

1 **Acoustic evidence for seasonal migration by a cryptic top predator of the deep**
2 **sea**

3
4 William K. Oestreich (Monterey Bay Aquarium Research Institute,
5 woestreich.research@gmail.com)

6
7 Kelly J. Benoit-Bird (Monterey Bay Aquarium Research Institute, kbb@mbari.org)

8
9 Briana Abrahms (Center for Ecosystem Sentinels, Department of Biology, University of
10 Washington, abrahms@uw.edu)

11
12 Tetyana Margolina (Naval Postgraduate School, tmargoli@nps.edu)

13
14 John E. Joseph (Naval Postgraduate School, jejoseph@nps.edu)

15
16 Yanwu Zhang (Monterey Bay Aquarium Research Institute, yzhang@mbari.org)

17
18 Carlos A. Rueda (Monterey Bay Aquarium Research Institute, carueda@mbari.org)

19
20 John P. Ryan (Monterey Bay Aquarium Research Institute, ryjo@mbari.org)

21
22 **Statement of authorship:** W.K.O., K.J.B., and J.P.R. conceived the study; W.K.O., K.J.B., and J.P.R.
23 designed the research; J.P.R. collected data; W.K.O., B.A., T.M., Y.Z., C.A.R., and J.P.R. developed
24 methods; W.K.O., T.M., and J.P.R. performed analyses; and W.K.O. wrote the manuscript with
25 contributions from all authors.

26
27 **Data availability statement:** Raw (256 kHz) and decimated (16 kHz) acoustic data from the MARS
28 hydrophone are available at <https://registry.opendata.aws/pacific-sound/>. Code for processing acoustic
29 data, analyzing sperm whale detections, and simulating individual-level movement strategies are
30 available at <https://doi.org/10.5281/zenodo.7860426>.

31
32 **Running title:** Seasonal migration in the deep sea

33
34 **Keywords:** deep sea, movement ecology, migration, resource tracking, phenology, bioacoustics, sperm
35 whale (*Physeter macrocephalus*), echolocation, marine megafauna, pelagic

36
37 **Type of article:** Letter

38 **Word count:** 150 words in abstract, 4894 words in main text

39 **Number of references:** 79

40 **Number of figures:** 6 (0 tables or text boxes)

41
42 **Correspondence:**

43 William K. Oestreich

44 7700 Sandholdt Road

45 Moss Landing, CA 95039

46 woestreich.research@gmail.com

47 +1 (412)-735-1106

1 **Abstract**

2

3 In ecosystems influenced by strong seasonal variation in insolation, the fitness of diverse
4 taxa depends on seasonal movements to track resources along latitudinal or elevational
5 gradients. Deep pelagic ecosystems, where sunlight is extremely limited, represent
6 Earth's largest habitable space and yet ecosystem phenology and effective animal
7 movement strategies in these systems are little understood. Analyzing seven years of
8 continuous population-level passive acoustic observations, we find evidence for seasonal,
9 latitudinal movements by sperm whales (*Physeter macrocephalus*) in the Northeast
10 Pacific. Integration of population-level empirical results with individual-level movement
11 simulations provides evidence of seasonal migration in this cryptic top predator, likely to
12 track deep-sea resources. We show that sperm whales track oceanographic seasonality
13 in a manner similar to many surface ocean predators, but with dampened seasonal-
14 latitudinal movement patterns. These findings shed light on the drivers of sperm whales'
15 long-distance movements and the shrouded phenology of the deep-sea ecosystems in
16 which they forage.

1 INTRODUCTION

2

3 The movement strategies that animals use to track resources in space and time
4 drive many aspects of their ecology, mediate their ability to respond to environmental
5 perturbations, and provide insight into the spatiotemporal dynamics of the ecosystems
6 they inhabit¹. These individual and group-level movement strategies typically result from
7 spatiotemporal patterns of resource availability², and manifest in distinct patterns of
8 population-level distribution in space and time³. For example, nomadic resource tracking
9 has evolved in aseasonal and unpredictable environments, leading to irregular patterns
10 of individual movement and population distribution⁴. Conversely, many species inhabiting
11 seasonal ecosystems have evolved to undertake seasonal migrations between distinct
12 ranges⁴ or perform partial migrations, whereby a specific demographic of the population
13 undertakes migration⁵. These seasonal migrations between distinct habitats (sometimes
14 referred to as “to-and-fro” migrations), as in the migrations of many baleen whales, are
15 distinguished by their persistent, relatively direct movements undistracted by proximate
16 resources⁶. Other seasonal migrants (e.g., many ungulates) undertake seasonal
17 movements to track the phenology of proximate resources (e.g., forage, favorable abiotic
18 conditions, etc.) en route as resource availability propagates across spatiotemporal
19 gradients such as latitudes or elevations^{7,8}. These resource-tracking migrations have
20 recently gained attention as an important connection between ecosystem dynamics and
21 animal movement, closely linking ecosystem phenology with that of seasonal animal
22 migrations^{1,9}. Such resource tracking has been shown to provide a number of individual
23 and population-level benefits, from enabling animals to have more prolonged access to
24 food¹⁰, to increasing fat gain¹¹ and allowing migratory populations to have higher growth
25 rates than sedentary populations¹². These linkages between resource dynamics and
26 animal movement strategies are increasingly well-understood in seasonal terrestrial^{2,7,9,13},
27 freshwater¹⁴, coastal marine¹⁵, and epipelagic¹⁶⁻²¹ ecosystems across the globe.

28 Few studies have assessed these connections between ecosystem dynamics and
29 animal movement in Earth’s largest habitable space: deep pelagic ecosystems. These
30 oceanic waters deeper than 200 meters, where little sunlight penetrates, have historically

1 been characterized as stable and aseasonal, but poorly understood²². However, a
2 growing body of evidence suggests elements of seasonality in the deep sea. For example,
3 oceanographic studies have documented seasonal variation in the transport of biomass
4 from the surface to the deep²³⁻²⁵. Further research has documented seasonality in
5 sightings and biomass of low and mid-trophic level organisms in the mesopelagic²⁶⁻²⁸. Yet
6 understanding of deep-sea phenology remains limited, particularly for highly mobile and
7 high-trophic-level animals. This knowledge gap is underpinned by the challenge of
8 making continuous and detailed observations in these ecosystems²². Given the global
9 extent, high endemic biodiversity, and major role in global biogeochemical cycles of deep
10 pelagic ecosystems, understanding the phenology of these ecosystems and the evolved
11 movement strategies of their inhabitants is important to advance fundamental ecology
12 and inform ecosystem management.

13 We address this gap by integrating long-term passive acoustic monitoring data and
14 movement simulations for a deep pelagic top predator, the sperm whale (*Physeter*
15 *macrocephalus*). Sperm whales are a deep-diving oceanic predator, diving to depths of
16 hundreds to thousands of meters²⁹ to forage on diverse deep pelagic prey³⁰. Thus,
17 studying the movement patterns of these ocean giants can provide a rare window into the
18 phenology of the deep-sea environment. In addition, sperm whales produce the loudest
19 known biological sounds³¹ which not only reveal the presence of this often-cryptic species
20 over large ocean volumes, but also transmit rich behavioral and demographic information
21 about detected individuals. Echolocation clicks are central to the foraging ecology of
22 sperm whales in the low-light conditions of the deep sea, and further indicate individuals'
23 behavioral state (foraging), size (both inter-click-interval³² and inter-pulse-interval within
24 individual clicks³³ correlate with size), and sex and age-class (sperm whales are sexually
25 dimorphic³⁴, allowing for sex and age-class identification via inter-click-interval³²). Sperm
26 whales use echolocation in both the meso- and bathypelagic³⁵ to locate a variety of squid
27 and fish prey species³⁰. As a result, patterns of sperm whale echolocation click detection
28 can provide insight into the phenology of both this top predator and the deep pelagic
29 ecosystems in which they forage.

1 In the Northeast Pacific, foraging sperm whales have been detected acoustically
2 year-round, specifically in the Gulf of Alaska (GoA)³⁶⁻³⁸. Individuals of this population have
3 expansive home ranges, exhibiting wide-ranging movements which include travel
4 between the GoA and the Central California Current System (CCCS; Figure 1A) among
5 other lower-latitude habitats³⁹⁻⁴¹. Yet the regularity, seasonality, and behavioral context of
6 such movements have historically remained unclear. Previous studies based on
7 individual-level sightings, genetic, and limited telemetry data have hypothesized that
8 latitudinal movements are likely irregular, resulting from aseasonal nomadic movements⁴⁰
9 consistent with the canonical view of aseasonal deep-sea ecosystems. Yet recent
10 acoustic studies in the GoA have suggested seasonality in foraging sperm whales'
11 presence³⁶⁻³⁸, challenging the hypothesis of aseasonal nomadic movements. Others have
12 suggested that long-distance latitudinal movements represent migration between distinct
13 high-latitude foraging and low-latitude breeding habitats⁴², akin to the seasonal migrations
14 of many baleen whales. Sex-specific partial seasonal migration (with only adult males
15 undertaking migration to higher latitudes) has also been hypothesized based on
16 individual-level sightings data^{34,43}, but both sexes have been observed in both the GoA⁴⁰
17 and CCCS^{40,44}. Further, individuals with small body size (females and juveniles) are heard
18 year-round in the GoA³⁸, refuting the hypothesis that only adult males undertake long-
19 distance movements to high latitudes. While individual-level telemetry data can often
20 provide sufficient sample sizes to understand population-level seasonal movement
21 strategies¹⁶, individual tracks of sufficient duration to assess seasonal movement are
22 extremely limited for this sperm whale population³⁹. As with most inhabitants of deep
23 pelagic ecosystems, this murky understanding of sperm whales' movement strategies
24 arises from the challenge of observing their behavior persistently at sufficient scale^{45,46}
25 and limited understanding of phenology in their foraging habitat.

26 Here, we investigate the strategies underlying movements of this deep pelagic top
27 predator in the Northeast Pacific. We consider four hypothesized movement strategies.
28 Three have previously been hypothesized: nomadic resource tracking⁴⁰, seasonal to-and-
29 fro migration between distinct habitats^{39,42}, and sex-specific partial seasonal migration^{34,42},
30 The fourth, seasonal resource-tracking migration akin to that observed in many surface

1 ocean and terrestrial predators^{16,19}, is hypothesized here based on growing evidence of
2 seasonality in the deep sea at lower trophic levels²³⁻²⁸. We first characterize seasonal
3 patterns of foraging sperm whale presence in the Central California Current System as
4 compared to previously published results from the Gulf of Alaska by applying automated
5 acoustic detection methods to more than seven years of passive acoustic recordings.
6 Passive acoustic monitoring approaches provide a valuable Eulerian lens to assess
7 population-level animal presence and behavior⁴⁷, particularly in largely inaccessible
8 oceanic ecosystems when Lagrangian tracking data (e.g., telemetry) is scarce (as with
9 sperm whales in the Northeast Pacific), and in cases where information beyond presence
10 alone (e.g., behavioral state) can be discerned from the properties of detected acoustic
11 signals⁴⁸. We then test the alternative hypotheses by comparing these empirical patterns
12 with emergent patterns derived from simulations of individual-level movement driven by
13 each of the hypothesized movement strategies. Finally, we compare empirically observed
14 seasonal-latitudinal patterns of foraging sperm whale presence to seasonal-latitudinal
15 patterns in the location of the North Pacific Transition Zone, the dominant foraging habitat
16 which numerous surface ocean predators track in the North Pacific^{16,49}. Hypothesis-
17 testing using this integrated approach allows us to (i) determine the unknown seasonality
18 and regularity of foraging sperm whale presence in the Central California Current System
19 and (ii) evaluate the individual-level strategies underlying sperm whales' wide-ranging
20 movements by comparing simulated and observed patterns.

