraging phase body angle; Table 1); however, dive parameters have only been documented for midwater column dives with unknown relative bottom depth (Watwood et al. 2006). Proportion of time spent in each bimodal dive is unknown for the GoMex population, as are the female source levels and beam patterns, which have only been measured for adult males in the Mediterranean, Bahamas, and Norway (Møhl et al. 2003, Zimmer et al. 2005, Nosal & Frazer 2007). For the unknown parameters for specific demographics or only documented for other populations, a grid-search was used to obtain a distribution of values from which to sample the mean and variability for the simulations (Hildebrand et al. 2019).

The grid-search approach, as described by Hildebrand et al. (2019), optimized parameter values by assessing a range of potential values, pre-selected based on literature for other populations or demographics (Table S2). Simulations were run with all other parameters held constant while the distribution of the parameter of interest was inspected for best goodness-of-fit by comparing model results with actual data. All possible combinations of parameters for optimization within the pre-selected ranges were tested. Optimized parameter values best fitting the model output were determined by minimizing the sum of the squared misfit for clicks or bins of any amplitude with RLs ≥ 135 dB_{DD} re: 1 μ Pa (Table 1). Final values were used in 6 scenario simulations per site to estimate the probability of detection (\hat{P}_k) . As expected, more clicks were detected at lower received levels as the distance from the hydrophone

Table 1. Monte Carlo detectability simulation parameters, along with their modeled sensitivity. Unique demographic class, site, or counting method parameters are listed separately. Cue counting uses individual click detections, while group counting uses 5 min time bins with group detections as the unit of analysis. RL: received sound pressure level; TL: transmission loss; MC: Mississippi Canyon; GC: Green Canyon; DT: Dry Tortugas

Parameter		Distribution sample mean	SD	Reference	\hat{P}_k cue sensitivity	\hat{P}_k group sensitivity
Max. detection range (km) Min. click RL (dB_{pp})		35 135	_ _	_ _		
Signal characteristics (demog	raphic-specif	ic)				
Source level (dB _{pp})	Social group Mid-size Adult male	237 ± 5^{a} 238 ± 5^{a} 242 ± 5^{a}	2-3	Grid search (Møhl et al. 2003, Zimmer et al. 2005)	$17\% dB^{-1}$ $19\% dB^{-1}$ $17\% dB^{-1}$	$6\% dB^{-1}$ $5\% dB^{-1}$ $6\% dB^{-1}$
Peak frequency (kHz)	Social group Mid-size Adult male	10 9 9	_	Data	_ _ _	_ _ _
Directivity (dB)	Social group Mid-size Adult male	30.5 ± 2.5 32.5 ± 2.5 32.5 ± 2.5	_	Grid search (Møhl et al. 2003, Zimmer et al. 2005, Nosal & Frazer 2007)	$-9\% dB^{-1}$ $-4\% dB^{-1}$ $-5\% dB^{-1}$	_ _ _
90° off-axis TL (dB)		37.5 ± 2.5	_	Zimmer et al. (2005), Nosal & Frazer (2007)	$-5\% \text{ dB}^{-1}$	_
180° off-axis TL (dB)		27.5 ± 2.5	_	Zimmer et al. (2005)	$-5\% \text{ dB}^{-1}$	_
Minimum off-axis TL (dB)		35.5 ± 2.5	2-3	Grid search (Zimmer et al. 2005)	_	$-1\% dB^{-1}$
Diving behavior (site-specific)					
Water column dive altitude (m		450 ± 50 650 ± 50 850 ± 50	10-30	Grid search (Watwood et al. 2006)	$0\% 50 m^{-1}$ $0\% 50 m^{-1}$ $2\% 50 m^{-1}$	$-3\% 50 \text{m}^{-1}$ $-2\% 50 \text{m}^{-1}$ $-1\% 50 \text{m}^{-1}$
Water column dive fraction (%) MC GC DT	55 ± 5 55 ± 5 95 ± 5	1-2	Grid search (Watwood et al. 2006)	$6\% 10\%^{-1}$ $3\% 10\%^{-1}$ $4\% 10\%^{-1}$	$-3\% 10\%^{-1}$ $-5\% 10\%^{-1}$ $-1\% 10\%^{-1}$
Benthic dive altitude (m)		10	10-30	Mate et al. (2017)	_	_
Benthic dive fraction (%)	MC GC DT	45 ± 5 45 ± 5 5 ± 5	1-2	Grid search (Mate et al. 2017, Irvine et al. 2017)	$-3\% 10\%^{-1}$ $-4\% 10\%^{-1}$ $-5\% 10\%^{-1}$	5% 10% ⁻¹ 4% 10% ⁻¹ 3% 10% ⁻¹
Start clicking depth (m)		210 ± 10	65-75	Watwood et al. (2006)	$1\% 50 m^{-1}$	$0\% 50m^{-1}$
Foraging phase	Cue	$0_{\rm p}$	0-50	Watwood et al. (2006)	$1\% \mathrm{deg^{-1}}$	_
Vertical body angle (deg)	Group	$0_{\rm p}$	0-65	Watwood et al. (2006)	_	$1\% {\rm deg^{-1}}$
Foraging phase horizontal bod angle (deg)	ly	0 ± 360	-	_	_	_

^aSite DT estimated source levels were 2 dB lower than the other sites (social group: $235 \pm 5 \, dB_{pp}$, mid-size animals: $236 \pm 5 \, dB_{pp}$, and adult males: $240 \pm 5 \, dB_{pp}$)

increased, due to the expanded observable area. The misfit of the number of clicks at lower received levels between the model and the actual data suggested missed detections, providing an opportunity to quantify them. False negatives (\hat{c}_{nk}) were estimated as the sum of the misfit between the final model and the measured data for the received level clicks or bins immediately above the threshold (135–140 dB_{pp} re: 1 μ Pa).

