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The importance of tropical tuna for seabird foraging over a marine
 productivity gradient

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## 11 Abstract

Foraging with tuna is a well-documented seabird strategy, referred to as facilitated foraging. 12 However, despite this behaviour being considered almost obligatory in nutrient-poor tropical 13 waters, little data exist on its relative importance to individual colonies. Therefore, to examine 14 facilitated foraging under different patterns of nutrient availability we tracked Wedge-tailed 15 Shearwaters Ardenna pacifica from two colonies, one tropical and one subtropical, situated in 16 waters of contrasting productivity. Shearwater foraging behaviour was assessed relative to 17 oceanographic covariates and predicted distributions for multiple tropical tuna species and age-18 classes, simulated by an existing ecosystem model (SEAPODYM). Shearwaters from both colonies 19 undertook long-trips to deep, pelagic waters close to seamounts and foraged most often at fronts and 20 eddies. Micronektonic and adult tuna age-classes were highly correlated in space. Predation 21 between these tuna age-classes represents a likely source of facilitated foraging opportunities for 22 shearwaters. At broad-scales, shearwaters consistently foraged in areas with higher predicted adult 23

skipjack and micronektonic tuna densities and avoided adult bigeye tuna. At finer-scales, dynamic 24 ocean features aggregated tuna of all sizes. Enhanced tuna density at these locations increased the 25 likelihood of shearwater foraging activity. Long-trips in the tropics targeted oligotrophic waters 26 with higher tuna densities. Long-trips in the subtropics targeted enhanced productivity, but in some 27 years shifted to target the same oligotrophic, tuna-dense waters used by tropical conspecifics. We 28 conclude that facilitated foraging with tuna is consistently important to the tropical breeding 29 population and becomes increasingly important to the subtropical population in years of low marine 30 31 productivity.

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33 Key words: facilitated foraging, micronekton, productivity, SEAPODYM, tuna, Wedge-tailed
34 Shearwater

## 35 Introduction

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Tropical waters are generally considered less productive than their temperate counterparts 37 (Longhurst & Pauly 1987). In temperate and polar waters, predictable physical oceanographic 38 processes e.g. fronts, upwellings, ice and shelf edges reliably aggregate seabird prey, whereas in 39 tropical waters seabird prey are scarcer and more patchily distributed (Ainley & Boekelheide 1983, 40 Ballance & Pitman 1999. Weimerskirch 2007). Despite such constraints, abundant communities of 41 seabirds exist in tropical waters (King 1974). For tropical seabirds the single most important 42 foraging strategy, believed to overcome poor prey predictability, is to feed in multi-species flocks in 43 association with sub-surface predators, primarily tunas (Au & Pitman 1986, Ballance & Pitman 44 1999, Spear et al. 2007). Sub-surface predators are believed to be crucial for driving prey upwards 45 and making them available to surface feeding seabirds (facilitated foraging) (Ashmole & Ashmole 46 1967, Clua & Grosvalet 2001). Although facilitated foraging occurs in polar (Thiebot & 47 Weimerskirch 2013), temperate (Goyert et al. 2014) and sub-tropical waters (Clua & Grosvalet 48 2001, Vaughn et al. 2008), the level of seabird community reliance on sub-surface predators is 49 unparalleled in the tropics (Ballance & Pitman 1999, Spear et al. 2007). 50

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However, the difficulty associated with monitoring seabird and sub-surface predator activity concurrently means that facilitated foraging is often inferred. Inference comes from overlap in seabird and sub-surface predator prey items (Ashmole & Ashmole 1967, Ménard et al. 2013) and trophic niches (Kojadinovic et al. 2008, Young et al. 2010a), although often the results of such studies are inconclusive. Facilitated foraging has also been inferred, but not quantified, from overlap between seabird foraging tracks and areas of high sub-surface predator activity in commercial fisheries (Catry et al. 2009, Weimerskirch et al. 2010, McDuie & Congdon 2016).

