Declining Abundance of Beaked Whales (Family Ziphiidae) in the California Current Large Marine Ecosystem

Jeffrey E. Moore*, Jay P. Barlow

Protected Resources Division, Southwest Fisheries Science Center, National Marine Fisheries Service, National Oceanographic and Atmospheric Administration, La Jolla, California, United States of America

Abstract

Beaked whales are among the most diverse yet least understood groups of marine mammals. A diverse set of mostly anthropogenic threats necessitates improvement in our ability to assess population status for this cryptic group. The Southwest Fisheries Science Center (NOAA) conducted six ship line-transect cetacean abundance surveys in the California Current off the contiguous western United States between 1991 and 2008. We used a Bayesian hidden-process modeling approach to estimate abundance and population trends of beaked whales using sightings data from these surveys. We also compiled records of beaked whale stranding events (3 genera, at least 8 species) on adjacent beaches from 1900 to 2012, to help assess population status of beaked whales in the northern part of the California Current. Bayesian posterior summaries for trend parameters provide strong evidence of declining beaked whale abundance in the study area. The probability of negative trend for Cuvier’s beaked whale (Ziphius cavirostris) during 1991–2008 was 0.84, with 1991 and 2008 estimates of 10771 (CV = 0.51) and ~7550 (CV = 0.55), respectively. The probability of decline for Mesoplodon spp. (pooled across species) was 0.96, with 1991 and 2008 estimates of 2206 (CV = 0.46) and 811 (CV = 0.65). The mean posterior estimates for average rate of decline were 2.9% and 7.0% per year. There was no evidence of abundance trend for Baird’s beaked whale (Berardius bairdii), for which annual abundance estimates in the survey area ranged from ~900 to 1300 (CV~1.3). Stranding data were consistent with the survey results. Causes of apparent declines are unknown. Direct impacts of fisheries (bycatch) can be ruled out, but impacts of anthropogenic sound (e.g., naval active sonar) and ecosystem change are plausible hypotheses that merit investigation.

Introduction

Beaked whales (Family Ziphiidae, Fig. 1) are one of the most diverse groups of marine mammals, comprising 21 (24%) of the 87 extant cetacean species currently recognized by the Society for Marine Mammalogy [1]. Among all marine mammal taxa, only the dolphin family (Delphinidae) is more speciose. Yet, the ecology and conservation status of ziphiids are the least understood for all marine mammal groups, owing to their deep-water oceanic existence and typically inconspicuous surface behavior. Feeding in depths often exceeding 1000 m [2], [3], most species are rarely seen; some have never been identified alive at sea and are known only from beach-stranded carcasses [4]. According to the IUCN Red List, approximately 40% of marine mammal species are considered Data Deficient [5], whereas for the Ziphiidae, 90% are Data Deficient (all except Cuvier’s beaked whale, Ziphius cavirostris, and southern bottlenose whale, Hyperoodon planifrons, which are Least Concern). Population trends for all beaked whale species are listed as unknown on the IUCN Red List.

While little is known about beaked whale ecology, and in spite of their cryptic existence, there is nevertheless a long list of documented human impacts to beaked whales [6], [7]. Beaked whales are hunted (mainly Baird’s beaked whale, Berardius bairdii), entangled unintentionally in fishing nets, affected behaviorally and physiologically by naval active sonar (sometimes with lethal effects), disturbed to an unknown extent by other ocean noise sources such as from large commercial vessels or oil and gas seismic surveys, susceptible to health effects of ingesting plastic debris, and potentially vulnerable to deepwater ecosystem changes driven by climate-related oceanographic forcing or other human impacts such as demersal fishing. Given this diverse set of mainly anthropogenic threats, there is an obvious need for improving our ability to assess beaked whale population status and impacts of anthropogenic activities.

The challenge of assessing abundance trends for rare or cryptic wildlife species is a long- standing problem in ecology [8], [9]. Cetacean abundance trends can be notoriously difficult to estimate based on monitoring programs because of typically low precision associated with individual abundance estimates [10], [11]. However, studies in terrestrial systems first demonstrated the value of using Bayesian hierarchical analyses to improve abundance trend inference by making efficient use of information contained within a time series of replicate-survey or capture-
for beach-stranded marine mammals along the western coasts of U.S. and Canada have been archived since ca. 1900 by local and regional stranding networks and museums, providing supplementary information about temporal patterns of beaked whale abundance and distribution. The sum of available information permits an analysis that may help assess conservation status of beaked whales in this part of the California Current. Beaked whale species known to occur in the study area (Figs. 1, 2) include Baird’s beaked whale (*Berardius bairdii*), Cuvier’s beaked whale (*Ziphius cavirostris*), and at least six species of the genus *Mesoplodon* that cannot be easily distinguished in the field – Hubbs’ beaked whale (*M. calihubbsi*), Blainville’s beaked whale (*M. densirostris*), Gingko-toothed beaked whale (*M. ginglyodon*), Perrin’s beaked whale (*M. perrini*), Pygmy beaked whale (*M. peruvianus*), and Stejneger’s beaked whale (*M. stejnegeri*) [15], [16]. Our analysis provides evidence of declining abundance trends for *Z. cavirostris* and for *Mesoplodon* spp. as a pooled group. To our knowledge, these are the first abundance trend assessments published for any beaked whale species. We discuss some plausible hypotheses for the apparent declines.