2.2.3. Vocal activity

Click production rate (\hat{r}_k) , which is required for cue counting, may vary among regions, locations, age groups, and sexes (Table S3). To account for this variability, click production rate was calculated as:

$$\hat{r}_k = \frac{\hat{P}_{cyc}}{\sum \widehat{\text{ICI}}_k} \tag{5}$$

^bNormal distribution, left truncated at 0 degrees

where \hat{P}_{cyc} is the mean proportion of time spent clicking, and \widehat{ICI}_k is the ICI at each site k, and of each demographic class when estimating demographicspecific estimates. Proportion of time spent clicking (\hat{P}_{cvc}) was computed using behavioral data reported from acoustic tag records in the GoMex (Watwood et al. 2006), multiplying the percentage of time spent in search phase producing regular clicks during a foraging dive (81 \pm 5% SD) by the total time of a foraging dive, including resting and socializing periods at the surface without echolocation (72 \pm 33%). Tag data in the GoMex only included female and immature animals, but populations in sub-tropical and temperate latitudes, including both sexes, spend equivalent amounts of time clicking (Watwood et al. 2006, Ward et al. 2012), hence the same mean proportion of time clicking was used for all classes.

ICI was computed per class and site using exclusive clicks from single demographic bins. ICI distribution per class was fit with a Gaussian distribution to estimate mean and coefficient of variation. A relatively constant click rate is maintained during a dive (Madsen et al. 2002b, Zimmer et al. 2003, Douglas et al. 2005) but variability exists (e.g. during terminal buzzes, descent, and ascent phases; Madsen et al. 2002a,b, Thode et al. 2002, Teloni et al. 2008). Only dominant click rate per class was estimated using this approach, restricted between 0.3 and 1.4 s based on the demographic classification method, as not all dive variability could be equally quantified with these data. This restriction excluded the tail of the distributions, acknowledging the limitations in accounting for all dive variability.

Assuming that the probability of detecting a group of animals during a given time window increases with group size and the synchrony in their clicking behavior (i.e. the extent to which the click sequences of animals within the group overlap), we estimated the required probability of a group being vocally active (\hat{P}_v) for group counting using the method of Hildebrand et al. (2019):

$$P_{v} = \left\{ \hat{P}_{cyc} \times [s - (s - 1) \times o] \right\} \tag{6}$$

where \hat{P}_{cyc} is the mean proportion of time spent clicking by one animal, s is the group size, and o is the synchrony of clicks among group members in a bin. Each animal added to the group is assumed to contribute both overlapped (synchronous) and non-overlapped (asynchronous) echolocation time to the bin. Adult females and their young form social groups with an average of 6.1 ± 4.8 animals in the GoMex (Jochens et al. 2008). Group sizes of adult males in the GoMex remain unknown but were expected to be 1-2 animals, con-

sidering their typical solitary behavior and limited interactions with social groups (Best 1979, Gero et al. 2014, Cantor et al. 2019). Sub-adult males form bachelor groups in which the group size reduces with age (Gaskin 1970, Ohsumi 1971, Best 1979), and aggregation of males have been documented in various latitudes dispersed over ~10—30 km (e.g. Gillespie 1979, Leaper & Scheidat 1998). An ad hoc test directly counting the number of interleaved received click sequences as a proxy for group size when animals were detected near the sensor (Hildebrand et al. 2015) supported the assumptions of group sizes of social groups and adult males (Text S1), and group sizes of 1—2 mid-size animals were estimated to be detected simultaneously, similar to adult males.

Three GoMex sperm whales from the same group were simultaneously tagged in a previous study (Jochens et al. 2008). This offered a rare opportunity to estimate synchrony of clicks among group members. Synchrony (o) and variation were estimated by measuring proportion of time, defined in 5 min bins, during which each pairwise combination of the 3 animals overlapped (Fig. S2). This showed a synchrony of $77 \pm 4\%$ among pairs of whales within the group.

2.3. Long-term trend estimation

Site trends were estimated by removing seasonal variability from weekly densities to characterize trends independently of the significant seasonality present among demographic classes of the population in the northern GoMex (Solsona-Berga et al. 2022). Seasonality was removed using a monthly seasonal pattern decomposition procedure (Cleveland et al. 1990), and linear trend and rate of change per 1000 km² yr⁻¹ was estimated using the Theil-Sen slope estimator (Sen 1968). This estimator computed the slope between each pair of points in a time series, using the true time difference, and estimated the overall slope as the mean across all pairs. This ensured a robust fit insensitive to outliers and shortterm inter-annual variability. Means and associated 95% confidence intervals were generated using a parametric bootstrap with 100 iterations, each with 500 points selected to compute the associated slope.

3. RESULTS

Sperm whales were present year-round in 31% of 5 min bins at MC and 14% at GC over a cumulative monitoring effort of 5.4 yr per site, after removal of ship

noise-dominated time periods. Cumulative monitoring effort at the southern site (DT) was 4.7 yr, and sperm whales were detected in 5% of bins during that period (Table 2). Ship passage periods removed from detection effort varied by site and over time (Fig. S3). Northern sites had the most ship passages, with MC having the highest noise levels during the oil spill (April—August 2010) with more than 20 h wk⁻¹ removed. In subsequent years, less than 10 h wk⁻¹ were removed at this site. GC had its peak noise level between 2012 and 2016, removing 20 h wk⁻¹, compared to 5 h wk⁻¹ removed at other times. The southeastern site (DT) had the fewest ship passages, removing less than 5 h wk⁻¹, with 2 short periods in 2012 and 2016 reaching 20 h wk⁻¹.

Most bins with sperm whale presence were assigned into 1 or more demographic classes (97% at MC, 96% at GC, 91% at DT; Table 2). All 3 classes were detected across all sites, with putative social groups being the most common (92% of the positive detection bins at MC, 87% at GC, and 74% at DT). Mid-size animals were moderately common (4% of detection bins at MC, 7% at GC, 14% at DT), and adult males were rare (1% of detection bins at MC, 3% at GC, 4% at DT). Presence of 2 classes in a bin were infrequent (0.3% of detections bins at MC, 0.1% at GC, 0.075% at DT), and cases where all 3 classes were detected simultaneously occurred only in 1 bin at MC (Table 2).