21

22 **MATERIALS AND METHODS**

23

24 **Hydrophone recordings**

25 To assess seasonal and interannual patterns of sperm whale presence in the
26 CCCS, we analyzed passive acoustic recordings between 2015 and 2022 with nearly
27 continuous (>95%) temporal coverage. Acoustic recordings were collected on the
28 continental slope outside Monterey Bay, CA, via two icListen hydrophones sequentially
29 deployed on the Monterey Accelerated Research System (MARS) cabled observatory (36°
30 42.75'N, 122° 11.21'W; depth 891 m; Figure 1A). These hydrophones recorded at 256

1 kHz; all recordings were decimated⁵⁰ to a sample rate of 16 kHz before analysis. While directional components of sperm whale echolocation clicks can have a peak frequency exceeding the Nyquist frequency of these 16 kHz audio files³¹, this sample rate allows for reliable detection of the omnidirectional low-frequency component of these clicks. Previously, these clicks have been reliably detected in audio files with a sample rate as low as 1 kHz³⁶.

8 **Passive acoustic analyses**

9 Sperm whales produce a variety of click types associated with distinct behaviors. The present analysis focused only on “usual” clicks, which are used for echolocation³⁴ and are hereafter referred to as clicks. We used a two-step automated workflow (detection and filtration) to determine presence or absence of sperm whale clicks at daily resolution.

13 Candidate detections of individual clicks were generated using a band limited energy detection (BLED) approach implemented in Raven Pro v1.6⁵¹. We manually tuned the parameters of a BLED (Table S2) to maximize the chances of detecting sperm whale clicks under a range of background noise scenarios, but this first step in acoustic processing also generated many false positives. These false positives were filtered out in the second step of our automated workflow by searching BLED results for repetitive, evenly-spaced sequences of detections matching the known inter-click interval (ICI) range of sperm whale clicks (~0.5 – 2.0 seconds⁵²). Because the intervals between clicks in sperm whale echolocation sequences are largely regular but not exactly constant (Figure 1C), we calculated the time difference between each BLED detection (inter-detection interval; IDI), then rounded to the nearest quarter second to enable a search for sequences of detections with a near-constant IDI. Each day of recording was automatically searched for IDI sequences matching three criteria: (1) rounded IDI must be between 0.5 and 2.0 seconds (inclusive); (2) rounded IDI must be constant; and (3) the number of consecutive IDI values meeting criteria (1) and (2) must meet a sufficient number of repetitions (r) to confidently determine sperm whale echolocation click presence. We considered any day with at least one sequence meeting these criteria to have sperm whale clicks present; all other days were considered to have such clicks

1 absent. Setting the number of repetitions required to consider clicks present can
2 significantly impact the accuracy of this automated workflow at daily resolution (Figure
3 S1). The optimal value for this parameter was determined via comparison to manual
4 identification of sperm whale search clicks. Manual assessments were completed for one
5 randomly chosen day of each month in both 2016 and 2020, as well as two days of known
6 sperm whale presence near our recording location in late 2022. These 26 days provided
7 a representative range of soundscape conditions by covering the full seasonal cycle,
8 including periods recorded by each of the two consecutively-deployed hydrophones, and
9 including recording periods both affected (2020) and unaffected (2016, 2022) by the
10 change in anthropogenic noise conditions associated with the COVID-19 pandemic⁵³. We
11 found optimal performance at $r=6$, yielding a daily balanced accuracy of 97%
12 (precision=100%, recall=94%) and false positive rate of 0% (Figure S1).

13 Using this time series of daily-resolution presence and absence, we then
14 calculated monthly percent of recording days with foraging sperm whales present over
15 the time series. This metric is effective in the study context for multiple reasons: (1) it
16 provides sufficient temporal resolution to assess seasonal trends, the primary timescale
17 of focus in this study; (2) automated detector performance is very high at daily resolution
18 (Figure S1), providing high confidence in this metric; and (3) this metric matches that used
19 in previous studies of foraging sperm whale presence at Ocean Station PAPA in the Gulf
20 of Alaska (GoA) over the years 1999-2001³⁶ and 2007-2012³⁷, allowing for comparison
21 of seasonal presence of foraging whales across a large latitudinal range. Monthly percent
22 presence values from the GoA were determined by digitizing the figures presenting this
23 information in previous studies^{36,37} and were later used in comparison to simulation
24 results. The seasonal patterns from these earlier studies^{36,37} match those recorded more
25 recently in the GoA³⁸ (2011-2019), with all studies showing a summer maximum and
26 winter minimum of foraging sperm whale presence in the GoA.

27 Seasonality in the detection of foraging sperm whales in the CCCS was assessed
28 statistically via a generalized additive model of monthly percent presence as a function of
29 month with year nested as a random effect, to test for the deviance in percent presence
30 explained by the seasonal cycle alone. Finally, because inter-click-interval (ICI) correlates

1 with body size and demographic group³² and therefore can help assess the hypothesis of
2 sex-specific partial migration, we calculated the ICI of all detected click sequences in the
3 time series. We used ANOVA to test for seasonal effects on natural-log-transformed ICI
4 distribution. To test for correlation between monthly mean ICI and monthly foraging sperm
5 whale presence, we used linear regression.

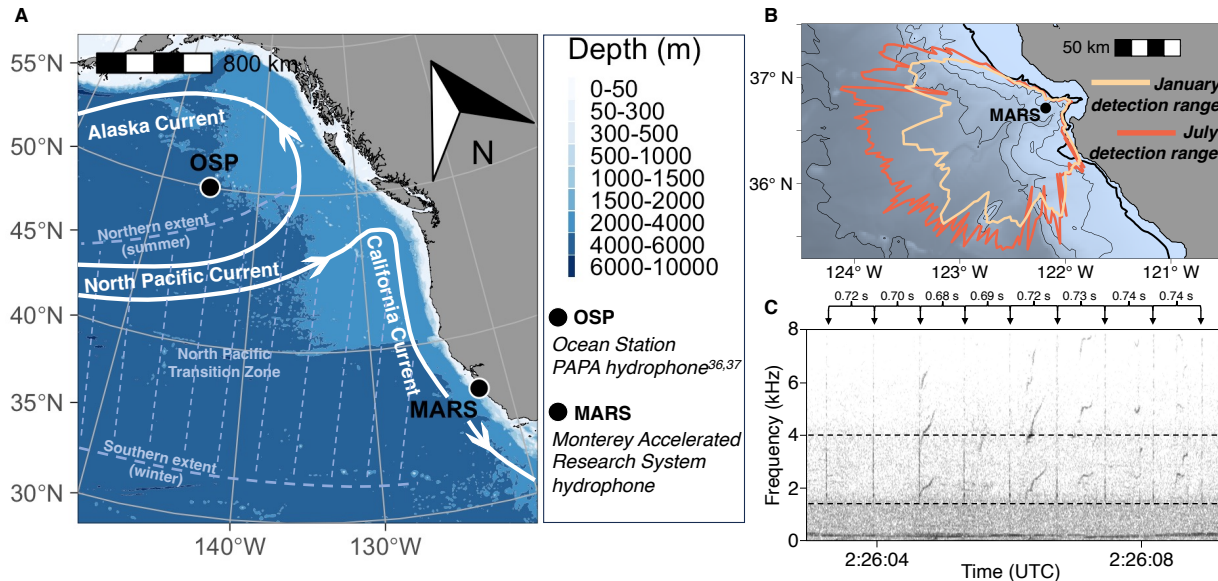
6

7 **Estimation of detection range**

8 Because seasonality in foraging sperm whale detection could be influenced by
9 seasonal differences in detection range, we assessed seasonality in both ambient noise
10 levels and acoustic propagation loss between sound source and the acoustic receiver at
11 MARS. From daily files of 16 kHz audio data spanning the full study period, daily mean
12 noise levels (single-sided mean-square sound pressure spectral density) were computed
13 for the frequency band targeted by the click detector (1.4-4 kHz). These daily ambient
14 noise values were binned by month across years to examine seasonality.

15 Acoustic propagation loss was modeled for January and July to assess seasonality
16 in click detection range (Figure 1B). This acoustic propagation loss modeling was
17 conducted for a sound source matching the characteristics of sperm whale echolocation
18 at the frequencies targeted by the automated detector. For a complete description of
19 propagation loss modeling methods, see the Supporting Information.

20



1
2 **Figure 1. Study system and acoustic methods.** (A) The Northeast Pacific Ocean, showing the location
3 of passive acoustic recordings from the present study (Monterey Accelerated Research System (MARS) in
4 the Central California Current System) and previous studies^{36,37} (Ocean Station PAPA (OSP) in the Gulf of
5 Alaska). Some map elements adapted from ^{16,37}. (B) The Central California Current System, indicating
6 winter and summer detection ranges for sperm whale echolocation clicks produced at 500m depth (see
7 Methods and SI for additional depths) based on average January and July oceanographic conditions over
8 the period 2016-2022. The circle indicates MARS (891m depth), with contours representing the 200 m
9 isobath (thicker line) and multiples of 1000m (thinner lines). (C) Example spectrogram of audio recorded at
10 MARS on November 30, 2022, showing a period when a single foraging sperm whale's echolocation clicks
11 (impulsive, broadband signals) were clearly visible and audible. Dashed horizontal lines indicate the
12 minimum and maximum frequencies of the automated energy detector used to detect sperm whale
13 echolocation clicks. Note the near-constant inter-click-interval used to discern echolocating sperm whales
14 from other impulsive sound sources in this frequency range.

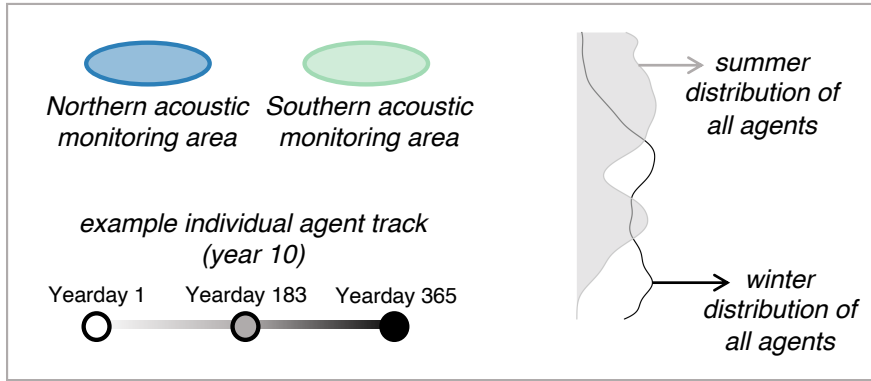
15

16 Simulation of individual-level movement strategies

17 To test hypotheses regarding the individual-level movement strategies underlying
18 empirically observed patterns of foraging sperm whale presence, we developed
19 individual-based movement simulations which we compared to empirical patterns of
20 whale detection. We employed simulations in which agents move through a spatial
21 domain with two hydrophone “listening ranges” (one at higher latitude and one at lower
22 latitude), analogous to passive acoustic monitoring of sperm whales in the GoA^{36,37} and
23 the CCCS (present study). In all simulations, 100 agents moved daily according to
24 strategy-specific decisions over a ten-year period. The spatial domain in which these
25 simulations occurred is not meant to specifically represent the spatial dimensions of the
26 North Pacific or hydrophone listening ranges used in the present or previous studies.

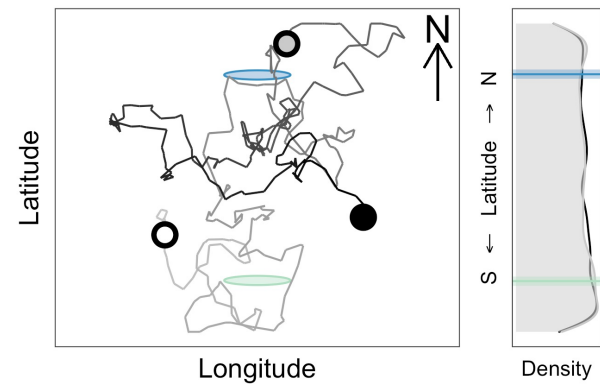
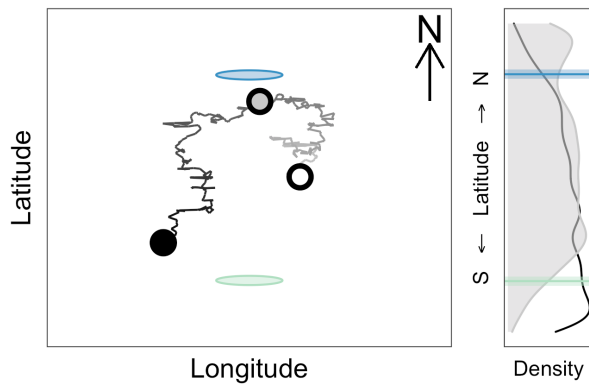
1 Instead, this spatial domain (described in greater detail in the Supporting Information)
2 provides a simplified arena for testing realistic individual movement strategies⁵⁴ and their
3 influence on population-level spatiotemporal patterns of acoustic detection (Figure 2).