Finally, oceanographic data has shown that the densities of tropical, diurnal, piscivorous seabirds
are driven by a well-stratified, deep thermocline, which is associated with higher sub-surface
predator densities (Ballance et al. 1997, Spear et al. 2001).

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The bulk of our knowledge on facilitated foraging comes from at-sea observations (Au & Pitman 63 1986, Spear et al. 2007). At-sea observations are able to quantify instances of seabirds using 64 facilitated foraging, but have their limitations. The cost of vessel hire has led some studies to make 65 observations from fishing vessels. However as these vessels target sub-surface predators, results are 66 biased towards overestimation of facilitated foraging events (Jaquemet et al. 2004, Hebshi et al. 67 2008). At-sea observations using transects are unbiased and have contributed significantly to our 68 understanding of facilitated foraging, primarily in the eastern and central tropical Pacific (Au & 69 Pitman 1986, Spear et al. 2007) and tropical Indian oceans (Thiebot & Weimerskirch 2013). 70 However, all at-sea survey methods are limited by the prohibitive cost of undertaking simultaneous 71 surveys in different regions and an inability to determine the provenance of birds being observed. 72 As such, no previously available method has been able to quantify the individual, or population-73 level decisions of seabirds on whether to preferentially target sub-surface predators or not. 74 Consequently, there has been no way to determine the relative importance of facilitated foraging 75 opportunities for specific seabird colonies. Our current study provides a framework to do so. 76

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Limited availability of suitable nesting habitat can lead to seabird colonies being located where local marine resources are sub-optimal (Navarro & González-Solís 2009). Under these constraints many pelagic foraging seabirds use a bimodal foraging strategy, where 'short-trips' (1-3 days in shearwaters; Baduini & Hyrenbach 2003) in resource-poor local waters are used almost exclusively to provision chicks at the expense of adult condition. Following a series of short-trips adults

undertake a 'long-trip' (5-17 days) to more distant foraging sites where they can quickly regain 83 condition (Weimerskirch 1998, Weimerskirch & Cherel 1998). To achieve this, long-trips are said 84 to access 'productive distant waters' (Weimerskirch 1998). As most bimodal foraging studies come 85 from temperate and polar regions 'productive' has become synonymous for high primary 86 productivity or chlorophyll-a concentration, aggregated by shelf or frontal features (Waugh et al. 87 1999, Catard et al. 2000, Klomp & Schultz 2000, Stahl & Sagar 2000, reviewed in Baduini & 88 Hyrenbach 2003). Seabirds breeding in the subtropics may have the opportunity to access similar 89 large-scale areas of high primary productivity using bimodal foraging (Paiva et al. 2010). However, 90 since such features are rare in tropical systems, we hypothesise that tropical-breeding seabirds target 91 increased sub-surface predator densities as an alternative, because of the greater facilitated foraging 92 opportunities this provides. 93

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The wedge-tailed shearwater is a tropical seabird known to associate heavily with tuna when 95 foraging in multiple regions (Au & Pitman 1986, Jaquemet et al. 2004, Hebshi et al. 2008), and has 96 been shown to capture most of its prey through facilitated foraging (Spear et al. 2007). The wedge-97 tailed shearwater populations breeding off eastern Australia adopt a bimodal foraging strategy in 98 tropical waters of the Coral Sea (Congdon et al. 2005) and a more unimodal strategy in sub-tropical 99 waters of the Tasman Sea, although some long-trips are made (Peck & Congdon 2005). East 100 Australian wedge-tailed shearwater populations have access to a range of marine habitats (shelf, 101 102 seamount and frontal systems; Hobday et al. 2011) and sub-surface predator populations in the region include numerous tuna species (Young et al. 2010b). 103