3.1. Demographic class differences in ICI and peak frequency

ICI varied by recording site within each class (Fig. 2, Table 3). ICI for social groups was consistent between the northern sites, and slightly larger at DT, resulting in an overall mean ICI of 482 ms with low

variability between sites (SD = 55 ms, CV = 11%). Mid-size animals had consistent ICI between sites, with an overall mean ICI of 670 ms and small variability (SD = 63 ms, CV = 9%), while adult males exhibited greater variability between the northern sites, resulting in an overall mean ICI of 812 ms with slightly higher but still low variability between sites (SD = 96 ms, CV = 12%). Click rates were estimated using the ICI per class and site (Table 4), which resulted in a higher clicking rate for social groups than for adult males and mid-size animals.

Click peak frequencies varied by class, with social groups having higher peak frequencies than adult males or mid-size animals (Fig. 3). Mean peak frequencies of social groups were 10.2–10.5 kHz, those of mid-size animals were 9.0–9.4 kHz, and those of adult males were 8.8–9.8 kHz (Table 3). Estimation of mean peak frequencies at MC was complicated by a bimodal distribution (Fig. 3), with a low-frequency peak at ~7 kHz in addition to the primary peaks at >~10 kHz (social groups) and ~9–10 kHz (mid-size animals).

3.2. Grid-search parameter optimization

The grid-search approximated sperm whale diving behavior parameters and signal characteristics unknown for a particular demographic or only known from other populations. Table 1 contains the values that provided the best fit between modeled received levels and observed distributions per site and class for both cue and group counting (Fig. 4). Based on the grid-search optimization for signal characteristics, estimated echolocation source levels and directivity varied by class but not between sites. The one excep-

Table 2. Number of 5 min bins assigned to a single demographic class, to multiple classes, and where assignment was not possible per site. Total bins with sperm whale presence and monitoring effort as number of bins are indicated in *italics*. The percentage of relative presence of each class is reported by the total bins with positive presence at each site. MC: Mississippi Canyon; GC: Green Canyon; DT: Dry Tortugas

Bin category	MO	C	—— GC	:	——DT ——		
J ,	Bins (n)	%	Bins (n)	%	Bins (n)	%	
Social group	162595	91.8	68471	86.5	17691	74.1	
Mid-size animals	6687	3.8	5501	6.9	3249	13.6	
Adult males	1499	8.0	2129	2.7	850	3.6	
Social group and mid-size	417	0.2	54	0.1	14	0.1	
Social group and adult males	144	0.1	34	< 0.1	2	< 0.1	
Mid-size and adult males	45	< 0.1	2	< 0.1	0	< 0.1	
Social group and mid-size and adult males	1	< 0.1	0	< 0.1	0	< 0.1	
Unassigned	5775	3.3	2992	3.8	2064	8.6	
Total bins positive presence	177163		79183		23870		
Total effort bins	563593		562812		489990		

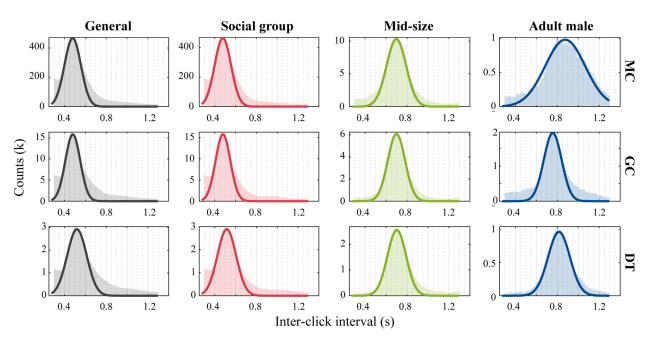


Fig. 2. Distribution of inter-click intervals (ICIs) for sperm whales by site, demographic class, and in general (all detections regardless of class), based on a histogram (shaded area) of ICIs with a bin width of 10 ms. A Gaussian distribution (line) omitting long tails was fit to obtain the mean and coefficient of variation. Refer to Fig. 1 for site abbreviations

Table 3. Inter-click interval (ICI) and peak frequency for sperm whale demographic classes per site. Mean and coefficient of variation (CV) of Gaussian fit for ICI and peak frequency distribution were estimated per class using only data from bins with a single class. MC: Mississippi Canyon; GC: Green Canyon; DT: Dry Tortugas

Site	Social group Mean CV	Mid-size Mean CV	Adult males Mean CV	Generalized Mean CV
ICI (n	ıs)			
MC	478 0.117	698 0.095	876 0.154	483 0.116
GC	479 0.104	699 0.083	757 0.079	481 0.107
DT	515 0.124	702 0.090	817 0.091	524 0.127
Peak	frequency (kHz)			
MC	10.2 0.106	9.0 0.097	9.8 0.101	10.1 0.111
GC	10.4 0.083	9.4 0.077	8.8 0.074	10.3 0.090
DT	10.5 0.110	9.3 0.088	9.4 0.083	10.2 0.120

tion was DT, where best fits resulted from source levels 2 dB lower for all classes (Table 1). Estimated source levels for social groups (237 \pm 5 dB $_{pp}$ re: 1 μPa @ 1m) and mid-size animals (238 \pm 5 dB $_{pp}$ re: 1 μPa @ 1m) were similar but lower than those estimated for adult males (242 \pm 5 dB $_{pp}$ re: 1 μPa @ 1m). Estimated directivity was the same for mid-size animals and adult males (32.5 \pm 2.5 dB) and lower for social groups (30.5 \pm 2.5 dB).

Optimized diving behavior parameters using gridsearch resulted in different proportions of time spent on mid-water or benthic dives, as well as dive altitude across all sites (Table 1). Estimated midwater column dive altitude was different at each site (MC = 450 ± 50 m; GC = 650 ± 50 m; DT = 850 ± 50 m). However, based on the varying bathymetric depths at each site, these altitudes suggest that the mid-water-column foraging depth was consistently near 500 m at all sites. A bimodal dive pattern with similar proportion of time spent in the midwater ($55 \pm 5\%$) and benthic dives ($45 \pm 5\%$) was estimated at MC and GC. In contrast, foraging dives at DT were almost exclusively estimated to occur in mid-water ($95 \pm 5\%$).