4 We used empirically determined information about step length and turn angle
5 distributions, as well as seasonality of movement, for well documented movement
6 strategies across diverse taxa and ecosystems⁵⁴ to formulate movement decision rules
7 for agents representing the four hypothesized movement strategies (Table S2). We
8 examined the population-level acoustic detection patterns resulting from each of these
9 four movement strategies via four separate simulations with agents subject to these
10 decision rules. At each daily timestep of each ten-year simulation, we recorded each
11 agent's position and presence or absence in each of the simulated hydrophone listening
12 ranges. The population-level patterns resulting from each simulation were compared to
13 empirical observations of foraging sperm whale seasonality in the GoA^{36,37} and the CCCS
14 (present study) by calculating the root-mean-square deviation of simulated monthly mean
15 acoustic detection results from both listening ranges relative to empirical results. For a
16 complete description of simulation parameters (following methods established by ⁵⁴), see
17 the Supporting Information and code⁵⁵ accompanying this manuscript.



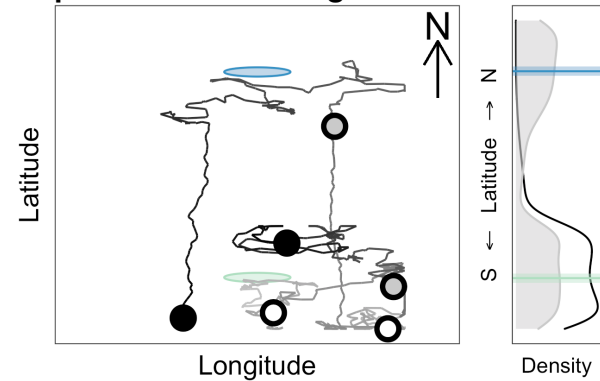
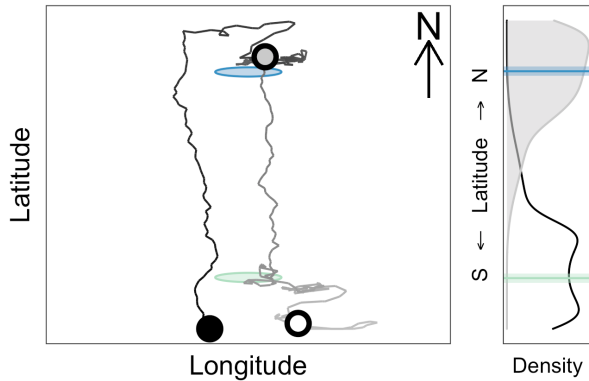
A: Seasonal resource-tracking migration

B: Nomadic resource tracking



C: Seasonal to-and-fro migration between distinct habitats

D: Sex-specific partial seasonal migration



1
2
3
4
5
6
7
8
9
10

Figure 2. Simulated individual-level movement strategies. Top panel provides a legend for the simulation domain. In each of the panels A-D, one individual's track (two individuals, one female and one male, in the case of sex-specific partial seasonal migration) is shown from year 10 of the simulation alongside the summer and winter distribution of all individuals over years 2-10. Circular acoustic monitoring areas appear elliptical due to distortion of the simulation domain in this visualization to highlight individual track details.

1 **Comparison to oceanographic seasonality**

2 To consider whether presence of foraging sperm whales tracks seasonality in
3 oceanographic habitat in a manner similar to many surface ocean predators¹⁶, we
4 compared seasonal patterns of foraging sperm whale presence to seasonal patterns in
5 the location of the North Pacific Transition Zone (NPTZ; Figure 1A). The NPTZ is a major
6 oceanographic feature in the North Pacific Ocean, representing a transition in surface
7 primary productivity between the subpolar and subtropical gyre⁵⁶ and serving as important
8 foraging habitat for a wide range of predators in the surface ocean^{16,49}. The latitudinal
9 position of the NPTZ varies seasonally, reaching its southern extent in the winter and
10 northern extent in the summer (Figure 1A, ⁵⁶). We calculated the monthly latitude of the
11 NPTZ for each month of the acoustic time series as in ⁵⁶, identifying the mean latitude of
12 the 18 °C sea surface temperature (SST) isotherm between 160-180 °W using monthly
13 composite Aqua MODIS 0.025° daytime SST imagery (for comparison to 2015-2022
14 CCCS acoustic metrics) and Pathfinder v5.3 0.0417° daytime SST imagery (for
15 comparison to pre-2006 GoA acoustic metrics and to fill Aqua MODIS data gaps). We
16 then compared the monthly percent of days with foraging sperm whale present to the
17 monthly NPTZ latitude via model II (ranged major axis) linear regression, given
18 uncertainty in both the independent and response variables.

19 20 **Software**

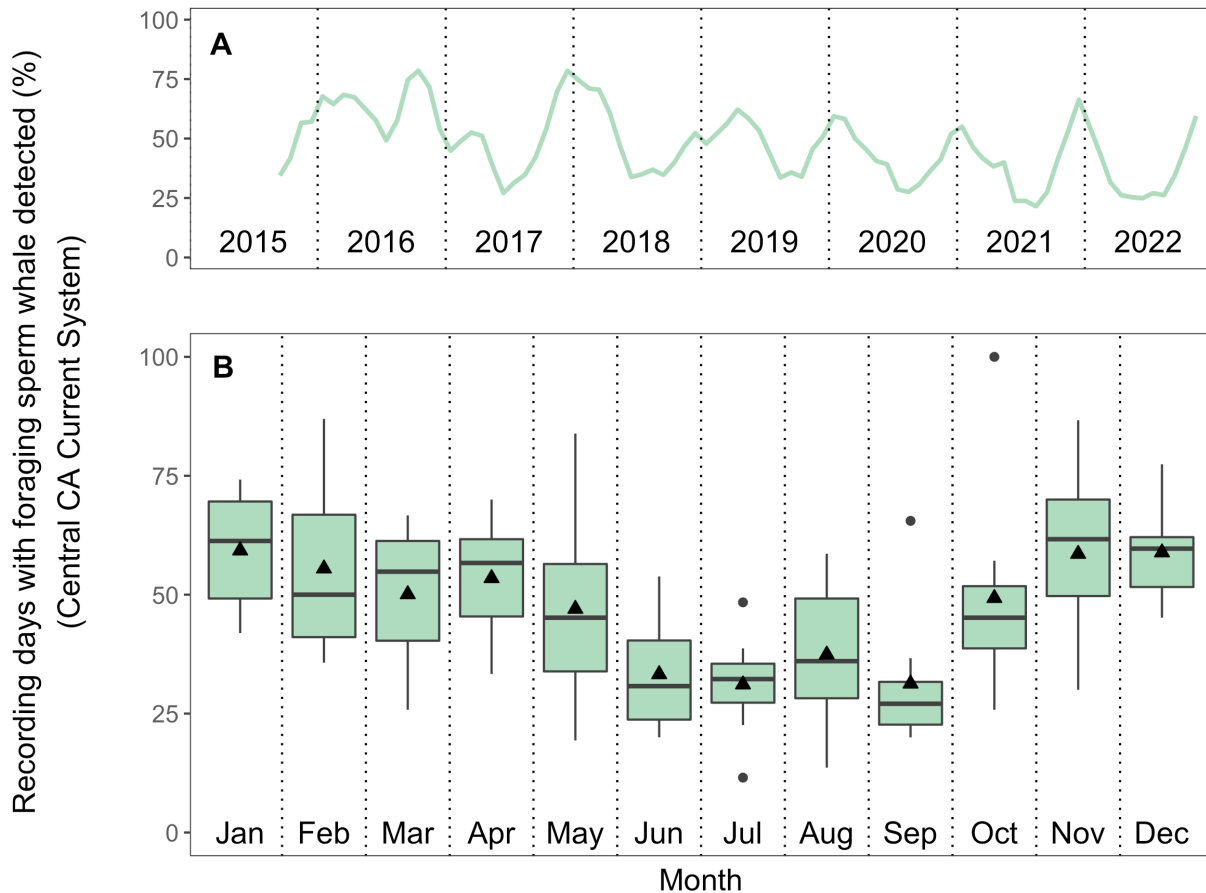
21 All analyses of click detections and individual-level movement simulations were
22 conducted in R v4.2.0⁵⁷. The maps in Figure 1 were created using the packages
23 “ggOceanMaps”⁵⁸, “geosphere”⁵⁹, and “marmap”⁶⁰. Background noise, acoustic
24 propagation, and satellite-based oceanographic analyses were conducted in Matlab⁶¹.
25 Candidate click detections were generated using Raven Pro v1.6⁵⁰.

1 RESULTS

2

3 Seasonality in acoustic detection

4 Acoustic detection revealed year-round, seasonally varying presence of foraging
5 sperm whales in the Central California Current System (CCCS; Figure 3). The frequency
6 of foraging sperm whale presence in the average annual cycle reached a maximum in
7 January (mean of 59.3% of days present) and a minimum in July (mean of 31.1% of days
8 present). A generalized additive model revealed a significant relationship between
9 monthly percent of days with presence and month, with year nested as a random effect
10 ($p < 0.001$; 45.4% deviance explained; Figure S2), indicating seasonality in foraging
11 sperm whale presence in the CCCS. Detection seasonality did not result from seasonal
12 changes in ambient noise or maximum detection range. Maximum click detection range
13 was slightly greater during the summer minimum in click detections relative to detection
14 range during the winter detection maximum (Figures 1B, S3), indicating that the degree
15 of seasonality shown here (Figure 3B) is a conservative estimate. Interannually, the
16 percent of recording days on which foraging sperm whales were detected varied little,
17 with the exception of 2016 (Figure 3A). Foraging sperm whales were detected on 63.4%
18 of recording days in 2016, whereas the percentage in all other years varied between 38.6-
19 49.9%.



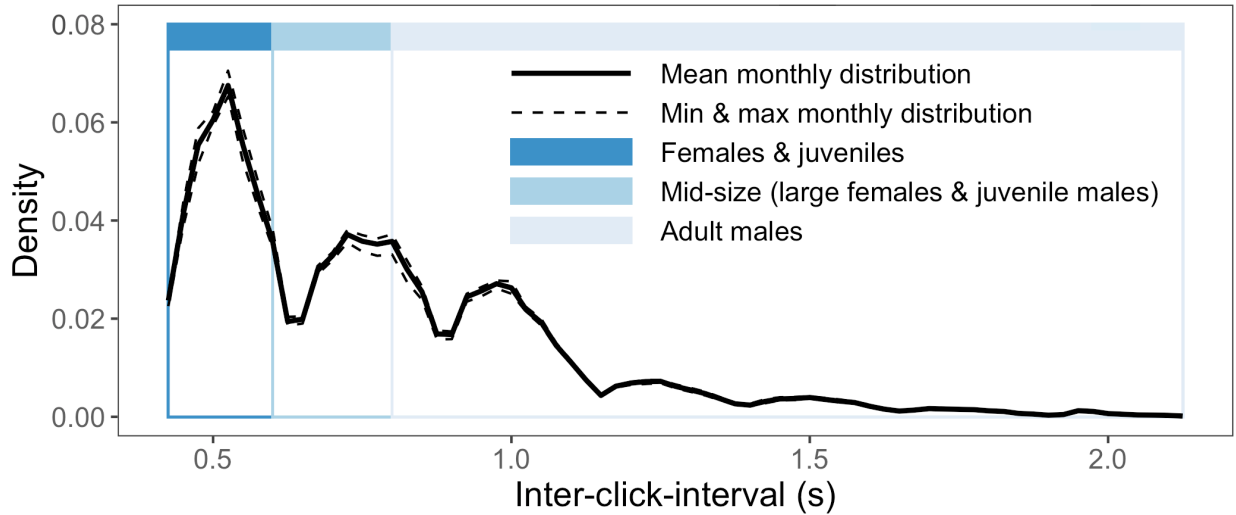
1
 2 **Figure 3. Empirically observed foraging sperm whale presence in the Central California Current**
 3 **System. (A)** Monthly percent presence over the full study period (smoothed with a 3-month running mean).
 4 **(B)** Annual cycle of echolocating sperm whale presence over the full study period (Aug 2015 – Dec 2022).
 5 Boxplots show the median (center line), mean (triangle), 25th-75th percentile (box), ± 1.5 *IQR (whiskers),
 6 and outlying points. A generalized additive model (GAM) revealed a significant relationship between
 7 monthly percent of days with presence and month, with year nested as a random effect ($p < 0.001$; 45.4%
 8 deviance explained; Figure S2),
 9

10 Seasonality of acoustically detected demographic groups

11 Inter-click-interval (ICI) can be used as a proxy for body-size and therefore
 12 demographics of acoustically detected individuals in this sexually dimorphic population³².
 13 Similar to acoustic results from the GoA³⁸, we detect three clear modes of ICI (Figure 4).
 14 We found no seasonality or interannual variation in the distribution of detected ICIs (and
 15 therefore, demographics): ANOVA on natural-log-transformed ICI data indicated no
 16 significant relationship between month ($F=1.52$, $df = 11,70$, $p > 0.14$) or year ($F = 1.70$, df
 17 $= 7,70$, $p > 0.12$) and ICI. We detected individuals with both large body size (adult males,

1 ICI>0.8 s^{32,38}) and small body size (females and juveniles, ICI<0.6 s^{32,38}) in every
2 individual month of the seven-plus year study period.