**Methods**

Cetacean abundance surveys

Shipboard line-transect surveys for marine mammals were conducted in the California Current by the SWFSC in summer/autumn of 1991, 1993, 1996, 2001, 2005, and 2008 (Fig. 2). The study area has been consistently divided into four strata from north to south: Oregon-Washington, Northern California, Central California, and Southern California. However, because of small sample sizes for beaked whales, the survey strata were collapsed into a single large study area for this analysis. Waters off Oregon and Washington (>28% of the study area) were not surveyed in 1991 or 1993 (implications of this discussed in Results – Sensitivity analysis). Transects followed a uniform grid pattern anchored to a different random starting point each survey year. Two observers each used mounted 25× binoculars to search for cetaceans from the flying bridge (≈15 m above the water surface) of NOAA research vessels. The vessels closed on all sightings to record group size and confirm species identification. For each sighting, group size was estimated, and a perpendicular distance to the transect line was calculated from the estimated radial distance and a measured sighting angle; various covariates associated with each detection were recorded (e.g., various visibility measures, environmental conditions). These survey methods have been used by many NOAA survey cruises in different areas of the Pacific and are well documented in the literature; for additional details, see [17], [18].

**Table 1. Survey data summary.**

| Species                  | 1991 | 1993 | 1996 | 2001 | 2005 | 2008 |
|--------------------------|------|------|------|------|------|------|
| *Berardius bairdii*      | 2    | 3    | 5    | 2    | 3    | 5    |
| *Ziphius cavirostris*    | 18   | 12   | 15   | 10   |      |      |
| *Mesoplodon* spp.        | 6    | 7    | 15   | 0    | 3    | 1    |
| unidentified ziphiid     | 0    | 3    | 3    | 2    | 4    | 4    |

Number of beaked whale groups detected, and total km of survey effort (L) in each year of cetacean line-transect surveys. Only whale groups <4 km from the transect line, and only survey effort during Beaufort sea state ≤5 are included. *Effort in these years are from three survey strata only (i.e., Oregon-Washington stratum not surveyed in these years).*
Detected and effort occurring during sea state conditions of Beaufort 0–5 were included in the analysis (although there were no detections in Beaufort 0 conditions, which rarely occur in the region). Distance data were truncated to only include observations <4 km from the transect line; this eliminated 17% (4 of 24) of B. bairdii groups and 6% (7 of 112) of groups of Z. cavirostris, Mesoplodon, and unidentified beaked whales (which belonged to either Z. cavirostris or Mesoplodon). These data truncations are consistent with recommendations for distance sampling analysis [19]. Total survey effort (on-effort transect length) and counts of beaked whale groups in the full study area are summarized in Table 1.

Analytical methods for survey data

Analytical methods generally follow those described in our previous analysis of fin whale abundance trends [14]. A brief description is provided here.

**Process and Observation Models.** Models were developed separately for B. bairdii, Z. cavirostris, a single Mesoplodon species group, and a group of unidentified ziphiids (which were either Z. cavirostris or Mesoplodon), although sighting-distance data were pooled across these groups for purposes of estimating parameters of the detection function. Recognizing that B. bairdii are more easily detectable than other ziphiids (they are larger, occur in larger groups, and have more conspicuous blows and surface behavior), the detection model included covariates for inter-species differences (see below).

Following [14], the model for each species group is partitioned into process and observation components. The process model describes how animal density \( D \) changes through time, so that abundance at time \( t \), \( N_t = D_t \cdot A \), where \( A \) is the size of the study area. The most general model we considered describes variation in animal density simply as a function of a single temporal trend parameter \( \beta_1 \) and a stochastic error component (random variable, \( \gamma_t \), for each year \( t \)). Small sample sizes precluded more complex (e.g., geographically stratified) models. If the population is changing exponentially, the full density model is:

\[
D_t = \exp(\beta_0 + \beta_1 t + \gamma_t),
\]

\( \gamma_t \sim \text{Normal}(0, \sigma) \).

The observation model links the state process to the observed data. Following line-transect sampling theory [19], and treating the observed counts of groups each year as a Poisson random variable [14]:

\[
n_t \sim \text{Pois}(E[n_t]),
\]

\[
E[n_t] = \frac{D_t}{s} \frac{2L_t g(0)}{f_t(0)},
\]

where \( n_t \) is number of groups detected; \( s \) is a single mean group size estimate for the species (there was no evidence of annual variation) with overdispersed Poisson variance (see [14]; \( f_t(0) \) is the value at distance \( y = 0 \) of \( f_t(y) \), which is the pdf of the detection probability function \( g(y) \), with \( g(0) \) being the detection probability on the transect line; and \( L_t \) is the on-effort transect length (km), considered to be measured without error (Table 1). If variance in the observed counts is over-dispersed (i.e., extra-Poisson), this should be handled implicitly by the process error term in equation 1. This can be seen by substituting the expression for \( D_t \) (eqn 1) into equation 2 and re-arranging slightly so that the error term, \( \gamma_t \), moves outside of the density term:

\[
E[n_t] = \exp(\beta_0 + \beta_1 t) \frac{2L_t g(0)}{f_t(0)} \exp(\gamma_t).
\]

Thus, we may think of \( \gamma_t \) as the sum of \( \gamma_t^p + \gamma_t^s + \gamma_t^{p+s} \), where subscripts \( p \) and \( s^+ \) refer to process error and extra-Poisson sampling error,
respectively. Estimates of $\gamma_{v,i}$ in individual years from bootstrapping methods [e.g., \cite{18}] could potentially be used to obtain more explicit estimates of process variance; this would be useful for projecting future abundance estimates with greater precision.