Received click levels predicted by the final simulations for the 3 classes

and sites were in good agreement with measured received levels in the range of 135–160 dB $_{pp}$ re: 1 μPa (Fig. 4). Group counting, however, resulted in a lower overall fit. Above 160 dB $_{pp}$ re: 1 μPa , fewer detections were measured than predicted using both counting methods for all classes and sites, likely as a result of too few recorded high-amplitude clicks. Near the detection received level threshold, fewer detections were measured than predicted, particularly using group counting. Greater numbers of low received level clicks are always expected relative to higher received level clicks because of a larger area for animals

Table 4. Sperm whale weekly mean densities and coefficient of variation (CV) derived from cue counting by site and demographic class. A maximum horizontal detection range (w) of 35 km was used to estimate densities. Estimated density trends regardless of class are referred to as generalized. MC: Mississippi Canyon; GC: Green Canyon; DT: Dry Tortugas

Class	Site	\hat{D}_{kt} Mean density (animals $1000~{ m km}^{-2}$)	$\stackrel{ ext{CV}}{(\hat{D}_{kt})}$	$n_{kt} T_{kt}^{-1}$ (n s ⁻¹)	\hat{c}_{pk} False positive (% clicks)	$ ext{CV} (\hat{c}_{pk})$	\hat{c}_{nk} False negative (% clicks)	$ ext{CV}$ (\hat{c}_{nk}) rate	\hat{r}_k Click (n sec ⁻¹)	$\operatorname{CV}_{(\hat{r}_k)}$	\hat{P}_k Probability of detection	$\begin{array}{c} \mathrm{CV} \\ (\hat{P}_k) \end{array}$
Demographic	Demographic-specific											
Social group	MC	2.259	0.014	0.098	5.1	0.02	2.8	0.01	1.220	0.19	0.009	0.022
	GC	0.825	0.018	0.033	2.7	0.03	2.2	0.01	1.218	0.19	0.008	0.023
	DT	0.345	0.030	0.009	8.2	0.07	0.1	0.01	1.132	0.19	0.006	0.025
Mid-size	MC	0.048	0.029	0.002	5.1	0.02	2.7	0.01	0.836	0.19	0.012	0.022
	GC	0.032	0.029	0.001	2.7	0.03	5.7	0.01	0.834	0.19	0.011	0.023
	DT	0.023	0.042	0.001	8.2	0.07	1.3	0.01	0.831	0.19	0.008	0.026
Adult male	MC	0.007	0.030	0.0004	5.1	0.02	0.0	0.01	0.666	0.19	0.022	0.020
	GC	0.009	0.135	0.0005	2.7	0.03	4.0	0.01	0.770	0.19	0.020	0.020
	DT	0.006	0.036	0.0002	8.2	0.07	0.0	0.01	0.714	0.19	0.015	0.023
Generalized												
	MC	2.319	0.014	0.100	5.1	0.02	2.8	0.01	1.208	0.19	0.009	0.022
	GC	0.868	0.017	0.034	2.7	0.03	2.2	0.01	1.213	0.19	0.008	0.023
	DT	0.379	0.028	0.010	8.2	0.07	0.1	0.01	1.113	0.19	0.006	0.025

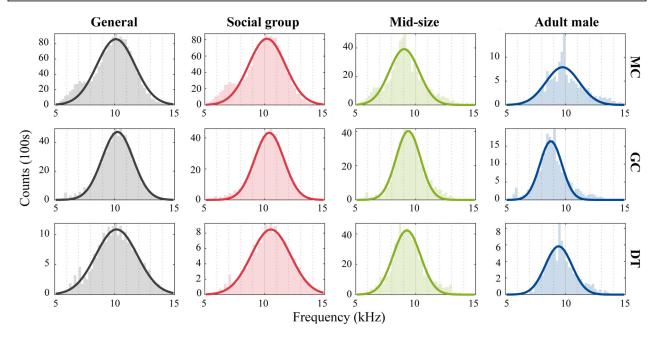
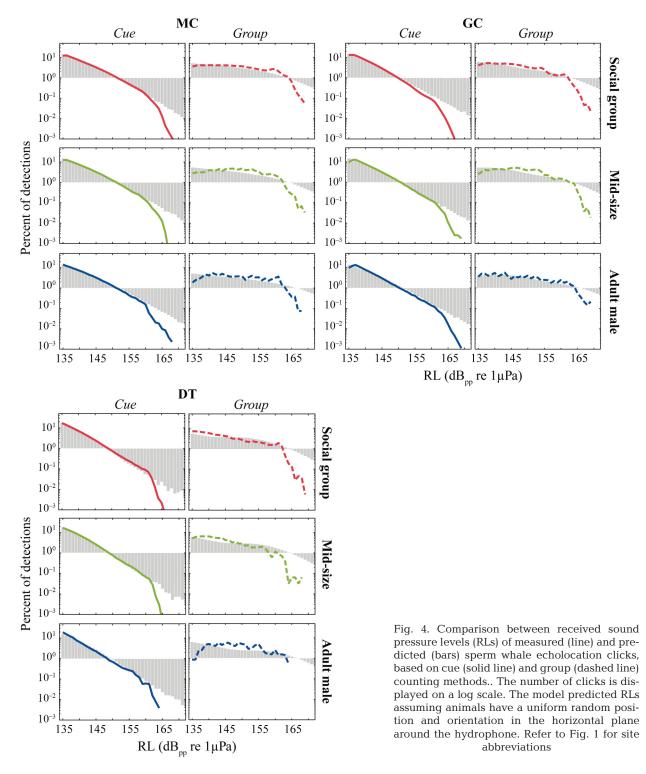


Fig. 3. Distribution of peak frequencies for sperm whale clicks by site, demographic class, and in general (all detections regardless of class). A Gaussian distribution (line), omitting long tails, was fit to obtain the mean and coefficient of variation. Refer to Fig. 1 for site abbreviations

to occupy as distance from the sensor increases. Since simulations selected random samples from a uniform distribution of animal positions and orientations in the horizontal plane around the hydrophone, the models indicate that groups were missed near the detection received level threshold, particularly for mid-size animals and adult males (Fig. 4, see Tables 4 & 5).