3



4

5 **Figure 4. Inter-click-interval (ICI) monthly distributions (relative density).** Solid line represents the
6 mean monthly distribution of ICI for detected sperm whale echolocation clicks over the full study period.
7 Dashed lines represent the minimum and maximum monthly ICI distributions at each ICI value. Colors
8 indicate the demographic groups associated with ICI values as per ^{32,38}.

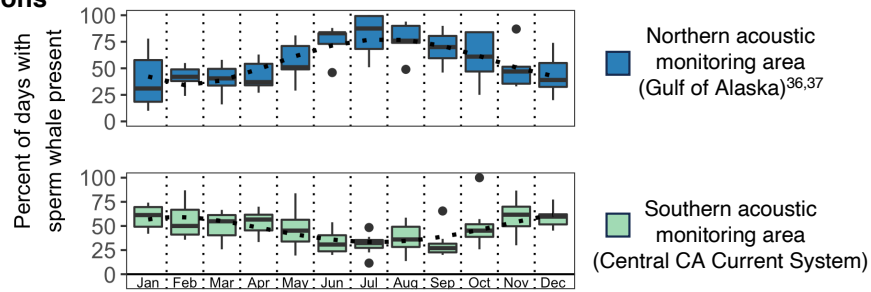
9

10 Individual-level movement simulations

11 Simulations of individual-level movement yielded qualitatively and quantitatively
12 distinct patterns in seasonal-latitude distribution (Figure 2) and seasonal acoustic
13 detection (Figure 5), dependent on the movement strategy employed. The simulation of
14 seasonal resource tracking individuals yielded year-round presence with moderate
15 seasonality at both southern and northern listening ranges (Figure 2A), peaking in the
16 winter and summer for the southern and northern listening ranges, respectively (Figure
17 5B). The seasonal patterns of acoustic detection arising from seasonal resource-tracking
18 migration represented the only simulated results matching the defining qualities of
19 empirically observed patterns: year-round presence with substantial and opposite
20 seasonality at both higher and lower-latitude listening ranges (Figure 5). Agents following
21 nomadic resource tracking decision rules showed no seasonality in detection at northern
22 or southern listening ranges (Figure 5B), driven by similar winter and summer latitudinal
23 distributions (Figure 2B). Agents undertaking seasonal to-and-fro migrations between

1 distinct habitats showed strong and opposite seasonality in latitudinal distribution (Figure
2 2C). This simulation yielded a detection peak during winter and zero detections during
3 summer at the southern listening range, while the northern listening range showed a
4 summer peak in detections and zero detections during winter (Figure 5B). Simulation of
5 sex-specific partial seasonal migration resulted in strong detection seasonality at the
6 northern listening range (high levels of detection in summer, zero detections in winter)
7 and year-round detection with moderate seasonality at the southern listening range
8 (Figure 2D; Figure 5B). Simulated acoustic detection patterns for seasonal resource-
9 tracking migration were also quantitatively most similar to empirical acoustic detection,
10 yielding a root-mean-square deviation among monthly means of only 15.6% (Figure 5B).
11 All other simulated movement strategies resulted in greater deviance from empirical
12 observations (22.4% for nomadic resource tracking, 31.7% for seasonal to-and-fro
13 migration between distinct habitats, 31.9% for sex-specific partial seasonal migration;
14 Figure 5B).

A: Empirical observations



B: Simulations

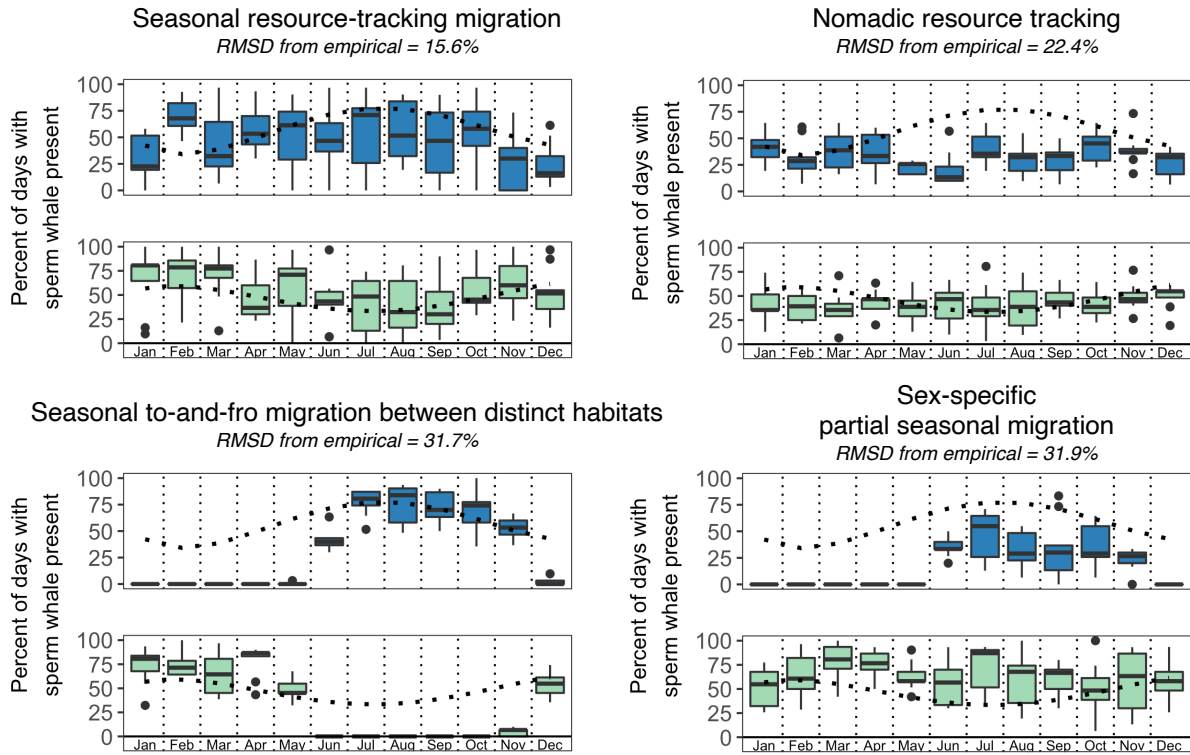


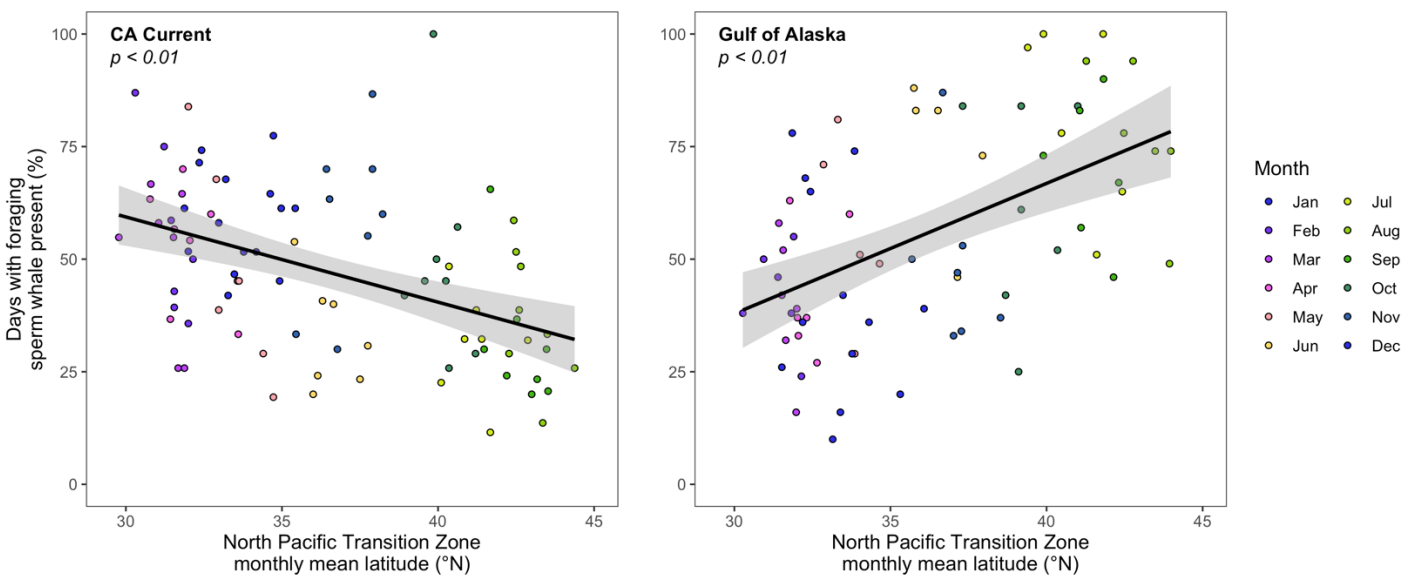
Figure 5. Comparison of empirical and simulated acoustic detection seasonality under hypothesized individual movement strategies. (A) Empirical acoustic detections from the Central California Current System (green; present study) and the Gulf of Alaska (blue^{36,37}). Dotted curves represent a fourth-order polynomial fit to empirical monthly data from each recording site. (B) Acoustic detection at northern (blue) and southern (green) listening ranges for simulated agents following each of the hypothesized movement strategies. Boxplots show the median (center line), 25th-75th percentile (box), $\pm 1.5 \times IQR$ (whiskers), and outlying points of monthly acoustic detection over years 2-10 of each simulation. RMSD refers to the root-mean-square deviation of each simulation's monthly mean acoustic detection results across both hydrophones relative to empirical observations. Empirical data fourth-order polynomial from (A) is overlaid on all simulated results.

1
2
3
4
5
6
7
8
9
10
11
12
13
14
15

1 **Comparison to seasonally shifting oceanographic habitat**

2 Monthly percent presence of foraging sperm whales correlated with oceanographic
3 seasonality in the Northeast Pacific Ocean (Figure 6). The latitude of the North Pacific
4 Transition Zone (NPTZ) was inversely correlated with foraging sperm whale presence in
5 the CCCS (highest detection rate with NPTZ at its southern extent) and positively
6 correlated with foraging sperm whale presence in the GoA (highest detection rate with
7 NPTZ at its northern extent).

8



9

10 **Figure 6. Foraging sperm whale presence follows oceanographic seasonality in the Northeast**
11 **Pacific.** Monthly empirically observed acoustic detection of foraging sperm whales in the Central California
12 Current System and the Gulf of Alaska^{36,37} relative to the monthly mean latitude of the North Pacific
13 Transition Zone. p -values reported for model II (ranged major axis; RMA) linear regression.