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105 This study aims to estimate the relative importance of facilitated foraging with tropical tuna for two 106 wedge-tailed shearwater populations in waters with contrasting productivity. Relative importance is

quantified from spatial association between shearwater tracking data, oceanographic covariates and 107 modelled tropical tuna distributions. We tested two facilitated foraging hypotheses at opposing 108 spatio-temporal scales, thus expecting that: 1) tropical tuna distributions influence the selection of 109 wedge-tailed shearwater core-area locations (defined by the 50% Utilization Distribution (UD) from 110 kernel analysis) at broad scales; and 2) tropical tuna distributions influence the likelihood of wedge-111 tailed shearwater foraging activity at fine scales. We additionally test two facilitated foraging 112 hypotheses at opposing ends of a productivity gradient, expecting that: 3) wedge-tailed shearwater 113 in tropical waters do not adhere to the temperate model of 'productive' long-trip destinations and 114 instead target tuna; and 4) wedge-tailed shearwater breeding in sub-tropical waters adhere to the 115 temperate model of 'productive' long-trip destinations and do not target tuna. 116

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## 118 Methods

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#### 120 Study area and logger deployment

This study was carried out at Heron Island (23°26' S, 151°51' E) in the Capricorn Bunker Island 121 Group, Great Barrier Reef and Lord Howe Island (31°33' S, 159°05' E) in the northern Tasman 122 Sea, Australia. Both islands support large breeding colonies of wedge-tailed shearwaters (Marchant 123 & Higgins 1990). We deployed GPS loggers on wedge-tailed shearwaters at Lord Howe Island in 124 2014, 2015 and 2016 and concurrently at Heron Island in 2015, and deployed Platform Terminal 125 Transmitter (PTT) loggers at Heron Island in 2011 and 2013 (McDuie et al. 2015). All loggers were 126 deployed during the chick-rearing period (Feb-April). I-gotU GT-120 GPS loggers (Mobile Action 127 Technology) were modified to use smaller 100mAh batteries, sealed in heat-shrink tubing and 128 programmed to obtain fixes every 10 minutes (Freeman et al. 2013). Solar-powered ARGOS PTTs 129 130 (PTT-100; Microwave Telemetry, USA) relay data via satellite and were programmed to obtain

fixes continuously. We deployed both GPS and PTT loggers to 3 central tail feathers using Tesa®
4651 Tape, total deployment weight of both logger types was ~ 10.5-12 g, within the 3-5% body
weight limit for the species (McDuie et al. 2015).

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#### 135 Tracking data preparation and analyses

All data handling and statistical analyses were performed in the statistical software environment 136 program R, version 3.2.4 (R Core Team 2016). Tracking data were pre-processed prior to analyses 137 using a speed filter, removing points exceeding a maximum velocity of 50 km/h (McDuie et al. 138 2015), additionally GPS tracks were gap filled using interpolation to 10 minute interval 139 (Weimerskirch et al. 2006). As loggers were active for several days multiple foraging trips were 140 observed for most individuals. To split long-trips from short-trips within multi-day GPS tracks we 141 used the R 'tripsplit' function from the 'marine IBA' package (Lascelles et al. 2016) and isolated 142 long-trips (>4 days duration; Congdon et al. 2005) for further analysis. We could not determine 143 individual foraging trips within multi-day PTT tracks (due to near colony positional error) so 144 removed all fixes from tracks within the 'short-trip zone' of 300 km around Heron Island (McDuie 145 et al. 2015). 146

To identify core-areas used by each colony for each year of long-trip tracking data, we employed 147 kernel analysis using the package 'adehabitatHR' (Calenge 2006). All locations within each colony-148 year combination were used with a grid size of 0.5 km and, for consistency, the same smoothing 149 parameter (h) of 20 km for GPS and PTT data following McDuie et al. (2015). The 50% UD was 150 selected from resultant kernels to represent the core-area used by wedge-tailed shearwaters in each 151 year (Hamer et al. 2007). To identify behavioural states and thereby identify foraging locations, we 152 applied Hidden Markov Models (HMM) to the GPS data. We constructed a single HMM using the 153 full GPS tracking dataset, including an identifier for each trip, using the package 'moveHMM' 154

(Michelot et al. 2016). For each consecutive GPS point the step length and turning angle were calculated, producing three distributions consistent with foraging, resting and transiting behaviours observed in HMM studies of boobies (Oppel et al. 2015) and shearwaters (Dean et al. 2012). The fitted HMM was then used to classify each GPS point as either foraging, resting or transiting.