A more intuitive expression of equation 2 is:

$$E[n_t] = \frac{D_t}{s} - 2L_s w q_t,$$

where $w$ equals the data truncation distance (4 km in our case) and $q_t$ is the average detection probability of a group within the surveyed area $2L_s w$. Equation 3 thus indicates that the expected number of groups detected equals the group density, multiplied by the area surveyed and the average detection probability within the area surveyed, defined as $q_t = g(0)/f(0)\cdot 1/w$. In other words, $q_t$ is the “effective strip half-width” $[1/f(0)]$ divided by the total distance from the vessel within which searching takes place and corrected for imperfect detection on the trackline. The effective strip half-width is a mathematical re-interpretation of the distance function $g(0)$ into a single theoretical distance from the transect line within which groups have a detection probability of 1 and beyond which the probability is zero.

Detection probability decreases as Beaufort sea state increases. Thus the estimate of $q_t$ in equation 3 is:

$$\bar{q}_t = \frac{\sum b=1^{5} q_b L_b, t}{\sum b=1^{5} L_b, t},$$

where $L_b, t$ is the amount of survey effort in each of five Beaufort categories ($b=1$ for classes 0 and 1, 2, … 5) in year $t$, and $q_b = g_b(0)/f_b(0)\cdot 1/w$. Note, the estimate for $\bar{q}_t$ is calculated from the effort-weighted mean of the ratio $[g_b(0)/f_b(0)]$, not the ratio of the means $g_b(0)/f_b(0)$. Based on previous analyses in our case study system \cite{18} we assume a half-normal detection function for $g_b(0)$:

$$g_b(0) \sim \exp \left( -\frac{y^2}{2\sigma_{h,b}} \right),$$

where $h$ denotes half-normal parameters and the proportionality sign is used since $g(0)$ may be less than 1. We estimated the scale parameter $\sigma_{h,b}$ and hence $f_b(0)$ as a function of covariates \cite{20}, assuming the following model:

$$\sigma_{h,b} = \exp[\beta_{h0} + \beta_{h1}(h) + \beta_{h2}(\log s)],$$

where $\beta_{h0}$ is the intercept; and $\beta_{h1}$ and $\beta_{h2}$ are the coefficients for Beaufort sea state and the log of mean group size for the species, respectively (we use log of group size following convention of earlier SWFSC cetacean abundance analyses \cite{14,18}). The covariate model is based on the one used by Barlow and Forney \cite{18} for beaked whales, the main difference being that we did not include a categorical variable for the ship on which observations occurred. Preliminary analyses did not reveal this variable to have much importance on parameter estimates, while it complicated the weighted-mean estimation of $\bar{q}_t$. Species group (\textit{B. bairdii} vs. other/ smaller species) was initially considered as a covariate as well (and was included in a Sensitivity analysis – see Results), but the sample size for \textit{B. bairdii} was small (Table 1); preliminary analyses suggested that group size was a more useful variable overall and sufficiently acted as a proxy for \textit{B. bairdii} since they usually occur in larger groups. As sample sizes for \textit{B. bairdii} increase with future surveys, a separate variable for them should be included. The parameters for equation 4 were estimated from data for individual detections:

$$\sigma_{h,b} = \exp[\beta_{h0} + \beta_{h1}(h) + \beta_{h2}(\log s)],$$

where $i$ denotes each observed group (all species detections pooled).

Trackline detectability, $g_b(0)$, for \textit{Z. cavirostris} and \textit{Mesoplodon} beaked whales declines strongly with deteriorating Beaufort sea conditions. Barlow and Forney \cite{18} reported estimates of $g_b(0)$ for Beaufort states 0–1 (from \cite{21}); these account for the combination of perception bias and availability bias. Estimates for these and Beaufort 2–5 (Barlow, unpublished data) are included in Table 2. The CVs of the $g_b(0)$ estimates are based on Beaufort 0 and 1 conditions \cite{21}; this CV was used for the other sea state levels as well as there are no independent CV estimates available for states 2 and higher.

### Table 2. Estimates of trackline detection probability, $g(0)$.

| Beaufort sea state | Berardius | Ziphius | Mesoplodon |
|--------------------|-----------|---------|------------|
| 0 & 1               | 0.87*     | 0.230   | 0.450      |
| 2                  | 0.87*     | 0.148   | 0.290      |
| 3                  | 0.87*     | 0.110   | 0.215      |
| 4                  | 0.87*     | 0.043   | 0.085      |
| 5                  | 0.87*     | 0.024   | 0.048      |
| CV                 | 0.23      | 0.35    | 0.23       |

Estimates for each beaked whale genus are a function of Beaufort sea state. Estimates for sea state 0–1 and CV are from \cite{21}; Estimates for \textit{Ziphius} and \textit{Mesoplodon} in sea states 2+ are from unpublished data (J. Barlow).

*Barlow (1999) reported a point estimate of 0.96. Value reported here is the mean of Barlow’s bootstrap distribution, for compatibility with the reported CV.