14

15

16 **DISCUSSION**

17

18 Animals' movement strategies shape their ecology and their ability to respond to
19 environmental perturbations. Moreover, these strategies offer a window into the
20 spatiotemporal dynamics of the ecosystems they inhabit¹. Our findings provide evidence
21 for seasonal movements by a cryptic top predator in the deep ocean, the sperm whale,
22 likely to track deep-sea resources. Below, we discuss several lines of evidence supporting

1 this conclusion and consider how these findings advance understanding of seasonal
2 movements in this population. More broadly, we discuss how these results advance
3 knowledge of phenology in the poorly understood deep ocean ecosystems in which sperm
4 whales forage.

5 The long-term acoustic detection results presented here indicate seasonality in the
6 latitudinal movements of foraging sperm whales, with greater frequency of echolocation
7 click detection in California during winter (Figure 3B; Figure S2), opposite the known
8 summer peak of detection in the Gulf of Alaska³⁶⁻³⁸ (Figure 5A). Despite this opposite
9 seasonality, foraging sperm whales are detected year-round in both locations. Based on
10 several lines of evidence, we posit that these patterns indicate a seasonal migration in
11 this population, likely driven by proximate resource tracking in an ecosystem with
12 dampened seasonality. Seasonal resource-tracking migration is the only hypothesized
13 movement strategy allowing for both year-round presence and significant seasonality in
14 presence across latitudes (Figure 2A; Figure 5B), matching empirical observations
15 (Figure 5A). Other hypothesized strategies yield either year-round presence (as in
16 nomadism) or seasonality in acoustic detection across latitudes (as in full and sex-specific
17 partial migration between distinct habitats), but do not match both of these key attributes
18 of the empirical observations (Figure 5). Additionally, if sex-specific partial seasonal
19 migration were occurring, we would expect the migratory demographic (previously
20 hypothesized to be adult males^{34,43}, with larger body sizes and higher inter-click-intervals
21 (ICI)) to drive seasonal patterns in the distribution of detected ICIs. Yet we do not observe
22 any significant seasonal shifts in the monthly distribution of detected ICIs in California,
23 instead detecting clicks consistent with female, juvenile, and adult male body sizes year-
24 round (Figure 4). We also find no relationship between monthly mean ICI and monthly
25 percent presence (Figure S4), further indicating that the seasonal pattern observed in
26 Figure 3 is not driven by adult males alone. These results are consistent with long-term
27 acoustic results from the GoA which also show year-round use of high latitudes by
28 females and juveniles³⁸. This growing body of evidence from long-term, population-level
29 observations via passive acoustics is inconsistent with the individual-sightings-based
30 hypothesis of sex-specific latitudinal segregation, likely arising from differences in the

1 scale and persistence of observation^{45,46}. Even though significant uncertainty about the
2 specific processes underlying these seasonal patterns remains, such continuous and
3 detailed deep-sea acoustic observations provide useful insights toward enhancing our
4 understanding of the phenology of sperm whale behavior and the vast and opaque
5 ecosystem they inhabit.

6 Despite seasonality in the frequency of foraging sperm whale presence, whales
7 are still detected year-round across latitudes (Figure 5A). This would be unexpected for
8 a population migrating to track proximate resources in a strongly seasonal (e.g., terrestrial
9 or surface ocean) ecosystem, but one might expect subtle population-level seasonality of
10 this nature for predators tracking resources in an ecosystem with a dampened seasonal
11 cycle. There is growing evidence that deep sea ecosystems exhibit such dampened
12 seasonality²⁶⁻²⁸, resulting from an indirect relationship with seasonal solar variation
13 mediated by organic matter falling from the directly seasonal surface ocean²³⁻²⁵. Seasonal
14 resource-tracking migration in such an ecosystem can be considered an intermediate
15 strategy between the seasonal resource-tracking movements previously studied in
16 strongly seasonal ecosystems and the nomadic resource-tracking movements found in
17 aseasonal ecosystems. Given that our simulation of nomadic resource tracking yielded
18 the second-closest match to empirical observations (Figure 5B), future work might test
19 for individual-level variation along this continuum of nomadic to strongly seasonal
20 resource tracking movements.

21 Our findings imply that sperm whales seasonally track a specific resource or
22 resource-rich habitat in the Northeast Pacific. We tested whether sperm whales'
23 acoustically inferred seasonal movements track seasonal-latitudinal patterns in the NPTZ,
24 the dominant foraging habitat which numerous surface ocean predators track in this
25 ocean basin^{16,49}. We find support for this hypothesis, with higher detection of foraging
26 sperm whales at lower latitudes when the NPTZ is at its southern extent (and vice versa;
27 Figure 6). The considerable variation around this trend likely arises from the indirect link
28 between surface biophysical processes (as measured via NPTZ latitude) and the
29 behavior of a deep-sea top predator. Nevertheless, that this top predator of the deep
30 ocean exhibits similar resource tracking behavior to that previously documented for

1 surface ocean predators¹⁶ suggests ecological links between surface and deep ocean
2 processes and seasonality. Diel vertical migration of animals between the deep and
3 surface ocean can vary seasonally in terms of depth distribution, total biomass, and
4 carbon transport^{27,62-64}. In Monterey Bay specifically, total biomass throughout the meso-
5 and epipelagic is at a minimum in spring and summer, rises in the fall, and remains
6 elevated through the winter²⁷, allowing for greater transport of biomass between surface
7 and deep waters during the seasons when foraging sperm whale detections peak in this
8 region (Figure 3B). It is important to note that we do not directly measure tracking of a
9 forage resource here, and resource-tracking migrations can also include movements to
10 track non-forage resources (e.g., predator-free habitat, favorable abiotic conditions,
11 etc.^{1,65}). Still, the intensive energetic demands of raptorial feeding at sperm whales'
12 extreme body size⁶⁶ point to forage availability as a likely motivator of their movements in
13 space and time. While our findings shed light on the likely resource-tracking seasonal-
14 scale movements of sperm whales in the Northeast Pacific, future work might explore the
15 habitat characteristics which sperm whales track at finer scale similar to previous studies
16 on southern elephant seals (*Mirounga leonine*)^{67,68}, another highly mobile mesopelagic
17 predator.

18 Seasonal resource-tracking migrations in terrestrial and epipelagic populations
19 typically evolve as a strategy to maximize resource gain in dynamic, seasonal
20 ecosystems^{1,4,11}. Interannual variability around the average seasonal-latitude patterns
21 exhibited by foraging sperm whales (Figure 3) suggests that the cues driving their
22 latitudinal movements are not fixed seasonal cues (e.g., day length), thus affording
23 flexibility to respond to environmental variation and change. Sperm whales were most
24 often detected in the CCCS during 2016 (Figure 3A), a year in which a persistent marine
25 heatwave combined with a strong El Niño to drive widespread biological impacts in both
26 the CCCS⁶⁸ and GoA⁶⁹. By exhibiting a movement strategy driven by resource tracking
27 rather than fidelity to a fixed foraging area or migratory schedule, sperm whales appear
28 to respond flexibly to interannual variability in oceanographic conditions (Figure 3A). Such
29 flexibility is often characteristic of greater resilience to environmental perturbations⁷⁰
30 including marine heatwaves⁷¹. Understanding the individual and population-level

1 outcomes of such flexibility in this sperm whale population remains an important and rich
2 area for future study.

3 While the specific cues that enable these seasonal movements remain unclear,
4 some combination of individual and social information is likely. As air-breathing predators,
5 sperm whales spend significant time in surface waters subject to seasonal variability in
6 solar irradiation, day length, and temperature. This provides a direct means of tracking
7 progression of the seasons, perhaps enabling movements influenced by spatiotemporal
8 memory similar to that observed in highly mobile epipelagic predators¹⁹. Because sperm
9 whales echolocate to find prey, long-distance acoustic information on the foraging
10 behavior of conspecifics might further direct this search, similar to the “mobile sensory
11 networks” formed by echolocating bats⁷³. Social learning of foraging and migration
12 strategies could also play a role^{74,75}, as sperm whales are highly social animals³⁴.

13 Taken together, our findings suggest that growing evidence for seasonal
14 processes in the deep ocean extend even to the seasonal movements of a top predator.
15 This study underscores the need for additional research to understand phenology across
16 trophic levels in light-limited deep pelagic ecosystems. A growing suite of technologies,
17 including remotely operated vehicles, autonomous underwater vehicles, and continuous
18 acoustic monitoring are providing an unprecedented opportunity to observe and
19 understand deep ocean ecosystems^{22,28,76}. Especially when integrated^{28,77}, these tools
20 can shed light on our murky understanding of seasonal processes and animals’ resource-
21 tracking strategies in the deep sea. In turn, we can provide more precise scientific insight
22 in support of spatiotemporally dynamic ecosystem management efforts which have to-
23 date been used on land and in the surface ocean⁷⁸, but which may be possible and
24 valuable in open and deep ocean ecosystems⁷⁹.

1 **ACKNOWLEDGEMENTS**

2

3 Thank you to Melissa Chapman and Megan McKenna for discussions which improved
4 this manuscript.

5

6 **FUNDING INFORMATION**

7

8 This research was supported by the David and Lucile Packard Foundation through the
9 Monterey Bay Aquarium Research Institute. The NSF funded installation and
10 maintenance of the MARS cabled observatory through awards 0739828 and
11 1114794. W.K.O. was supported by a postdoctoral fellowship from the David and Lucile
12 Packard Foundation through the Monterey Bay Aquarium Research Institute.

13

14 **REFERENCES**

15

- 16 1. Abrahms, B. *et al.* Emerging Perspectives on Resource Tracking and Animal Movement Ecology.
17 *Trends Ecol. Evol.* **36**, 308–320 (2021).
- 18 2. Mueller, T. *et al.* How landscape dynamics link individual- to population-level movement patterns:
19 a multispecies comparison of ungulate relocation data. *Glob. Ecol. Biogeogr.* **20**, 683–694 (2011).
- 20 3. Mueller, T. & Fagan, W. F. Search and navigation in dynamic environments – from individual
21 behaviors to population distributions. *Oikos* **117**, 654–664 (2008).
- 22 4. Teitelbaum, C. S. & Mueller, T. Beyond Migration: Causes and Consequences of Nomadic
23 Animal Movements. *Trends Ecol. Evol.* **34**, 569–581 (2019).
- 24 5. Chapman, B. B., Brönmark, C., Nilsson, J. Å., & Hansson, L. A. The ecology and evolution of
25 partial migration. *Oikos*, **120**, 1764-1775. (2011)
- 26 6. Dingle, H., & Drake, V. A. What is migration? *Bioscience*, **57**, 113-121 (2007).
- 27 7. Armstrong, J. B., Takimoto, G., Schindler, D. E., Hayes, M. M. & Kauffman, M. J. Resource
28 waves: phenological diversity enhances foraging opportunities for mobile consumers. *Ecology* **97**,
29 1099–1112 (2016).
- 30 8. Storrie, L., *et al.* Do beluga whales truly migrate? Testing a key trait of the classical migration
31 syndrome. *Mov. Ecol.*, **11**, 53 (2023).