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#### 160 Oceanographic data

We created a static covariate for depth (BTY) from the 30 arc-second General Bathymetric Chart of 161 the World (GEBCO) and calculated a static covariate for distance to seamount using the Global 162 Seamount Database (Kim & Wessel 2011) and package 'raster' (Hijmans 2016). Gridded 163 oceanographic data were accessed from NOAA ERDDAP servers using the 'rerddap' package 164 (Chamberlain 2016) (Table 1). We accessed oceanographic climatologies in the form of long term, 165 monthly averages: primary productivity (PRO), 1997-2016; sea surface temperature (SST), 1985-166 2016. We used the March data product as our covariate as it is central to the wedge-tailed 167 shearwater chick-rearing period. We accessed very fine temporal scale Ekman upwelling (EKM, 168 1d) and sea surface height anomaly (SSHA, 1d) data as these products were cloud free. We also 169 accessed coarser scaled data on Chlorophyll-a concentration (CHL, 8d), sea surface temperature 170 171 (SST, 1-8d) and sea surface temperature anomaly (SSTA, 1-8d), to counter daily missing values due to cloud cover. In a further step to fill data gaps due to clouds, we blended two CHL, SST and 172 SSTA products from different data sources to create single covariates (Table 1). 173

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#### 175 **Tropical tuna data**

Tuna covariates were generated using the Spatial Ecosystem and Population Dynamics Model (SEAPODYM). SEAPODYM is an age-structured population model describing spatial and temporal dynamics of top predator species in a direct link with prey biomass and environmental

variability (Lehodev et al. 2008). SEAPODYM uses a sub-model to predict the distribution of prev. 179 which provides a habitat-quality index for tuna that varies by age-class and species (Lehodey et al. 180 2010). This habitat is based on the distribution of simulated micronekton, which we here define as 181 mobile and free-swimming macro-zooplankton, fish and squid species between ~10 mm and ~250 182 mm. Micronekton are classified into functional groups by their vertical habitat and diel migration 183 pattern, with the spatio-temporal transfer of energy between them described using allometric scaling 184 equations and ocean currents. The top predator model describes the age-structured spatial 185 population of tunas across four distinct life stages (termed larval, juvenile, immature young, and 186 mature adult) and includes anthropogenic forcing in the form of effort and catch from multiple 187 fisheries. Driving both models are ocean biophysical variables (temperature, currents, oxygen, and 188 primary production) that characterise the marine environment of predator dynamics (Lehodey & 189 Senina 2009) and age-dependent accessibility functions describing the sub-model micronekton 190 biomass available to top predators. SEAPODYM solves these internal models using advection-191 diffusion-reaction equations over a network of regularly spaced grid points and a discrete time step 192 (e.g. 1° square x 1 month), outputting predictions on the spatial dynamics of large pelagic predators 193 (Lehodey et al. 2008, Senina et al. 2008). SEAPODYM can be optimised and parameterised for 194 195 different marine predator species and regions (Abecassis et al. 2013, Dragon et al. 2014) or different exercises, such as climate change prediction (Senina et al. 2016). 196

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For our region SEAPODYM predictions were available on the distribution of biomass (g m<sup>-3</sup>) for three tuna species: yellowfin tuna *Thunnus albacares* (YFT), bigeye tuna *T. obesus* (BET) and, skipjack tuna *Katsuwonus pelamis* (SKJ). For each species we selected distributions of adult (ADU) and micronektonic (MIC) age-classes which vary spatially due to their differing access to prey, behaviour and density-dependent mortality representing cannibalism (Lehodey et al. 2008). Adult

tuna are not shearwater prey. Therefore, we assume associations between wedge-tailed shearwaters 203 and adult tuna represent facilitated foraging. Micronektonic tuna are between 1-3 months old with 204 fork lengths from ~30 mm up to ~100 mm (SKJ) and ~250 mm (YFT and BET) (Davies et al. 2014, 205 Harley et al. 2014, Langley et al. 2014), wedge-tailed shearwater association with micronektonic 206 tuna could represent direct predation of smaller individuals or facilitated foraging with larger 207 individuals. We also selected SEAPODYM predictions under fished conditions to represent real-life 208 tuna distributions, such as heavier long-lining effort closer to the Australian east coast (Trebilco et 209 al. 2010). 210