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**Parameter Estimation.** Parameter estimation was conducted using a Bayesian MCMC approach in WinBUGS 1.4.3 \cite{22}, \cite{23}. Likelihoods were Poisson for the $n_i$ data, overdispersed Poisson for group size $(s_i)$ data, and truncated half-normal for the distance $(y_i)$ data. See Appendices S1 and S2 in Moore and Barlow \cite{14} for example WinBUGS code and likelihood expressions. Vague priors were used on all parameters except for $g_b(0)$. Informative Beta priors were used for $g_b(0)$ corresponding to Beaufort 0 and 1 estimates in Table 2. At each MCMC sample, $g_b(0)$ for the other sea state levels were calculated by multiplying $g_b(0)$ by a constant to preserve the $g_b(0)/g_0(0)$ ratios in Table 2. Normal priors with mean $= 0$ and large variance (e.g. 10,000) were used for most intercept and slope coefficients (e.g. $\beta$s). Positive uniform distributions (e.g., U[0, 10]) were used for standard deviations of random effects. For each model, MCMC runs consisted of two chains with a burn-in of 25,000 samples and a posterior distribution based on 75,000 samples for each chain thinned by 4 (i.e., posterior distributions constructed from 37,500 samples total); this was sufficient to achieve low Monte Carlo errors (<5% of MCMC sample standard deviation) and $R=1$ for key parameters.

**Abundance of “unidentified beaked whales”.** The abundance of the “unidentified ziphiid” group was modeled as a separate species, but these animals are believed to belong to either \textit{Mesoplodon} or \textit{Z. cavirostris}. Adult \textit{B. bairdii} are larger and with distinctive blow that make them unlikely to be confused with the
smaller genera. Therefore, \( q_1 \) for the unknown group was estimated as a weighted average of \( q_1 \) for *Mesoplodon* and *Ziphius cavirostris*, with weights at each MCMC sample given by the posterior estimates of relative abundance for these two groups. These weights were also used to proportionally attribute abundance estimates for the unidentified group to *Mesoplodon* and *Ziphius cavirostris*, thus providing revised estimates of annual abundance and trends.

**Strandings data**

Beaked whale stranding records (species, dates, and locations) dating back to 1900 were compiled from two sources: museum collections and four U.S. and Canada west coast regional stranding networks (Alaska, British Columbia, Washington-Oregon, and California). Museum records were accessed mostly through ManiS (http://manisnet.org) and Arctos (http://arctos.database.museum/home.cfm) internet data portals in July 2012 (see Table S1 for stranding data summaries and sources). Collections managers from all museums verified that these databases accurately reflected museum inventories at date of download. Regional stranding networks developed formally in the early and mid-1980s, although some of them opportunistically collected earlier records as well. There is redundancy between records from the two source types (some animals recorded by stranding networks also reside in museums); therefore, the data sets were evaluated separately, rather than combined into a single data set. These data sources do not comprise all known historical stranding records, but we consider them the most representative for describing large-scale spatio-temporal reporting patterns. Some formal stranding networks with a more local focus were established prior to regional networks (e.g., [24]), and earlier records exist from a variety of sources [15], though many of these are included in the museum or stranding network record as well.

Spatio-temporal stranding patterns can provide information about species abundance and distribution, but they also reflect patterns in detection rates, variation in ocean currents that carry carcasses to shore, changes in mortality rates, and other unquantified factors that limit ecological inference. Therefore, we did not attempt formal analysis of strandings data but simply looked for qualitative patterns that seemed obviously consistent or inconsistent with survey results. We limited our evaluation to unique stranding events rather than total numbers of stranded individuals. Thus, multiple individuals stranding together or in nearby locations (within ≈2 degrees latitude or longitude) within a one-week period constituted one event for purposes of generating and visually assessing data plots.

| Table 3. Posterior distribution summaries for coefficients of the covariate-dependent detection function (see equation 4 in text). |
|---------------------------------------------------------------|
| **Mean** | **SD** | **95% CRI** |
| \( b_0 \) [Intercept] | 1.63 | 0.37 | 0.99, 2.46 |
| \( b_1 \) [Beaufort sea state] | −0.39 | 0.10 | −0.60, −0.22 |
| \( b_2 \) [log(groupSize)] | 0.16 | 0.11 | −0.04, 0.40 |

Coefficients in the table describe the scale parameter \( (s_{bh2}) \) (equation 4 in text). For a given truncation distance \( w \) and estimate of \( g(0) \), a smaller scale parameter \( (s_{bh2}) \) implies a greater value for \( f(0) \) and thus a shorter effective strip width and smaller average detection probability \( (q) \). Thus, for example, the negative coefficient for the linear covariate for Beaufort sea state implies decreasing detectability with worsening (increasing) sea state.

**Results**

**Group size and detection**

Mean (SD) of the Bayesian posterior distributions for group sizes \( g \) across all surveys were 9.6 (8.7) for *B. bairdii*, 1.81 (0.13) for *Ziphius cavirostris*, 1.77 (0.17) for *Mesoplodon*, and 1.51 (0.20) for unidentified ziphii. The smaller mean group size for the unidentified ziphii may indicate that smaller groups in the field are less likely to be identified, or that groups not seen well enough to identify also tend to be underestimated in size, or the difference could be due to chance. The estimates for *Mesoplodon* and *Ziphius cavirostris* were slightly lower than the average of previously reported estimates using data from the same surveys [18], [25]. For both groups, the mean group size in our full dataset (Beaufort 0–5 observations) was approximately 1.8, compared to 2.0 and 2.2 for *Mesoplodon* and *Ziphius*, respectively, in the earlier studies, which used Beaufort 0–2 observations only. Differences could reflect sampling error, since we estimated group size from a larger dataset. Alternatively, group size estimates recorded in rougher seas could be biased low, driving down our estimates, although a post hoc linear regression of
group size vs. Beaufort sea state suggested this possible bias only for the *Z. cavirostris* data.