3.3. Detection probability

Probability of detecting sperm whales over a horizontal range (35 km) varied between demographic class and site for both counting methods (Fig. 5). Adult males were detectable at greater distances than social groups and mid-size animals, influenced main-



ly by higher mean source levels. A 1 dB $_{pp}$ re: 1 μ Pa increase in mean source level led to a 17–19% increase in the mean detection probability of individual clicks and 5–6% increased detectability of a group of whales (Table 1). Mid-size animals had a slightly longer detection range than social groups, with mar-

ginally higher mean source levels and directivity. For all classes, individual clicks were rapidly less detectable between 2 and 5 km from the sensor at all sites, with the exception of DT, which was on average 10% lower than MC and GC. Differences in the detectability of individual clicks were likely driven by the steep

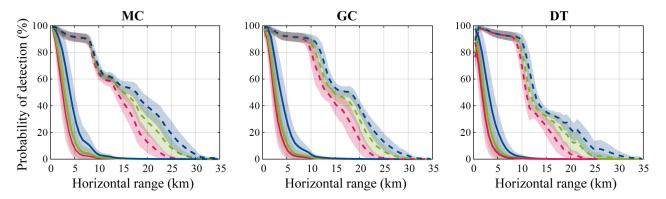


Fig. 5. Estimated mean detection probability for individual sperm whale clicks (solid lines) and groups (dashed lines) in 5 min bins for social groups (red), mid-size animals (green), and adult males (blue). Shaded areas show \pm 1 SD. Refer to Fig. 1 for site abbreviations

bathymetry of this site. Clicks from a group of whales were detectable 70-90% on average at 8 km and declined rapidly after 8-15 km. Detectability of a group was also lower at greater distances at DT, with a probability of detecting 50% of a group at a distance below 15 km compared to 15-20 km at MC and GC.

3.4. Densities and trends in sperm whale demographic segments

Mean weekly densities per demographic class and site were estimated between 2010 and 2017 incorporating class-specific parameters using cue (Table 4) and group counting (Table 5). Social groups dominated across all sites, accounting for 92–98% of the overall densities, followed by mid-size animals representing 1–6% and adult males only 0.2-1.4%. Social groups occurred in high densities at MC, near the wellhead (2.3 and 2.7 animals $1000~\rm km^{-2}$ for cue and group counting, respectively) and in lower densities at GC (cue and group: $0.8~\rm and~1.1$ animals $1000~\rm km^{-2}$). At these 2 northern sites, social groups were present year-round, but densities gradually decreased over the 7 yr period (Fig. 6, Table 6), with a greater annual decline at MC (cue and group: $-3.4~\rm and~-7.4\%$) than at GC (cue and group: $-1.6~\rm and~-2.4\%$). The southern site, DT, had the lowest den-

Table 5. Sperm whale weekly mean densities and coefficients of variation (CV) derived from group counting by site and demographic class. A maximum horizontal detection range (w) of 35 km was used to estimate densities. Estimated density trends regardless of class are referred to as generalized. MC: Mississippi Canyon; GC: Green Canyon; DT: Dry Tortugas

Class	Site	\hat{D}_{kt} Mean density(animals 1000 km ⁻²)	(\hat{D}_{kt})	$n_{kt} T_{kt}^{-1} $ (n s ⁻¹)	\hat{c}_{pk} False positive (% bins		\hat{c}_{pk} False negative (% bins)		\hat{s} Mean group siz (animals		\hat{P}_v Prob group vocal	$ \begin{array}{c} \text{CV} \\ (\hat{P}_v) \end{array} $	\hat{P}_k Probability of detection	$ \begin{array}{c} \text{CV} \\ (\hat{P}_k) \end{array} $
Demographic	-speci	fic												
Social group	MC GC DT	2.736 1.102 0.437	0.013 0.014 0.024	0.122	0.1 0.1 0.1	0.01 0.01 0.01	3.5 3.5 0.5	0.01 0.01 0.01	6.1 6.1 6.1	0.18 0.18 0.18	1.00 1.00 1.00	0.06 0.06 0.06	0.174 0.181 0.132	0.007 0.008 0.007
Mid-size	MC GC DT	0.037 0.027 0.024	0.035 0.033 0.038	0.013 0.010 0.007	0.1 0.1 0.1	0.01 0.01 0.01	9.5 7.5 5.0	0.01 0.01 0.01	1.5 1.5 1.5	0.25 0.25 0.25	0.65 0.65 0.65	$0.04 \\ 0.04 \\ 0.04$	0.228 0.233 0.173	0.008 0.007 0.009
Adult male	MC GC DT	0.007 0.008 0.006	0.044 0.083 0.052	0.003 0.004 0.002	0.1 0.1 0.1	0.01 0.01 0.01	8.0 2.0 18.0	0.01 0.01 0.01	1.5 1.5 1.5	0.25 0.25 0.25	0.65 0.65 0.65	0.04 0.04 0.04	0.283 0.279 0.218	0.007 0.006 0.009
Generalized														
	MC GC DT	2.971 1.273 0.589	0.012 0.014 0.021	0.314 0.141 0.049	0.1 0.1 0.1	0.01 0.01 0.01	3.5 3.5 0.5	0.01 0.01 0.01	6.1 6.1 6.1	0.18 0.18 0.18	1.00 1.00 1.00	0.06 0.06 0.06	0.174 0.181 0.132	0.007 0.008 0.007