- 1 9. Aikens, E. O. *et al.* The greenscape shapes surfing of resource waves in a large migratory
2 herbivore. *Ecol. Lett.* **20**, 741–750 (2017).
- 3 10. Deacy, W. W. *et al.* Phenological tracking associated with increased salmon consumption by
4 brown bears. *Sci. Rep.* **8**, 11008 (2018).
- 5 11. Middleton, A. D. *et al.* Green-wave surfing increases fat gain in a migratory ungulate. *Oikos* **127**,
6 1060–1068 (2018).
- 7 12. Fryxell, J. M. & Sinclair, A. R. E. Causes and consequences of migration by large herbivores.
8 *Trends Ecol. Evol.* **3**, 237–241 (1988).
- 9 13. Bastille-Rousseau, G. *et al.* Migration triggers in a large herbivore: Galápagos giant tortoises
10 navigating resource gradients on volcanoes. *Ecology* **100**, e02658 (2019).
- 11 14. Brönmark, C. *et al.* There and back again: migration in freshwater fishes. *Can. J. Zool.* **92**, 467–
12 479 (2014).
- 13 15. Lok, E. *et al.* Spatiotemporal associations between Pacific herring spawn and surf scoter spring
14 migration: evaluating a ‘silver wave’ hypothesis. *Mar. Ecol. Prog. Ser.* **457**, 139–150 (2012).
- 15 16. Block, B. A. *et al.* Tracking apex marine predator movements in a dynamic ocean. *Nature* **475**,
16 86–90 (2011).
- 17 17. Boustany, A. M., Matteson, R., Castleton, M., Farwell, C. & Block, B. A. Movements of pacific
18 bluefin tuna (*Thunnus orientalis*) in the Eastern North Pacific revealed with archival tags. *Prog.*
19 *Oceanogr.* **86**, 94–104 (2010).
- 20 18. Oestreich, W. K. *et al.* Acoustic signature reveals blue whales tune life-history transitions to
21 oceanographic conditions. *Funct. Ecol.* **36**, 882–895 (2022).
- 22 19. Abrahms, B. *et al.* Memory and resource tracking drive blue whale migrations. *Proc. Natl. Acad.*
23 *Sci. U.S.A.* **116**, 5582–5587 (2019).
- 24 20. Ryan, J. P. *et al.* Oceanic giants dance to atmospheric rhythms: Ephemeral wind-driven resource
25 tracking by blue whales. *Ecol. Lett.* **25**, 2435–2447 (2022).
- 26 21. Shuert, C. R. *et al.* Decadal migration phenology of a long-lived Arctic icon keeps pace with
27 climate change. *Proc. Natl. Acad. Sci. U.S.A.* **119**, e2121092119 (2022).
- 28 22. Robison, B. H. Deep pelagic biology. *J. Exp. Mar. Biol. Ecol.* **300**, 253–272 (2004).
- 29 23. Dall’Olmo, G., Dingle, J., Polimene, L., Brewin, R. J. W. & Claustre, H. Substantial energy input to
30 the mesopelagic ecosystem from the seasonal mixed-layer pump. *Nat. Geosci.* **9**, 820–823
31 (2016).
- 32 24. Billett, D. S. M., Lampitt, R. S., Rice, A. L., & Mantoura, R. F. C. Seasonal sedimentation of
33 phytoplankton to the deep-sea benthos. *Nature*, **302**, 520-522. (1983)
- 34 25. Lampitt, R. S., Hillier, W. R., & Challenor, P. G. Seasonal and diel variation in the open ocean
35 concentration of marine snow aggregates. *Nature*, **362**, 737-739. (1993).

- 1 26. Girard, F. *et al.* Phenology in the deep sea: seasonal and tidal feeding rhythms in a keystone
2 octocoral. *Proc. R. Soc. B* **289**, 20221033 (2022).
- 3 27. Urmy, S. S., Horne, J. K. & Barbee, D. H. Measuring the vertical distributional variability of pelagic
4 fauna in Monterey Bay. *ICES J. Mar. Sci.* **69**, 184–196 (2012).
- 5 28. Messié, M. *et al.* Coastal upwelling drives ecosystem temporal variability from the surface to the
6 abyssal seafloor. *Proc. Natl. Acad. Sci. U.S.A.* **120**, e2214567120 (2023).
- 7 29. Fais, A. *et al.* Sperm whale echolocation behaviour reveals a directed, prior-based search
8 strategy informed by prey distribution. *Behav. Ecol. Sociobiol.* **69**, 663–674 (2015).
- 9 30. Kawakami, T. A review of sperm whale food. *Sci. Rep. Whales Res. Inst.* **32**, 199–218 (1980).
- 10 31. Møhl, B., Wahlberg, M., Madsen, P. T., Heerfordt, A. & Lund, A. The monopulsed nature of sperm
11 whale clicks. *J. Acoust. Soc. Am.* **114**, 1143–1154 (2003).
- 12 32. Solsona-Berga, A., Posdaljian, N., Hildebrand, J. A. & Baumann-Pickering, S. Echolocation
13 repetition rate as a proxy to monitor population structure and dynamics of sperm whales. *Remote*
14 *Sens. Ecol. Conserv.* **8**, 827–840 (2022).
- 15 33. Gordon, J. C. D. Evaluation of a method for determining the length of sperm whales (*Physeter*
16 *catodon*) from their vocalizations. *J. Zool.* **224**, 301–314 (1991).
- 17 34. Whitehead, H. *Sperm whales: social evolution in the ocean*. (University of Chicago Press, 2003).
- 18 35. Davis, R. *et al.* Diving behavior of sperm whales in relation to behavior of a major prey species,
19 the jumbo squid, in the Gulf of California, Mexico. *Mar. Ecol. Prog. Ser.* **333**, 291–302 (2007).
- 20 36. Mellinger, D. K., Stafford, K. M. & Fox, C. G. Seasonal occurrence of sperm whale (*Physeter*
21 *macrocephalus*) sounds in the Gulf of Alaska, 1999–2001. *Mar. Mammal Sci.* **20**, 48–62 (2004).
- 22 37. Diogou, N. *et al.* Sperm whale (*Physeter macrocephalus*) acoustic ecology at Ocean Station
23 PAPA in the Gulf of Alaska – Part 1: Detectability and seasonality. *Deep Sea Res. 1 Oceanogr.*
24 *Res. Pap.* **150**, 103047 (2019).
- 25 38. Posdaljian, N. *et al.* Sperm Whales Demographics in the Gulf of Alaska and Bering Sea/Aleutian
26 Islands: An Overlooked Female Habitat. *bioRxiv* preprint. (2023).
27 <https://doi.org/10.1101/2023.04.16.537097doi>
- 28 39. Straley, J. *et al.* Depredating sperm whales in the Gulf of Alaska: local habitat use and long
29 distance movements across putative population boundaries. *Endanger. Species Res.* **24**, 125–
30 135 (2014).
- 31 40. Mizroch, S. A. & Rice, D. W. Ocean nomads: Distribution and movements of sperm whales in the
32 North Pacific shown by whaling data and Discovery marks. *Mar. Mammal Sci.* **29**, E136–E165
33 (2013).
- 34 41. Mesnick, S. L. *et al.* Sperm whale population structure in the eastern and central North Pacific
35 inferred by the use of single-nucleotide polymorphisms, microsatellites and mitochondrial DNA.
36 *Mol. Ecol. Resour.* **11**, 278–298 (2011).

- 1 42. Lefort, K. J., Hussey, N. E., Jones, J. M., Johnson, K. F. & Ferguson, S. H. Satellite-tracked
2 sperm whale migrates from the Canadian Arctic to the subtropical western North Atlantic. *Mar.*
3 *Mammal Sci.* **38**, 1242–1248 (2022).
- 4 43. Best, P. B. *Social organization in sperm whales, Physeter macrocephalus*. Springer US, (1979).
- 5 44. Pitman, R. L., Ballance, L. T., Mesnick, S. I. & Chivers, S. J. Killer Whale Predation on Sperm
6 Whales: Observations and Implications. *Mar. Mammal Sci.* **17**, 494–507 (2001).
- 7 45. Jaquet, N. How spatial and temporal scales influence understanding of Sperm Whale distribution:
8 a review. *Mammal Rev.* **26**, 51–65 (1996).
- 9 46. Levin, S. A. The Problem of Pattern and Scale in Ecology: The Robert H. MacArthur Award
10 Lecture. *Ecology* **73**, 1943–1967 (1992).
- 11 47. Oliver, R. Y. *et al.* Eavesdropping on the Arctic: Automated bioacoustics reveal dynamics in
12 songbird breeding phenology. *Sci. Adv.* **4**, eaaq1084 (2018).
- 13 48. Oestreich, W. K. *et al.* Animal-Borne Metrics Enable Acoustic Detection of Blue Whale Migration.
14 *Curr. Biol.* **30**, 4773-4779.e3 (2020).
- 15 49. Polovina, J. J., Howell, E., Kobayashi, D. R., & Seki, M. P. The transition zone chlorophyll front, a
16 dynamic global feature defining migration and forage habitat for marine resources. *Prog.*
17 *Oceanogr.*, **49**, 469-483. (2001).
- 18 50. Zhang, Y., McGill, P. R. & Ryan, J. P. Optimized design of windowed-sinc anti-aliasing filters for
19 phase-preserving decimation of hydrophone data. *J. Acoust. Soc. Am.* **151**, 2077–2084 (2022).
- 20 51. Charif, R. A., Waack, A. M., & Strickman, L. M. Raven Pro 1.4 user’s manual. (2010).
- 21 52. Wahlberg, M. The acoustic behaviour of diving sperm whales observed with a hydrophone array.
22 *J. Exp. Mar. Biol. Ecol.* **281**, 53–62 (2002).
- 23 53. Ryan, J. P. *et al.* Reduction of Low-Frequency Vessel Noise in Monterey Bay National Marine
24 Sanctuary During the COVID-19 Pandemic. *Front. Mar. Sci.* **8**, (2021).
- 25 54. Abrahms, B. *et al.* Suite of simple metrics reveals common movement syndromes across
26 vertebrate taxa. *Mov. Ecol.* **5**, 12 (2017).
- 27 55. Oestreich, W.K. Data and code for: Acoustic evidence for seasonal resource-tracking migration
28 by a top predator of the deep sea. Zenodo. <https://doi.org/10.5281/zenodo.7860426>. Deposited
29 April 24, 2023
- 30 56. Bograd, S. J., *et al.* On the seasonal and interannual migrations of the transition zone chlorophyll
31 front. *Geophysical Research Letters*, **31**, L17204. (2004).
- 32 57. R Core Team (2022). R: A language and environment for statistical computing. R Foundation for
33 Statistical Computing, Vienna, Austria. <https://www.R-project.org/>.
- 34 58. Vihtakari, M (2022). ggOceanMaps: Plot Data on Oceanographic Maps using ‘ggplot2’. R
35 package version 1.3.4. <https://CRAN.R-project.org/package=ggOceanMaps>.
- 36 59. Hijmans, R (2022). geosphere: Spherical Trigonometry. R package version 1.5-18,

- 1 <https://CRAN.R-project.org/package=geosphere>.
- 2 60. Pante, E., & Simon-Bouhet, B. marmap: a package for importing, plotting and analyzing
- 3 bathymetric and topographic data in R. *PLoS one*, **8**, e73051 (2013).
- 4 61. The MathWorks Inc. (2022). MATLAB version: 9.13.0 (R2022b), Natick, Massachusetts: The
- 5 MathWorks Inc. <https://www.mathworks.com>.
- 6 62. Hernández-León, S. et al. Large deep-sea zooplankton biomass mirrors primary production in the
- 7 global ocean. *Nat. Commun.* **11**, 6048 (2020).
- 8 63. Archibald, K. M., Siegel, D. A. & Doney, S. C. Modeling the Impact of Zooplankton Diel Vertical
- 9 Migration on the Carbon Export Flux of the Biological Pump. *Glob. Biogeochem. Cycles* **33**, 181–
- 10 199 (2019).
- 11 64. Fischer, J. & Visbeck, M. Seasonal variation of the daily zooplankton migration in the Greenland
- 12 Sea. *Deep Sea Res. Part 1 Oceanogr. Res. Pap.* **40**, 1547–1557 (1993).
- 13 65. Xu, F., & Si, Y. The frost wave hypothesis: How the environment drives autumn departure of
- 14 migratory waterfowl. *Ecol, Indic.* **101** 1018-1025 (2019).
- 15 66. Goldbogen, J. A. et al. Why whales are big but not bigger: physiological drivers and ecological
- 16 limits in the age of ocean giants. *Science* **366** 1367-1372 (2019).
- 17 67. McMahon, C. R. et al. Finding mesopelagic prey in a changing Southern Ocean. *Sci. Rep.* **9**,
- 18 19013.
- 19 68. Bailleul, F. et al. Successful foraging zones of southern elephant seals from the Kerguelen
- 20 Islands in relation to oceanographic conditions. *Philos. Trans. R. Soc. B* **362**, 2169-2181.
- 21 69. Morgan, C. A., Beckman, B. R., Weitkamp, L. A. & Fresh, K. L. Recent Ecosystem Disturbance in
- 22 the Northern California Current. *Fisheries* **44**, 465–474 (2019).
- 23 70. Walsh, J. E. et al. The high latitude marine heat wave of 2016 and its impacts on Alaska. *Bull.*
- 24 *Am. Meteorol. Soc.* **99**, S39–S43. (2018).
- 25 71. Merkle, J. A. et al. Site fidelity as a maladaptive behavior in the Anthropocene. *Front. Ecol.*
- 26 *Environ.* **20**, 187–194 (2022).
- 27 72. Cavole, L. M. et al. Biological Impacts of the 2013–2015 Warm-Water Anomaly in the Northeast
- 28 Pacific: Winners, Losers, and the Future. *Oceanography* **29**, 273–285 (2016).
- 29 73. Roeleke, M. et al. Insectivorous bats form mobile sensory networks to optimize prey localization:
- 30 The case of the common noctule bat. *Proc. Natl. Acad. Sci. U.S.A.* **119**, e2203663119 (2022).
- 31 74. Aikens, E. O., Bontekoe, I. D., Blumenstiel, L., Schlicksupp, A. & Flack, A. Viewing animal
- 32 migration through a social lens. *Trends Ecol. Evol.* **37**, 985–996 (2022).
- 33 75. Oestreich, W. K. et al. The influence of social cues on timing of animal migrations. *Nat. Ecol.*
- 34 *Evol.* **6**, 1617–1625 (2022).
- 35 76. Benoit-Bird, K. J. & Lawson, G. L. Ecological Insights from Pelagic Habitats Acquired Using
- 36 Active Acoustic Techniques. *Annu. Rev. Mar. Sci.* **8**, 463–490 (2016).