Detection probability, \( g(0) \), decreased strongly as Beaufort sea state level increased and appeared to increase some with group size, as indicated by posterior distributions for detection model coefficients in the model for \( \sigma_h \) (Table 3, Fig. 3). Average sea-state conditions and thus detection probability estimates for *Z. cavirostris* and *Mesoplodon* declined over the course of the study, and the average probability \( (g) \) of detecting a *Ziphius* or *Mesoplodon* group present within the 4-km truncation distance from the vessel was 0.03–0.05 (CV=0.36) and 0.07–0.10 (CV=0.25), respectively (Fig. 4). For *B. bairdii*, estimates of average detection probability declined slightly over the course of the study (due to trend in \( f(0) \)) but not \( g(0) \) from 0.49 in 1991 to 0.43 in 2008 (CV=0.27).

**Abundance and trends based on survey data**

Based upon analysis of four separate taxa (i.e., including unidentified ziphiids as a separate taxon), the posterior mean estimates for the trend parameters (\( \beta_1 \)) indicated annual rate of change of -4.5% per year (95% CRI: -11.5% to +2.2%) for *Z. cavirostris* and -8.4% (95% CRI: -18.9% to +0.3%) for *Mesoplodon* over the period 1991–2008. The probabilities of declining trend (i.e., \( \text{Prob}[\beta_1<0] \)) were 0.92 and 0.97, respectively. No trend was evident for *B. bairdii*, given high uncertainty in parameter estimates (posterior mean trend estimate = +0.8% per year, 95% CRI: -7.6% to +9.8%; probability of decline = 0.44).

The ‘unidentified ziphiid’ group showed evidence of increasing trend (mean trend = +5.0% annually; 95% CRI: -4.8% to +16%), with mean abundance estimates of ~700 (CV = 0.76) in 1991 and ~2000 (CV = 0.65) in 2008. One possible explanation for this is that, since observing conditions coincidentally worsened with each survey, there was an increasing trend in the number of sighted groups that could not be identified to genus. Assuming the abundance estimates for unidentified ziphiids comprised a mixture of *Z. cavirostris* and *Mesoplodon*, proportionally allocating the estimates to the two species groups and re-estimating the trend parameters weakened the evidence slightly for *Z. cavirostris* and *Mesoplodon* decline. Still, the revised annual growth rate estimate for *Z. cavirostris* was -2.9% per year (95% CRI: -8.8% to +3.3%) and for *Mesoplodon* was -7.0% (95% CRI: -16.7% to +1.0%). The revised estimates for probability of negative trend were 0.84 and 0.96, respectively. Survey abundance estimates that include prorating of individuals from the unidentified ziphiid group to *Z. cavirostris* and *Mesoplodon* are in Fig. 5 and Tables S2, S3, S4.

**Sensitivity analysis**

We conducted a *post hoc* analysis to make sure that our trend results for *Z. cavirostris* and *Mesoplodon* were not an artifact of pooling data across all survey strata even though the Oregon-Washington stratum had not been surveyed in 1991 and 1993. Conceivably, if this stratum has lower mean beaked whale densities than the California strata, this could reduce the overall density (and hence abundance) estimates for the 1996–2008 surveys, relative to the 1991 and 1993 survey estimates, leading potentially to a false trend result. Therefore, we repeated the analysis using count \( (n_i) \) and effort \( (E_i) \) and survey covariate data (e.g. Beaufort state) in all years from only the three California strata to estimate annual average detection rates \( (g) \), density, and abundance. Data from all strata were still used to estimate the detection function parameters (e.g., model for \( \sigma_h \)) and group size. We also included an indicator variable for *B. bairdii* in the detection covariate model, in case the trend estimates were sensitive in any way to how data pooling across species affects the detectability estimates.

This revised analysis did not fundamentally change our inference about trends for any species, including the unidentified group. The probabilities of declining trend for *Z. cavirostris* and *Mesoplodon*, after pro-rating the abundance and trend estimates by those of the unidentified ziphiid group, were 0.86 and 0.96, respectively, virtually identical to those in the primary analysis. Prior mean estimates for the trend parameters were actually slightly more negative in this *post hoc* analysis (-4.2% and -8.3%, respectively). Given this result, we proceed with discussion based on our primary results, to take advantage of precision and inference from the full dataset.

**Trends in stranding data**

Taken alone, patterns in the historical stranding record provide limited information about beaked whale abundance trends. However, the stranding record appears generally consistent with results of the line-transect survey analyses, thus providing secondary support for our primary analyses. Regional stranding networks originated during the early to mid-1980s, and beach coverage and reporting rates are thought to have increased throughout the 1990s and in to the early 2000s (e.g., [26], [24], pers. comm. with stranding network coordinators). Therefore, for a stable population, an overall increasing trend in stranding reports between the 1980s and 2000s might be expected. The strandings data for *B. bairdii* in the California Current are generally in line with this expectation (Fig. 6). Patterns of *Z. cavirostris*
stranding data are highly variable across stranding network regions, but an overall increasing trend from the 1980s through 2000s is not particularly evident across the California Current area, contrary to expectation if the number of reports simply tracked increasing reporting rates. Reports of Mesoplodon in the California Current area have clearly decreased over the course of three decades. For *M. carlhubbsi* and *M. stejnegeri*, the decline in reports has been monotonic (when binned by decade) since the 1980s. *M. carlhubbsi* had not been reported along the U.S. coast since 1996 until a report from Washington in 2010 (and none since). *M. pernix* is known as a species from only four stranding events in California, three between 1973 and 1979 and one in 1997. The warm-water species *M. ginkodens*, *M. densusrostris*, and *M. pernix* are represented by few records within the study area, all from California, the most recent having occurred in 2001 (*Mg*: 1954; *Md*: 1977, 1984, 1965; *Mf*: 1998, 2001).