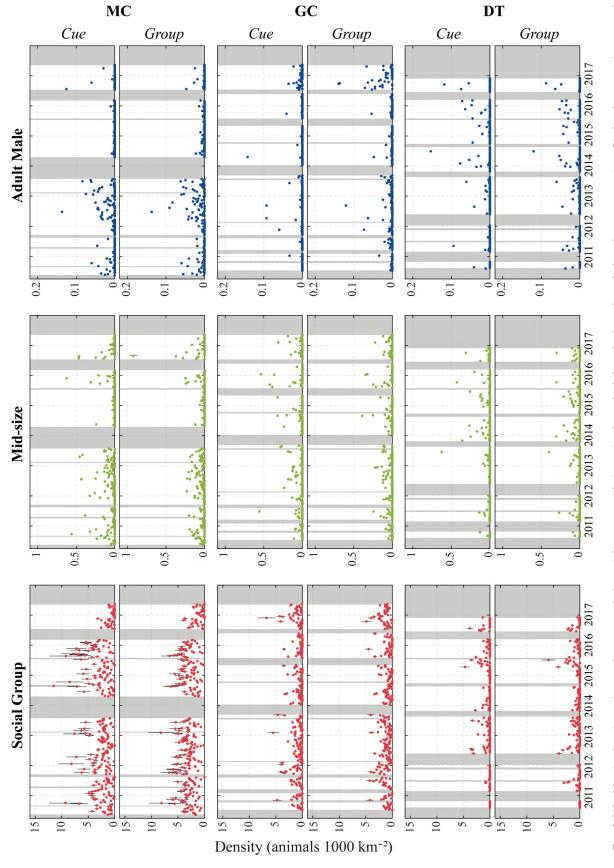


Fig. 6. Weekly site density estimates for sperm whale social groups, mid-size animals, and adult males based on click and group counting. Circles denote estimates and vertical lines show ± 1 SD. Shaded areas lack recording effort. Density scale varies for each demographic class. Refer to Fig. 1 for site abbreviations

Site	Class	Annu	Cue al % change		Group al % change	Trend
		Mean	95% CI	Mean	95% CI	
MC	Generalized Demographic-specific	-3.77	[-6.07, -1.20]	-6.51	[-8.32, -5.03]	¥
	Social group	-3.41	[-6.31, -0.78]	-7.41	[-9.47, -5.31]	¥
	Mid-size	-6.33	[-11.93, -1.62]	-5.12	[-9.52, -1.43]	¥
	Adult male	-7.43	[-14.21, -1.59]	-10.07	[-15.45, -5.56]	Ý
GC	Generalized	-0.25	[-4.01, 2.95]	-0.51	[-3.16, 1.51]	Ý
	Demographic-specific					,
	Social group	-1.61	[-5.36, 1.46]	-2.36	[-5.24, 0.96]	¥
	Mid-size	4.47	[0.86, 10.27]	3.39	[-0.07, 8.93]	À
	Adult male	2.03	[1.09, 3.82]	14.36	[9.64, 18.33]	Â
DT	Generalized Demographic-specific	16.27	[10.90, 25.02]	7.90	[5.02, 11.91]	Â
	Social group	15.87	[7.17, 21.80]	8.08	[4.13, 11.94]	A

[4.24, 14.49]

[9.11, 25.39]

Table 6. Mean density trends as an annual percent change and 95% confidence interval (CI) by site, demographic class, and counting method (cue, group) from 2010 to 2017. Estimated density trends regardless of class are referred to as generalized.

MC: Mississippi Canyon; GC: Green Canyon; DT: Dry Tortugas

sities of social groups (cue and group: 0.3 and 0.4 animals 1000 km^{-2}) where presence was most variable, consistently increasing in spring/summer (Fig. 6). In contrast with northern sites, social group densities increased over this period at DT (Table 6; cue and group: 16 and 8% annual change).

9.47

16.24

Mid-size

Adult male

Mid-size animal densities were similar across sites, with mean weekly densities marginally higher at MC (cue and group: 0.05 and 0.04 animals 1000 km^{-2}) than GC and DT (0.03 and 0.02 animals 1000 km^{-2} , respectively; Tables 4 & 5). Mean densities of mid-size animal across sites were an order of magnitude smaller than the lowest mean densities of social groups at DT. Mid-size animal densities fluctuated among the 7 years (Fig. 6), and different trends were observed across sites (Table 6). Densities declined at MC (cue and group: -6 and -5% annual change) but increases were observed at GC (cue and group: 5 and 3% annual change) and at moderately higher rates at DT (cue and group: 10 and 9% annual change).

Adult males had the lowest densities among classes, being relatively rare at all sites with mean densities of 0.007 and 0.006 animals 1000 km $^{-2}$ at MC and DT, respectively, and marginally higher at GC (0.008 animals 1000 km $^{-2}$; Tables 4 & 5, Fig. 6). Similar to mid-size animals, adult male densities declined at MC (cue and group: -7 and -10% annual change) but increases were observed at GC (cue and group: 2 and 14% annual change) and at moderately higher rates at DT (cue and group: 16 and 11% annual change).

3.5. Densities with and without demographic considerations

[3.68, 14.97]

[5.61, 17.88]

9.16

11.25

Generalized mean densities per site were estimated for the 7 yr period using parameters treating the entire population as a single entity for both cue (Table 4) and group counting (Table 5). A comparison between generalized mean densities and inclusive mean densities, which incorporated the sum of the 3 demographic class densities, showed that the bias from ignoring demographics in the GoMex was small (0.2–2%) for cue counting, but larger (7–23%) for group counting, consistently leaning towards overestimation (Table 7). The bias was lowest at MC and highest at DT. Discrepancies in counting methods improved at all sites from 25% for generalized densities to 18% for inclusive densities at MC, 38 to 27% at GC, and 43 to 23% at DT.

4. DISCUSSION

Accurate estimates of animal densities and trends are fundamental for effective wildlife management and conservation. PAM distance sampling methods require multiple assumptions when estimating parameters to convert acoustic detections into absolute numbers of animals. We illustrate that applying generalized assumptions about animal behavior at the species or population level can introduce bias in density estimates for sperm whales. This species exhibits

Table 7. Comparison of density estimates with and without demographics differences. 'Generalized' ignores demographic differences, whereas 'inclusive' adds the estimated density sum of each demographic segment: social group, mid-size animals, and adult males. Difference estimated as the density change between methods divided by average densities of methods per site.

MC: Mississippi Canyon; GC: Green Canyon; DT: Dry Tortugas

Counting method	Site	Generalized mean density (animals 1000 km ⁻²)	Inclusive mean density (animals 1000 km ⁻²)	Generalized- inclusive difference (%)
Cue	MC GC DT		2.314 0.866 0.372	0.2 0.3 2.0
Group	MC GC DT		2.780 1.138 0.467	6.7 11.2 23.1
Cue- group difference	MC GC DT	24.6 % 37.8 % 43.3%	18.3% 27.2% 22.7%	

sexual dimorphism in adult whales, characterized by variations in click characteristics and diving behavior, group sizes, and stratified spatial distribution, resulting in non-conformity to assumptions of equal detection probability between sex/age groups. Monitoring demographic structure in the northern GoMex using dominant ICI distributions as a proxy for size classes (Solsona-Berga et al. 2022) delineated acoustic detections of 3 demographic segments. Demographic specific parameters were considered when estimating densities and allowed for separate analysis of demographic trends.