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To test our hypotheses we compiled two datasets at opposing spatio-temporal scales, the first was 212 built with broad scale, decadal-averaged data (hereafter termed the climatology dataset/model), the 213 214 second built with fine scale, weekly-averaged data (defined as high resolution and hereafter abbreviated as the 'hi-res' dataset/model; Appendix 1). The climatology dataset included BTY, 215 SMT, PRO, SST and tuna covariates from INTERIM parameterised SEAPODYM predictions (1° x 216 1 month), for the month of March (SKJ: 1979-2010 average, BET and YFT: 1986-2010 average) 217 (Table 1). The hi-res dataset included BTY, SMT, CHL, SST, SSTA, SSHA, EKM and tuna 218 covariates from INDESO V2 parameterised SEAPODYM predictions (0.25° x 1 week). Tuna data 219 were not available for 2016, so hi-res models were constructed for 2014 and 2015 only. 220

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#### 222 Covariate extraction

To standardise climatology model covariate extraction we used a 0.1 degree grid, taking all pixels within the 50% UD core-areas as presence and generating pseudo-absence pixels, defined as locations where absence is probable but uncertain, for logistical regression. Pseudo-absence pixels were randomly generated, at a rate of random 3:1 presence pixels, within a hypothetical maximum

foraging range for each colony. The range was set at 1400 km from each colony (maximum distance observed in our data; Heron Island 2013), refined by removing land and areas beyond the species range e.g. south of the sub-tropical front (del Hoyo et al. 1992). For hi-res models we reclassified behaviour-classed tracking data for logistic regression, assigning foraging and resting locations as presences, as tropical seabirds can 'drift forage' (using a surface 'sit-and-wait' strategy; Conners et al. 2015), and transiting locations as absences. We extracted values from covariates for climatology and hi-res model locations using the package 'raster' (Hijmans 2016).

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#### 235 Multicollinearity and spatial autocorrelation

Collinearity between covariates is an ever present issue in regression-type analyses of ecological 236 data (Dormann et al. 2013), and continually persistent in marine habitat modelling exercises 237 (Goyert et al. 2014, Lavers et al. 2014, McDuie & Congdon 2016). We explored the climatology 238 and hi-res covariate datasets for collinearity using pairwise Pearson's correlations with scatterplots 239 of covariates (Zuur et al. 2010). We identified significant collinearity between covariates (Pearson's 240  $r^2 > 0.9$ ), particularly around SST and tuna covariates. To understand the sources of collinearity, 241 identify correlated clusters and select proxy covariates to use in models, we standardised our 242 243 covariates and carried out Principal Components Analyses (PCA) in the package 'vegan' (Oksanen et al. 2016). To help interpret PCA ordination plots, Pearson's R<sup>2</sup> values were calculated between 244 the covariates and the principal components (Quinn & Keough 2001). For each dataset we aimed to 245 retain covariates for modelling that had a pairwise Pearson's  $R^2 < 0.5$  with others and selected a 246 single tuna covariate that minimised correlation with non-tuna covariates, to act as a proxy for all 247 covariates correlated within tuna clusters (Dormann et al. 2013). 248