### Discussion

#### Comparison with previous estimates

We compared our estimates with previous estimates [18], [24] obtained using data from the same surveys. The most noteworthy difference is that our estimates were much more stable from year to year. Just as one example, the previous point estimates for *B. bairdii* ranged from 0 to 1591 across years, reflecting strong sensitivity to the small number of groups observed in a particular year, whereas our estimates varied only by a few hundred individuals from year to year (Fig. 5). A key feature of the hierarchical modeling process is to treat the observations as random variables and borrow from the strength of information in the whole dataset to improve individual year estimates, shrinking them more toward the mean trend estimate in more data-poor years and thus improving the precision of each. This, along with taking a probabilistic (Bayesian) approach to inference (sensu [27]), seem to allow for better assessment of population status than would be expected based on guidelines rooted in conventional power analyses. Taylor et al. [10] noted that when using simple regression and a null hypothesis-testing paradigm with significance criteria $\alpha = 0.05$, even annual surveys (for 15 years) of a species would fail half the time to detect a 5% rate of decline when the abundance estimate CV = 0.34. Our annual CVs for $\zeta$ (2000s) were substantially higher (0.40–0.65) and the estimated rate of decline for $\zeta$ was less than 5%. Thus, our analysis suggests we can be more optimistic about our ability to assess trends using methods such as those presented here than about using conventional regression-based approaches.

On average, the population abundance estimates for $\zeta$ were substantially higher (0.40–0.65) and the estimated rate of decline for $\zeta$ was less than 5%. Thus, our analysis suggests we can be more optimistic about our ability to assess trends using methods such as those presented here than about using conventional regression-based approaches.
This summary of evidence for declining trends in beaked whale abundance is based on data collected from 1991 to 2008. The analysis suggests that abundance of *Z. cavirostris* and *Mesoplodon* spp. in the California Current has likely declined over the 18-year study period. The evidence is particularly strong for *Mesoplodon*, based on our analysis of survey-cruise data as well as the strandings record. We cannot say definitively which particular *Mesoplodon* species have driven the observed trend; however, *M. densirostris*, *M. gingkoedens*, and *M. peruvianus* are considered tropical or warm-temperate species and are rare in our cold-temperate study area, having been documented in the California Current only from a handful of strandings in southern or central California [15]. Therefore, the declining trend probably does not reflect dynamics of these species but rather of the cold-temperate more abundant species in the California Current: *M. carlhubbsi* (ranges from British Columbia to Baja California), *M. perrini* (known only from stranding records in California), and *M. stejnegeri* (a more northern species; most records are from Alaskan waters).
waters with southern California being the southern extent of its known range). Of eight known *Mesoplodon* individuals entangled in the California-based pelagic drift net fishery between 1990 and 1995, five were *M. cartilagineus*, one was *M. stejnegeri* and two were not identified to species [28].

We do not know whether the observed trends in our study area reflect actual population declines or large-scale distribution shifts to outside of the study area, but there is no obvious evidence for the latter, and given the large study area (ca. 2000 km from north to south boundary), latitudinal distribution shifts would have to be major to substantially reduce the abundance estimates. Abundance increases for long-beaked common dolphins (*Delphinus capensis*) in southern California waters [26], [29] suggest a possible recent northward range shift of warm-water cetacean species in the southern California Current, but warm-water beaked whale species have not been observed in the California stranding record since single events for *M. panus* in 1998 and 2001, and before that since *M. densirostris* in the mid-1980s. For *M. cartilagineus* and *M. stejnegeri*, the number of reported strandings along the North American west coast peaked in the 1970s or 1980s and has declined more-or-less throughout their respective ranges since then (Fig. 6). In other words, a decline in the number of reports in say, American west coast peaked in the 1970s or 1980s and has *stejnegeri* level would not change our qualitative inferences, since both beaked whales, while declining error rates at the species or genus level (i.e., non-beaked whales are not likely to have been called an observation as a particular species) seems unlikely at the family measure with our survey-based abundance analysis.

We cannot base strong conclusions on the strandings record given the overall rarity of beaked whale occurrences, unquantified variation in stranding-observer effort through time, and diverse sources of variation that can underlie true stranding patterns. We simply make the point that the stranding record does not obviously suggest northward range shifts for beaked whales and seems consistent rather than at- odds with our survey-based abundance analysis.

We can largely rule out sampling design or error artifacts as an explanation for the estimated trends. Seasonal movement dynamics of beaked whales could have changed over the study period such that the timing of our surveys coincided with higher animal abundance in the study area during the first years of the survey; however, beaked whales are not known to migrate, although seasonal movements are poorly understood [7]. Beaked whale identification accuracy by observers could have improved throughout the study period, but decreasing trends in omission error (classifying beaked whale observations as “unidentified”) would imply stronger declining trends than we estimated, because there would have been even more potential beaked whale detections during the earliest surveys. This actually seems possible because the number of “unidentified small whales” (ziphiid spp., minke whale, or *Kogia* sp.) decreased across the course of surveys [18], [24]. Decreasing commission error (e.g., incorrectly labeling an observation as a particular species) seems unlikely at the family level (i.e., non-beaked whales are not likely to have been called beaked whales), while declining error rates at the species or genus level would not change our qualitative inferences, since both *Ziphius cavirostris* and *Mesoplodon* were estimated to have declined. A final possibility we considered is that visual detectability of beaked whales could have decreased over the course of the study for behavioral reasons or other factors apart from trends in observing conditions. We cannot rule out the possibility of a trend in availability bias (i.e., decreasing g(0)); however, preliminary models that included random year effects on f(0) suggested that if anything, detection probability of “available” whales may have increased slightly through time after accounting for effects of other covariates. We have no basis for assuming behavioral changes through time related to g(0), and such a hypothesis would not explain trends in strand data. In light of all the evidence, we suggest it is prudent to be precautionary and interpret the apparent declines as real based on the best available information.