Misdetection and classification of sperm whales can lead to biased density estimates. Accounting for imperfect detection required estimating the proportion of false positives and negatives. False positives were manually calculated from a random subset of the 7 yr data set at each site. False positive detections, attributed to other marine mammals and noise sources, were estimated regardless of class, and so differences in false positives between demographic classes are unknown but assumed to be relatively small given manual categorization by 2 analysts. This dual-analyst approach helped identify and resolve discrepancies, enhancing classification accuracy. However, manual classification still involves some subjectivity. This underscores the need for ongoing improvements and the potential integration of automated methods in future studies to ensure consistency. Deviation between predicted and measured clicks at low amplitude provided a means to estimate percentage of false negatives per site, class, and counting method and illustrated lower sensitivity of detecting clicks from a group of whales near the amplitude detection threshold (135 dB_{pp} re:

 $1~\mu Pa)$, particularly in mid-size animals and adult males. This reduced sensitivity likely stems from the difficulty of classifying demographics with only a few clicks. Although the number of unassigned time bins was low, those with a small number of clicks are inherently less reliable for classification. Distant and smaller groups of mid-size animals and adult males may only be detected through a few clicks, increasing the likelihood of these groups being unassigned and missed.

Cue counting involves estimating clicking rate, which varied slightly between sites for each class driven by the measured ICIs. Differences in ICI between sites were minimal for each demographic class (SD <100 ms), re-

sulting in minor differences in average click rate estimates. Some differences could be explained by differences in the total number of clicks available for analysis, with far larger numbers of clicks at the northern sites. The proportion of time spent clicking was computed using behavioral data from acoustic tag records of females and juveniles in the GoMex (Watwood et al. 2006). Although populations in subtropical and temperate latitudes involving both sexes spend equivalent amounts of time clicking (Watwood et al. 2006, Ward et al. 2012), future studies could validate this information by placing acoustic tags on GoMex adult males.

Group counting requires an estimate of group size and synchrony. While social group size is well documented in the northern GoMex based on extensive fieldwork (Jochens et al. 2008) and synchrony was estimated from a unique instance of simultaneously tagged whales from the same group (Jochens et al. 2008), information on adult male and mid-size animal group sizes is limited. Expected small group sizes for adult males and mid-size animals were supported by an ad hoc test directly counting simultaneously received click sequences as a proxy for group size (Hildebrand et al. 2015) when animals were detected near the sensor. Future studies could validate group size estimates and minimize uncertainty by acoustically tracking several animals from these demographic segments.

Parameters unknown for certain demographics or only known for other populations were approximated with a grid-search approach (Hildebrand et al. 2019) and provided information on differences in diving behavior between sites and signal characteristics between demographics. However, while grid-search optimization tested all parameter combinations within pre-selected ranges, it was computationally intensive and may have missed complex parameter interactions, leading to convergence on local minima rather than the global optimum. This is because it evaluates parameters independently and systematically. Implementing techniques like Bayesian optimization could address these issues, capturing parameter dependencies and providing more accurate parameter optimization.

Grid-search indicated a consistent water-column foraging depth of approximately 500 m across all sites, in line with tag data diving behavior from the northern GoMex (Watwood et al. 2006). At northern sites, modeling revealed that 40-50% of foraging dives occur near the seafloor (800–1000 m), showing a bimodal diving pattern with mid-water and benthic dives, consistent with tag data collected in the northern GoMex (Mate et al. 2017). In contrast, modeling suggested virtually exclusive mid-water column foraging at the southern site with deeper bottom depth (1300 m). Tagging animals near the southern site would be valuable for validating these potential foraging differences. While more complex models could explore dive patterns based on time of day (e.g. determining if benthic dives occur at specific times) or depth of prey fields, tracking or tagging studies could address these potential foraging differences. Grid-search estimated source levels of 237 ± 5 dB_{pp} re: 1 μ Pa for social groups, 238 \pm 5 dB_{pp} re: $1 \mu Pa$ for mid-size animals, and $242 \pm 5 dB_{pp}$ re: 1 μPa for adult males. Although source levels of females and juveniles have not been directly measured in previous studies, they are expected to be lower than those of adult males given their smaller body size (Jensen et al. 2018). Our findings align with measured source levels for adult males, with apparent source levels up to 229 dB_{peak} (Zimmer et al. 2005) and 235 dB $_{rms}$ re: 1 μPa (Møhl et al. 2003), which, assuming a sinusoidal wave, are equivalent to $235-244 \text{ dB}_{DD}$ re: 1 μPa . Notably, at DT, the modelderived source levels were 2 dB lower across all classes compared to the northern sites. These lower source levels were necessary to achieve a good fit between the predicted and measured received level distributions. The reason for this difference, and its relation to whales foraging exclusively at mid-water depths at DT, remains unclear. These differences may potentially be related to our model simulations assuming a uniform distribution of sperm whales around the sensor to facilitate detection probability estimation. Actual sperm whale distribution may be more complex, particularly in sites with varying bathymetry, such as the sharp slope of site DT. Gridsearch estimated that directivity was consistent for mid-size animals and adult males (32.5 ± 2.5 dB) but lower for social groups (30 ± 2.5 dB). These values are slightly higher than the previously measured directivity of 27 dB for adult males (Møhl et al. 2003, Zimmer et al. 2005, Nosal & Frazer 2007). Our estimates of directivity for mid-size animals and adult males may have slightly decreased the detectability of these classes, as small changes in this input parameter resulted in a maximum 4-5% change in overall detectability predictions.