- 1 77. Urmy, S. S. & Benoit-Bird, K. J. Fear dynamically structures the ocean's pelagic zone. *Curr. Biol.*
2 **31**, 5086-5092.e3 (2021).
- 3 78. Oestreich, W. K., Chapman, M. S. & Crowder, L. B. A comparative analysis of dynamic
4 management in marine and terrestrial systems. *Front. Ecol. Environ.* **18**, 496–504 (2020).
- 5 79. Maxwell, S. M., Gjerde, K. M., Conners, M. G. & Crowder, L. B. Mobile protected areas for
6 biodiversity on the high seas. *Science* **367**, 252–254 (2020).

Supporting Information for:

Acoustic evidence for seasonal migration by a cryptic top predator of the deep sea

William K. Oestreich^{a,*}, Kelly J. Benoit-Bird^a, Briana Abrahms^b, Tetyana Margolina^c, John E. Joseph^c, Yanwu Zhang^a, Carlos A. Rueda^a, John P. Ryan^a

^a Monterey Bay Aquarium Research Institute, Moss Landing, CA 95039, USA

^b Center for Ecosystem Sentinels, Department of Biology, University of Washington, Seattle, WA 98195, USA

^c Naval Postgraduate School, Monterey, CA 93943, USA

*Corresponding author: William K. Oestreich

Email: woestreich.research@gmail.com

Supporting information on simulation of individual movement strategies

The simulation methods used here follow methods introduced by Abrahms et al. (2017)¹. Briefly, Abrahms et al. (2017)¹ generated “computer-simulated idealized movement syndromes representing suites of correlated movement traits observed across taxa”. Their paper compared these idealized syndromes and their corresponding step length and turn angle distributions to 130 individual tracks from 13 vertebrate species exhibiting a range of movement strategies (migration, nomadism, territoriality, and central place foraging) in both marine and terrestrial ecosystems. These tracks ranged in duration from several months to multiple years, and were processed to standardized, 1-hour position sampling intervals given the importance of fix rate to step length and turn angle distributions. These authors found strong statistical evidence that their simple suite of seasonal movement rules, turn angle, and step length distributions for distinct movement strategies robustly represent the movement strategies in the diverse tracking dataset to which they were compared. These suites of distributions and movement rules therefore provide a useful set of dimensionless distributions representative of distinct vertebrate movement “syndromes”.

We applied these suites of distributions and movement rules to a dimensionless simulation domain which provides a simplified arena in which to explore the patterns of detection in two geographically-distant monitoring areas under hypothesized movement strategies. This domain is not meant to specifically represent the spatial dimensions of the North Pacific, but instead provides an arena for simplified hypothesis-testing roughly analogous to the geometry of our empirical sensors in the North Pacific. Agent step lengths, hydrophone listening ranges, and domain dimensions were scaled proportionally to give agents limited probability of acoustic detection even if present at the latitude of a listening range (i.e., listening ranges covered only a proportion of both the latitudinal and longitudinal dimensions). This approach allowed for realistic probabilities of acoustic detection for a large number of individual position-days (365,000 per simulation) without the extreme computational expense of simulating a number of agents comparable to the estimated population size of sperm whales in the eastern North Pacific (~2000²).

All distributions for step lengths and turn angle, as well as seasonality and other elements of movement rules, are summarized in Table S1. These simulations were implemented in a domain with longitude ranging from -2000 to 2000 and latitude ranging from 0 to 30000 (dimensionless units). Hydrophone “listening ranges” were positioned at (0, 5000) and (0, 25000).

Nomadic resource tracking

We simulated nomadic individuals using decision rules previously documented for nomads¹: low probability ($P=0.1$) of behavioral state switching between active foraging and searching, small step lengths and uniformly-distributed turn angles during active foraging, and longer step lengths during searching with normally-distributed turn angles (around the initial direction after switching from foraging to searching).

Seasonal resource-tracking migration

We simulated movements to track resources with a shifting seasonal-latitudinal distribution using decision rules similar to those for nomadic resource tracking as described above, but with differences in movement behavior between times and locations of active foraging. Rather than searching in a single direction with turn angles normally-distributed around a randomly-selected initial search direction (as in nomads), agents in this simulation moved between active foraging periods by tracking resources with headings normally-distributed around due north and due south. The probability of northward-centered or southward-centered heading distributions during resource tracking varied seasonally to mimic seasonal shifts in latitudinal resource availability.

Seasonal migration between distinct habitats

We simulated migration between distinct habitats again using the decision rules documented by¹: four months of foraging in a southern range (steps defined by uniform step length and turn angle distributions), two months of northward migration (longer step lengths and normal turn angle distribution centered on north), four months of foraging in a northern range (steps again defined by uniform step length and turn angle distributions), and finally two months of southward migration (longer step lengths and normal turn angle distribution centered on south).

Sex-specific partial seasonal migration between distinct habitats

We simulated sex-specific partial seasonal migration by assigning 50% of agents to a migratory (male) group and 50% of agents to a resident (female and juvenile) group. Migrants followed the decision rules described above for migration between distinct habitats; residents followed the decisions rules described above for nomadic resource tracking, but only in the southern portion of the simulation domain.

Table S1. Parameters for simulations of hypothesized individual-level movement strategies, derived from ¹. † Precise migration onset and duration varies slightly by individual (sd = 5 days).

Simulated movement strategy	Behavioral states	Step length	Turn angle (°)	Behavioral transition probability	Seasonality	Demographic elements
<i>Nomadic resource tracking</i>	Active foraging	gamma distribution (shape=2, rate=2) *100	uniform distribution (min=0, max=360)	P = 0.1 (switch to searching)	N/A	N/A
	Searching	uniform distribution (min=0, max=300)	If transitioning from forage to search: uniform distribution (min=0, max=360) If continuing search: normal distribution (mean=preceding step angle, sd=60)	P = 0.1 (switch to active foraging)	N/A	N/A
<i>Seasonal resource-tracking migration</i>	Active foraging	gamma distribution (shape=2, rate=2) *100	uniform distribution (min=0, max=360)	P = 0.1 (switch to searching)	N/A	N/A
	Searching	uniform distribution (min=0, max=300)	normal distribution, either northward (mean=0, sd=10) or southward (mean=180, sd=10)	P = 0.1 (switch to active foraging)	probability of northward search varies seasonally (max of 0.8 in summer; min of 0.2 in winter)	N/A

Simulated movement strategy	Behavioral states	Step length	Turn angle (°)	Behavioral transition probability	Seasonality	Demographic elements
<i>Seasonal migration between distinct habitats</i>	Active foraging	gamma distribution (shape=2, rate=2)*100	uniform distribution (min=0, max=360)	P = 0.1 (switch to searching)	During ~4-month periods between northward & southward migrations	N/A
	Searching	uniform distribution (min=0, max=300)	If transitioning from forage to search: uniform distribution (min=0, max=360) If continuing search: normal distribution (mean=preceding step angle, sd=60)	P = 0.1 (switch to active foraging)	During ~4-month periods between northward & southward migrations	N/A
	Northward migration	constant (distance between habitats divided by ~2-month migration duration)	normal distribution (mean=0, sd=5)	N/A	~2-month period in spring-summer †	N/A
	Southward migration	constant (distance between habitats divided by ~2-month migration duration)	normal distribution (mean=180, sd=5)	N/A	~2-month period in fall-winter †	N/A

Simulated movement strategy	Behavioral states	Step length	Turn angle (°)	Behavioral transition probability	Seasonality	Demographic elements
<i>Sex-specific partial seasonal migration</i>	Active foraging	gamma distribution (shape=2, rate=2) *100	uniform distribution (min=0, max=360)	P = 0.1 (switch to searching)	During ~4-month periods between northward & southward migrations	N/A
	Searching	uniform distribution (min=0, max=300)	If transitioning from forage to search: uniform distribution (min=0, max=360) If continuing search: normal distribution (mean=preceding step angle, sd=60)	P = 0.1 (switch to active foraging)	During ~4-month periods between northward & southward migrations	N/A
	Northward migration	constant (distance between habitats divided by ~2-month migration duration)	normal distribution (mean=0, sd=5)	N/A	~2-month period in spring-summer †	Assigned 50% of population undertakes migration
	Southward migration	constant (distance between habitats divided by ~2-month migration duration)	normal distribution (mean=180, sd=5)	N/A	~2-month period in fall-winter †	Assigned 50% of population undertakes migration

Supporting information on acoustic propagation loss modeling

We modeled acoustic transmission loss for an impulsive sound source at 2.7kHz (the center frequency of the BLED), 185 dB re: 1 μ Pa at 1m (peak level of the omnidirectional low-frequency component of sperm whale echolocation clicks²), and source depths of 100, 500 and 1000m (typical of echolocation in foraging sperm whales in many ecosystems³⁻⁶), received at the location of MARS. Propagation loss was modeled for January and July to assess seasonality in click detection range. Oceanographic water column properties for the January and July model runs were calculated as the climatological mean of oceanographic conditions over the period 2016-2022 as estimated by the HYCOM (HYbrid Coordinate Ocean Model) data assimilative system⁷ with 4.8-minute spatial resolution. Acoustic propagation loss was then calculated for each of 360 1° bearings from MARS (Figure 1B) using a wave-theory parabolic equation model that accounts for absorption in both the water column and the bottom, scattering in the water column and at the surface and bottom, geometric spreading (spherical and cylindrical), refraction, and diffraction⁸. Finally, detection range for each source depth and season was estimated for each of these 360 bearings, requiring received level at MARS to exceed 5.0 dB (SNR of the click detector, Table S2) above monthly median ambient noise levels (Figure S3).