Hypotheses for beaked whale declines

Causes of the apparent declines are unknown, but we consider three hypotheses: effects of incidental mortality from fishing; impacts of anthropogenic noise, namely Navy sonar; and ecosystem changes.

**Effects of direct mortality from fishing.** Bycatch mortality of beaked whales has been reported worldwide, particularly in high-seas drift net fisheries [30]–[32]; see additional references in [28]. The California large mesh drift gillnet fishery is the only fishery known to interact with beaked whales in the California Current within the US EEZ. Based on U.S. fishery observer program data, annual bycatch mortality estimates from 1990–1995 were 0–6 for *B. bairdi*, 0–44 for *Ziphius cavirostris*, 0–29 for *Mesoplodon* whales, and 0–15 unidentified ziphiids (based on observer coverage levels of 4.4% to 17.9%; [31]). Comparing the mean 1990–1995 bycatch estimates to our 1991 and 1993 abundance estimates, the mean estimated mortality rate would have been less than 0.005 for both *Ziphius cavirostris* and *Mesoplodon*. Using the 20th percentile abundance estimates (in line with estimation of Potential Biological Removal under the US Marine Mammal Protection Act), the average bycatch mortality rate would have been as high as 0.008 for *Mesoplodon* in 1991.

Since mid-1996, acoustic pinger deterrents have been used in the California drift net fishery; this effectively eliminated beaked whale bycatch [33], [28]. A declining trend in fishing effort (Appendix 1 in [16]) and additional regulation of the fishery (reviewed in [34]) – including a large time-area closure (central California to Oregon) in effect for 4 months each year since 2001 to protect leatherback sea turtles (*Dermochelys coriacea*) – have likely reduced the potential for fishery interactions with beaked whales even further.

In summary, it seems unlikely that apparent beaked whale trends in the California Current can be explained by fishery-related mortality inside the US EEZ. Estimated bycatch during the early 1990s appears to have been low relative to abundance estimates, and bycatch of beaked whales since 1996 (inclusive) has presumably been trivial.

**Navy sonar and other anthropogenic noise.** Underwater noise has increased substantially in recent decades [35], and numerous studies and reviews have described the potential and often realized impacts posed to beaked whales and other cetaceans by anthropogenic noise. Of primary concern for beaked whales is noise caused by naval active sonar, although noise associated with varied other sources such as ship traffic and seismic exploration is a potential yet unquantified issue [36]–[42]. *Ziphius* and *Mesoplodon* are the two beaked whale genera known to suffer impacts from naval sonar activities. They exhibit strong behavioral responses to certain types of active sonar, resulting in altered movements and space use for prolonged periods after exposure (e.g., several days; [43], [41]). In more extreme cases there can be physiological consequences leading to death or stranding [44], [45], [37].

Although the threats from naval acoustic activity have been described, population-level impacts have not been quantified. Mass strandings of beaked whales throughout the Northern Hemisphere have been associated with offshore military activity, but estimates of total mortality associated with these types of impacts do not exist. Certainly they exceed levels that have been recorded, however, since the probability of observing dead whales...
California Current is generally low, especially for deep-water species [46]–[48]. Indirect impacts associated with chronic stress are even more difficult to document, although it could be hypothesized that frequent intense stressors that alter behavior and displace individuals from their habitat could reduce fitness via mechanisms such as reduced foraging efficiency, failed reproduction, increased calf mortality, etc. [42], [49].

Ambient noise off the coast of California has increased manyfold over the past several decades [50], [51], and in the Southern California Bight, beaked whales are exposed to sonar activities in the vicinity of the U.S. Navy’s Southern California Anti-Submarine Warfare Range (SOAR). However, evidence to implicate noise from naval activity or other acoustic sources as a cause of apparent beaked whale declines in the California Current is equivocal. If Navy activities at SOAR are responsible, one might expect declines to be localized to the southern California portion of the study area, or even restricted to the SOAR area, depending on beaked whale home range size and movement patterns.

Unfortunately, our data do not support a formal evaluation of spatial variation in beaked whale abundance trends; sample sizes from surveys and beach strandings are too low to evaluate abundance trends at a fine spatial scale.

Some Navy ranges support high densities of beaked whales. For example, high densities of Ziphius cavirostris occur in the SOAR area [52], and the Navy’s AUTEC sonar test facility in the Tongue of the Ocean (The Bahamas) supports the highest densities of Mesoplodon densirostris that have ever been estimated [53], [54]. High densities are not obviously consistent with a hypothesis that declines are due to military sonar, but they do not refute the possibility that declines have occurred in these areas (i.e., that densities were previously even higher). Navy ranges occurring in high-quality beaked whale habitat could also act as population sinks where sonar-habituated adults persist but recruitment is compromised through direct or indirect mechanisms. Disproportionately high frequencies of immature animals occurring in mass stranding events associated with anthropogenic activities [55] provide some albeit inconclusive support of this hypothesis. Densities of Mesoplodon densirostris in the Abaco Island area, >100 km north of the AUTEC range, appear to have remained stable from 1998–2011 [56], suggesting that, at least for this species in the Bahamas region, any potential negative effects of navy sonar may have a limited geographic reach. However, major differences in deepwater canyon bathymetry and spatial dynamics of naval operations between AUTEC and SOAR (e.g., active sonar operations in the Southern California Bight can occur well outside of SOAR) make it difficult to extend inference for Mesoplodon in the Bahamas to Mesoplodon and Ziphius cavirostris in the California Current.