Modeling detection probabilities revealed substantial differences in the detection range of each demographic class and site. Adult males were detectable at larger distances than social groups and mid-size animals. The distance at which individual clicks of adult males could be detected was approximately 5 km farther than the other classes, and 10 km farther for the detection of a group. Differences in detectability of individual clicks of social groups and mid-size animals were marginal but more prominent when detecting a group of mid-size animals, which were detectable approximately 2-5 km farther away than social groups. Overall, individual clicks became less detectable at shorter distances (2-5 km) than for a group (20-25 km) under the same environmental conditions. This is because the probability of an individual click being on-axis is very low, but higher for the group during a 5 min window (Frasier et al. 2016). Larger detection ranges of groups were expected since on-axis clicks are detectable at relatively far ranges. Differences in detectability between the southern site (DT) and the northern sites (MC and GC) were observed, likely driven by the steep bathymetry of the southern site and our model simulations assuming a uniform distribution of sperm whales around the sensor. Detection probability models are also sensitive to assumptions about signal characteristics and diving behaviors. Source level, beam directivity, and bimodal dive fraction had the most influence on detectability predictions, with source level having the most substantial impact. Future studies would benefit from tagging or long-term acoustic tracking data to understand the non-uniform distribution of sperm whales and signal characteristics or diving behavior of the different demographics, providing more realistic simulation models and improving density estimation accuracy. In the absence of such data, grid-search optimization has provided approximate estimates despite its limitations in capturing complex parameter interactions.

4.1. Impact of demographic differences in density estimation

Ignoring population demographics biases sperm whale density estimation from passive acoustics. Larger differences between generalized and inclusive methods occurred in group counting, while small differences were found with cue counting. Variation among sites highlights non-uniform bias, with the lowest differences at MC and the highest at DT. Dominant presence of social groups at MC may explain the relatively modest bias when demographics were ignored, contrasting with the highest bias at DT with the greatest demographic variability. These findings may be important in ocean basins with less skewed demographic composition and greater variations in spatial and temporal presence, where neglecting demographic differences may lead to larger biases in densities.

Accuracy of each counting method relied on meeting assumptions and estimating parameters reliably. Large differences between cue and group counting occurred with a generalized approach. Inclusive density estimation improved agreement between counting methods, particularly at sites with larger discrepancies, attributed to a more accurate consideration of model assumptions regarding differences in detectability, click rates, and diving behavior associated with demographics.

4.2. Density and trends

We chose a weekly time scale to balance the need for fine-scale estimates and to meet our simulation assumptions about animal positions relative to the hydrophone. Weekly densities can capture relevant patterns in animal movements and population changes, which we plan to analyze alongside oceanographic variations for future habitat model predictions. We estimated a sperm whale mean density of 1.3 animals $1000 \ km^{-2}$ over the 3 sites (1.4 animals 1000 km⁻² when not considering population demographic differences). Compared with earlier large-scale vessel-based population estimates in the northern GoMex of 1180 animals (CV = 0.22) for the 2017-2018 period, with an average density of 1.7 animals 1000 km⁻² (Garrison et al. 2020, Hayes et al. 2022), our estimate aligns reasonably well given the limited spatial sampling with PAM recorders. Data collection at additional sites will provide better spatial coverage and improve application of these methods to the broader region.

Northern sites showed higher densities than the southern site, and different trends were observed among the 3 sites over the 7 yr period. MC in the north-central GoMex exhibited the highest densities among all sites, aligning with previous findings suggesting it is a core area for sperm whales (Jochens et al. 2008). In contrast, the southeast site, DT, had the lowest densities, potentially due to its location near the continental slope. Sperm whales have been sighted at high rates in this area, but most sightings occurred in waters of the abyssal plain with bottom depths greater than 2000 m (Garrison & Dias 2020). Social group presence at DT was most variable, consistently increasing in spring and summer (Solsona-Berga et al. 2022), potentially explained by inshore/ offshore movements across deeper and shallower isobaths. The impact of the GoMex Loop Current on the West Florida continental shelf, particularly near DT, where it meets shallow isobaths, can induce upwelling-favorable motion across the entire slope (Sorinas et al. 2023) and could drive variation in sperm whale presence.

Declines in sperm whale density at the northern sites and increases at the southern site suggest movements in the GoMex region. Trends aligned between the generalized and demographic-specific methods at MC and DT but varied at GC. While the generalized method documented declines at GC, only social groups exhibited such trends, highlighting the importance of demographic consideration to capture relevant shifts in the population. Declines may be linked to elevated noise from anthropogenic activities in the north-central GoMex (Wiggins et al. 2016, Estabrook et al. 2016), where we detected the most ship passages, and the potential lasting consequences of the DWH oil spill, particularly at MC near the wellhead, which had the most pronounced decline. While the full extent of DWH damage on deepwater ecosystems is uncertain, estimates suggest a significant impact in the northeastern quarter of the deepwater GoMex (Murawski et al. 2020), including benthic contamination (Brooks et al. 2015). Approximately 16.1% of the GoMex sperm whale stock was exposed to high oil concentrations (Aichinger Dias et al. 2017). The availability of limited pre-spill ecosystem data complicates damage assessment, relying heavily on postspill time series (Frasier et al. 2020). Population-level declines in intermediate trophic level organisms (Sutton et al. 2022), preyed on by sperm whales, may contribute to observed sperm whale declines in the northern GoMex as a result of the oil spill.

To interpret site-specific trends, a comprehensive understanding of sperm whale responses to local conditions, including both natural and anthropogenic processes, is crucial. Continued monitoring, combining long-term acoustic data with monitoring of the GoMex deep ecosystem (Cook et al. 2020, Woodstock et al. 2021), can aid in identifying the drivers influencing these trends.

5. CONCLUSIONS

This study highlights the importance of accounting for demographic differences in passive acoustics for accurate sperm whale density estimates. Considering differences in detectability, click rates, group sizes, and diving behavior improved the agreement between cue and group counting methods, allowing monitoring of demographic-specific density trends. Small biases for cue counting and moderate biases for group counting resulted at sites with skewed demographic composition, particularly in the GoMex population composed predominantly of social groups. In regions with balanced sex/age group presence, ignoring demographics could introduce higher bias in density estimates. Estimated long-term density trends aligned between generalized and demographic-specific methods and indicate an increase in the southeastern site and a decline in the northern site near the oil spill wellhead. Generalized estimates indicate declines at the northwestern site, whereas demographic-specific density estimates at this site indicate that only social groups exhibited such trends. These findings warrant further research on the drivers behind these changes and highlight the importance of monitoring demographic-specific changes.

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