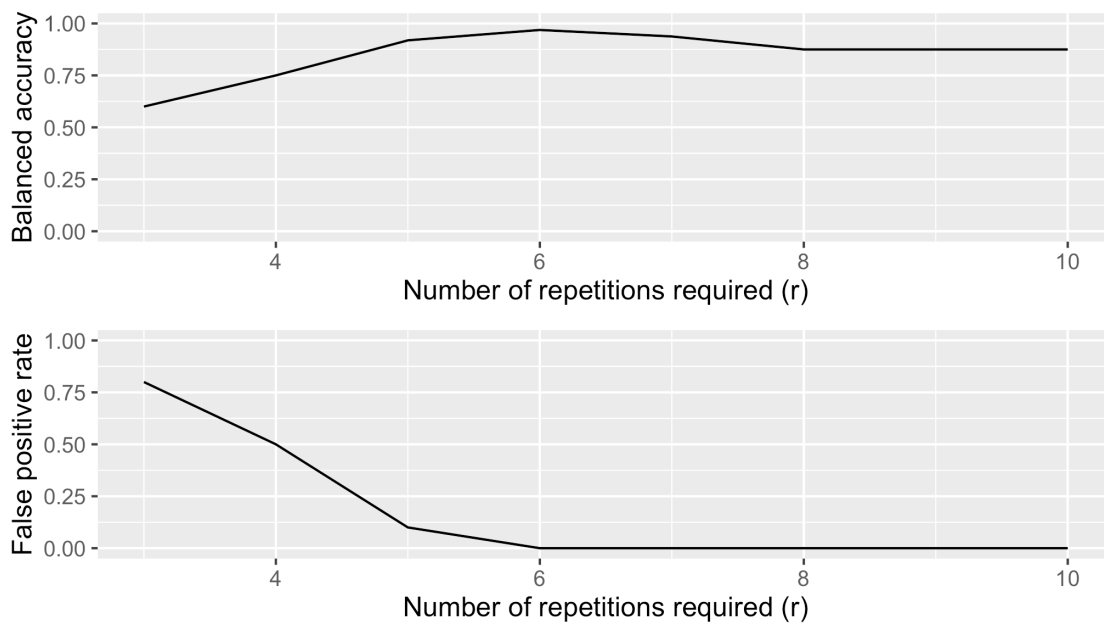


Figure S1. Performance of automated daily acoustic processing relative to manual assessment. Requiring six repetitions of click detection at near-constant inter-click interval ($r = 6$) yields a daily balanced accuracy of 97% and daily false positive rate of 0%.

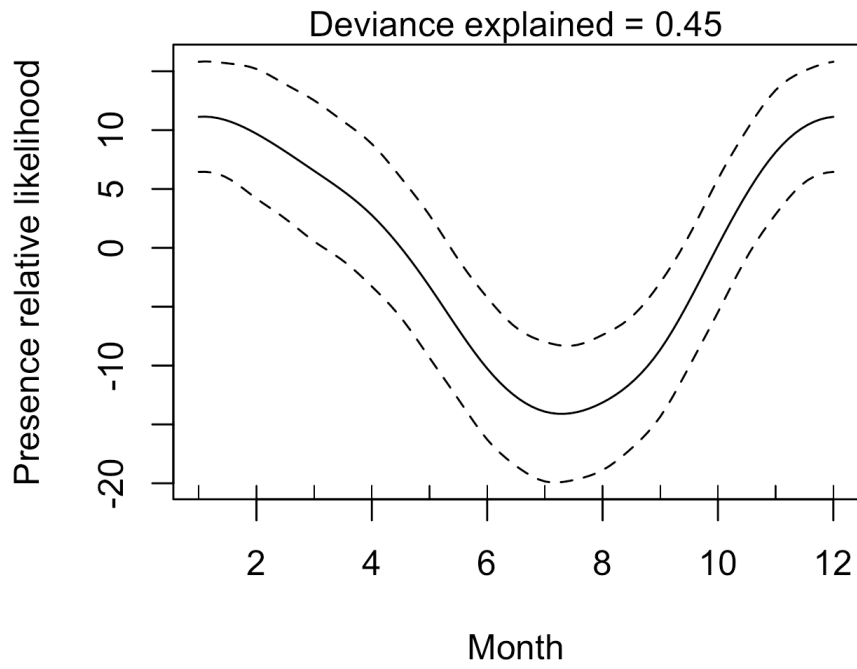


Figure S2. Generalized additive model fit relationship for monthly foraging sperm whale presence (% of days) and month, with year nested as a random effect.

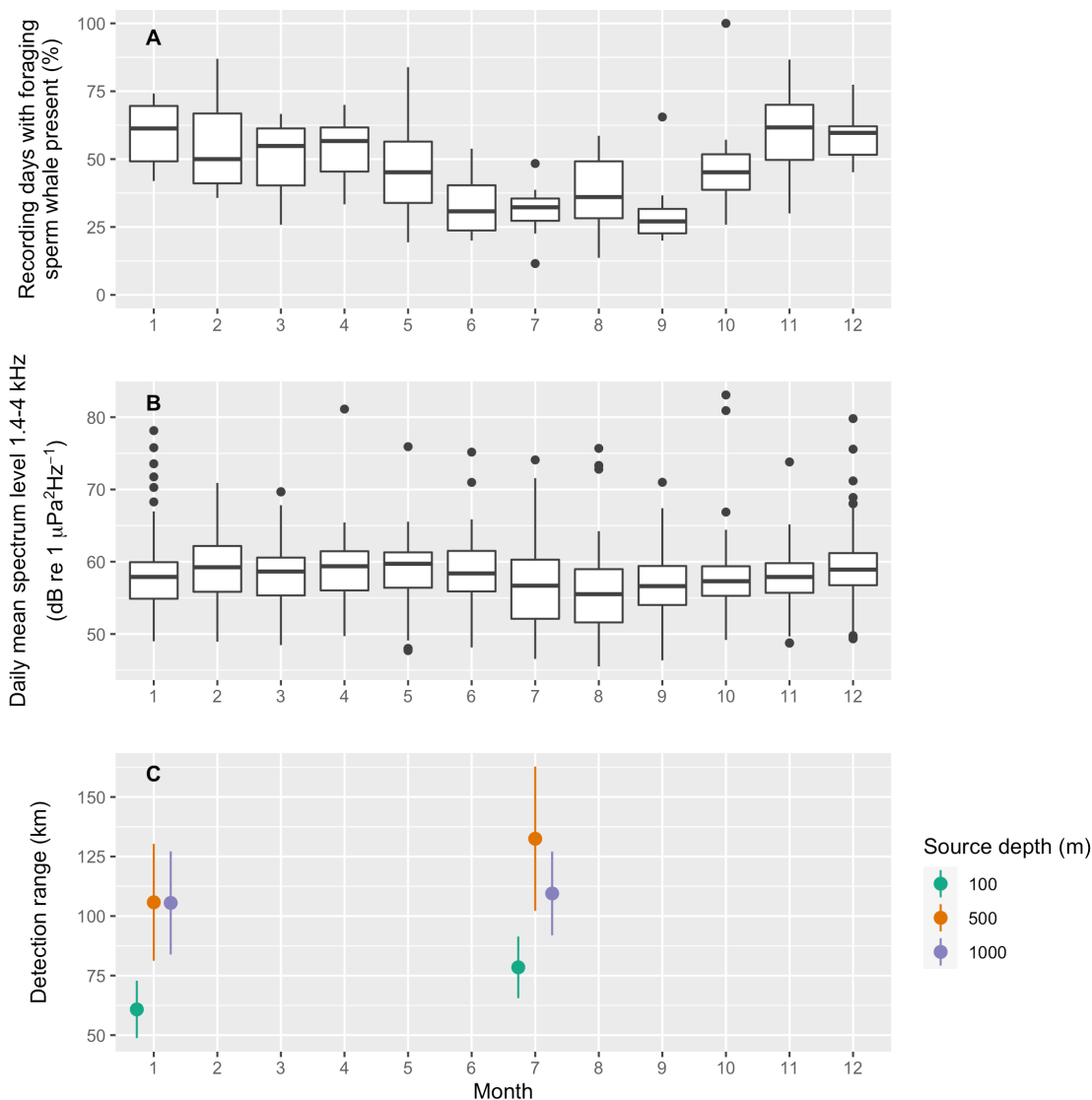


Figure S3. Seasonal variation in listening conditions at MARS. (A) Average annual cycle of echolocating sperm whale presence averaged over the full study period (Aug 2015 – Dec 2022), reproduced from Figure 3B in the main text. **(B)** Average annual cycle of ambient noise conditions at MARS in the frequency range (1.4-4kHz) targeted by the band limited energy detector employed to identify candidate sperm whale echolocation detections. **(C)** Estimated maximum detection range at MARS for sperm whale echolocation clicks produced at depths of 100, 500, and 1000m during the maximum (January) and minimum (July) months of foraging sperm whale presence. Points and lines represent the mean and standard deviation of 1-degree bearing ranges between 154-311° around MARS, representing the offshore area where 500m and 1000m source depth results are not limited by the shelf break (Figure 1B), and where sperm whales are most likely to be found. See Methods for information on modeling of acoustic propagation and detection range.

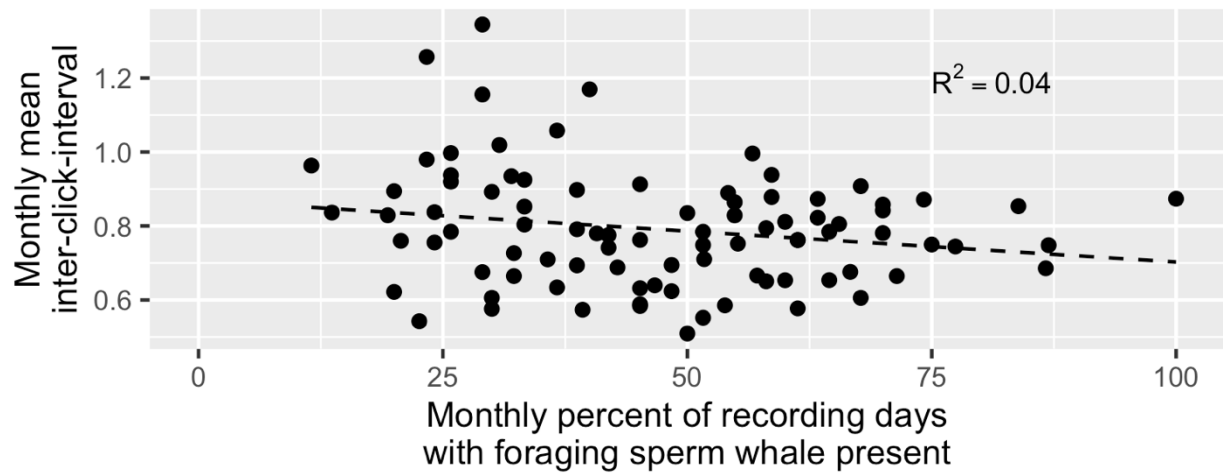


Figure S4. Additional inter-click-interval (ICI) comparison to monthly foraging sperm whale presence. Monthly mean ICI vs. monthly percent presence, indicating no significant relationship between these variables ($p > 0.05$).

Table S2. Band limited energy detector parameters.

BLED signal calculation	
Min. Frequency	1.4 kHz
Max. Frequency	4.0 kHz
Min. Duration	8.125 ms
Max. Duration	32.5 ms
Min. Separation	32.5 ms
BLED noise calculation	
Block size	2.0 s
Hop size	0.5 s
Percentile	20.0
Signal-to-noise parameters	
Min. Occupancy	70.0%
SNR Threshold	5.0 dB
Spectrogram calculation	
Window	Hann
Window Size	512 samples
Window Overlap	95%

References

1. Abrahms, B. *et al.* Suite of simple metrics reveals common movement syndromes across vertebrate taxa. *Mov. Ecol.* **5**, 12 (2017).
2. Moore, J. & Barlow, J. Improved abundance and trend estimates for sperm whales in the eastern North Pacific from Bayesian hierarchical modeling. *Endanger. Species Res.* **25**, 141–150 (2014).
3. Zimmer, W. M. X., Tyack, P. L., Johnson, M. P. & Madsen, P. T. Three-dimensional beam pattern of regular sperm whale clicks confirms bent-horn hypothesis. *J. Acoust. Soc. Am.* **117**, 1473–1485 (2005).
4. Mathias, D. *et al.* Acoustic and diving behavior of sperm whales (*Physeter macrocephalus*) during natural and depredation foraging in the Gulf of Alaska. *J. Acoust. Soc. Am.* **132**, 518–532 (2012).
5. Fais, A. *et al.* Sperm whale echolocation behaviour reveals a directed, prior-based search strategy informed by prey distribution. *Behav. Ecol. Sociobiol.* **69**, 663–674 (2015).
6. Davis, R. *et al.* Diving behavior of sperm whales in relation to behavior of a major prey species, the jumbo squid, in the Gulf of California, Mexico. *Mar. Ecol. Prog. Ser.* **333**, 291–302 (2007).
7. Chassignet, E. P. *et al.* The HYCOM (hybrid coordinate ocean model) data assimilative system. *J. Mar. Syst.* **65**, 60-83 (2007).
8. Collins, M. D. A split-step Padé solution for the parabolic equation method. *J. Acoust. Soc. Am.* **93**, 1736–1742 (1993).