Ecosystem change. Beaked whale feeding ecology is poorly known. Stomach content analyses from stranded animals suggest that many beaked whale species feed primarily on cephalopods as well as some mid-water and demersal fishes in the deep ocean. Mesoplodon whales eat smaller prey and more fish than Ziphius cavirostris, which seems to feed mainly on larger cephalopods [57]. B. bairdii feed substantially on demersal fishes, although cephalopods appear important as well [57]–[59].

Dynamics of beaked whale prey are unknown, so it is difficult to infer impacts of ecosystem change on beaked whales in the California Current. However, changes in beaked whale prey in the region may have occurred during our study period. The California Current is a highly variable system, characterized by interannual and interdecadal changes in oceanography (e.g., El Niño, Pacific Decadal Oscillation) that manifest as switches between temperature and associated biological “regimes” [60], [61]. In the middle of our study period (following the strong 1998 El Niño), the California Current switched from a “warm” phase to a “cool” phase. Bottomfish biomass has declined since the 1970s, and by 60% between 2003 and 2010, due in part to effects of demersal fishing [62], [63]. Mesopelagic fishes in the Southern California bight area have declined >60% in abundance since the 1980s in association with increasing deep-water hypoxia due to climate change [64]. Further projected expansion of oxygen minimum zones (OMZ) is expected to have significant ecosystem impacts [65]–[67]. Warming temperatures and OMZ expansion have facilitated invasion of the California Current by jumbo squid (Dosidicus gigas) since the late 1990s. D. gigas is an abundant, large-bodied (up to 50 kg), aggressive, schooling, hypoxia-tolerant generalist predator of fishes and cephalopods that may have played a role in declines of Pacific hake (Merluccius productus) [68], [69], [66]. How deepwater cephalopod communities have responded along with all these factors, either directly or by trophic-mediated pathways, is unknown, but it would be surprising if they have been unaffected. Meanwhile, competition for cephalopod prey between beaked whales and other species could be increasing. Prey species found in stomachs of M. pantheri and M. carlhubbsi in the California Current [70], [71], [58] are also consumed by sperm whales ( Physeter macrocephalus) [72], jumbo squid [68], and northern elephant seals (Mirounga angustirostris) [73], which breed on islands off the coasts of Baja California, Mexico, and southern and central California. Following commercial hunting that depleted elephant seal abundance to about 100 by 1890 [74], seals have recovered to over 170,000 as of 2005 [75]. Numbers in U.S. waters have increased roughly 6-fold since the late 1970s and nearly doubled during the course of our study period (M. Lowry, SWFSC, unpubl. data). Ultimately, we lack the information necessary to assess the impacts of ecosystem change and trophic dynamics on beaked whale populations, but additional research into these questions is merited.

Summary and research recommendations

The abundance of Ziphius and especially Mesoplodon beaked whales appears to have declined in the California Current since the early 1990s. This inference was made possible through a Bayesian hierarchical modeling approach. Drivers of apparent population declines are unknown, although direct fisheries (bycatch) impacts can probably be ruled out. Impacts from anthropogenic noise and human-mediated or other ecosystem change are plausible explanations, but additional research is required to more thoroughly evaluate these hypotheses.

Dedicated survey effort to estimate trends in the Navy SOAR area of the Southern California Bight and in additional control areas would help test hypotheses concerning the effects of naval sonar on trends. Comparisons of population age structure based on mark-resight data would also be insightful, while data on individual movement patterns would provide complementary information about the potential geographic reach of local impacts at SOAR to other areas of the system. Hypotheses related to ecosystem change could possibly be evaluated through dedicated surveys in areas differently affected by deepwater oxygen depletion, demersal fishing, jumbo squid range expansion, etc., combined with research on spatial or temporal variation in beaked whale diets (e.g., via stomach content and/or stable isotope analysis of free-ranging or stranded animals), prey abundance, and food web modeling. Increased use of acoustic methods to improve the amount of abundance information collected during surveys would be valuable. Additional large-scale surveys (especially augmented by acoustic data collection) will be useful for increasing sample sizes and the length of the time series to eventually permit geographically stratified analysis (ideally in relation to large-scale
variation in ecosystem characteristics). This would facilitate our ability to explore factors associated with apparent beaked whale declines.

Supporting Information

Table S1 Data sources for beaked whale stranding records from Alaska, British Columbia, Washington, Oregon, and California. Records in table indicate number of individual animals. These were reduced to the number of unique events (i.e., a group of animals 1 event) for analysis in main paper.

Table S2 Final abundance estimates (Bayesian posterior summaries) for Cuvier’s beaked whale (Ziphius cavirostris) in the California Current study area. Estimates include pro-rated allocation of unidentified beaked whales to this species.

Table S3 Final abundance estimates (Bayesian posterior summaries) for Mesoplodon beaked whales in the California Current study area. Estimates include pro-rated allocation of unidentified beaked whales to this species group.

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Table S4 Final abundance estimates (Bayesian posterior summaries) for Baird’s beaked whales (Berardius bairdii) in the California Current study area.

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Author Contributions

Conceived and designed the experiments: JPB. Performed the experiments: JPB. Analyzed the data: JEM. Contributed reagents/materials/analysis tools: JEM, JPB. Wrote the paper: JEM.