Spatial autocorrelation (SAC) is another issue inherent in species distributional data, and failure to 250 account for it can result in non-independence of model residuals causing bias in parameter estimates 251 and increasing type I errors (Dormann et al. 2007). We checked Pearson's residuals from 252 climatology and hi-res models for SAC using the package 'ncf' (Bjornstad 2016), calculating 253 Moran's I values over distances: 1-2500 km (climatology models) and 1-1000 km (hi-res models). 254 Climatology models showed high levels of SAC (Moran's I > 0.9), to accommodate the spatial 255 structure we calculated an autocovariate term over the 50 nearest neighbours following Bardos et al. 256 (2015), inclusion of this term reduced model SAC to acceptable levels (Moran's I  $\leq$  0.22) (McDuie 257 & Congdon 2016). hi-res models showed lower, but still present, levels of SAC (Moran's I < 0.4), 258 we reduced SAC in this tracking dataset by subsampling locations to every third point, resulting in 259 models with Moran's I  $\leq$  1.5 (Perotto-Baldivieso et al. 2012). 260

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#### 262 Statistical modelling

To test the influence of different broad scale oceanographic and tuna covariates on selection of 263 wedge-tailed shearwater core-area location, we used generalized linear models (GLM). GLMs had a 264 binary response, treating core-area locations as 1 and total foraging range pseudo-absences as 0, and 265 were constructed separately for Lord Howe and Heron Island. To test the effect of different high-266 resolution oceanographic and tuna covariates on likelihood of wedge-tailed shearwater foraging, we 267 used generalized linear mixed models (GLMM), fitted in package 'lme4' (Bates et al. 2015), with 268 bird identity as the random intercept (Hamer et al. 2007, Grecian et al. 2016). GLMMs had a binary 269 response, treating foraging or resting locations as 1 and transiting locations as 0, and were 270 constructed separately for each year and colony (Lord Howe Island 2014, 2015 and Heron Island 271 2015). To effectively accommodate non-linearity, but not over-fit relationships, we permitted 272 covariates in hi-res models to take either a linear or second-degree polynomial form. Model 273

residuals and diagnostics were plotted and checked as per Zuur et al. (2009) and models' terms were 274 selected using both forwards and backwards selection based on likelihood ratio tests ( $\chi^2$ ) and 275 confidence intervals (Bolker et al. 2009). Model explanatory power was evaluated by constructing 276 receiver operating characteristic (ROC) curves (Hanley & McNeil 1982) and calculating the 277 associated area under the ROC curve (AUC) in package 'verification' (NCAR 2015). Goodness of 278 fit was assessed for GLM using McFadden's pseudo R<sup>2</sup> (Azen & Traxel 2009) in package 'pscl' 279 (Jackman 2015) and assessed for GLMM split into marginal (var. explained by fixed effects) and 280 conditional (var. explained by fixed + random effects) pseudo  $R^2$  components (Nakagawa & 281 Schielzeth 2013) in package 'MuMIn' (Bartoń 2016). 282

283

### 284 **Results**

285

#### 286 Tracking data

In total, 62 long-trips were recorded during the study. GPS battery life allowed individual 287 shearwaters to be tracked with GPS for an average of  $5.1 \pm 1.1$  days when making long-trips lasting 288 on average  $9 \pm 1.4$  days; representing 57% of their time at-sea (Table 2). Wedge-tailed shearwaters 289 were tracked with PTTs for an average of  $8.9 \pm 0.9$  days. On average, wedge-tailed shearwaters on 290 long-trips from Lord Howe Island travelled up to 495 ± 166 km from their colony whereas 291 conspecifics from Heron Island travelled up to  $672 \pm 62$  km on long-trips from their colony (Fig. 1). 292 293 The core-areas of wedge-tailed shearwaters on long-trips from Heron Island were located in the same general region each year (centroid: 19°48' S, 155°36' E), whereas core-areas of conspecifics 294 from Lord Howe Island were located in the same general region in 2014 and 2016 (centroid: 31°48' 295 S, 156°42' E), but not in 2015 (Table 2). 296

#### 298 PCA and collinearity

PCAs of oceanographic and tuna covariates within the climatology and hi-res datasets showed high 299 levels of correlation between covariates and consistent clustering of the same covariates in both 300 datasets. PCA of 10 covariates within the Heron Island climatology dataset revealed that the first 2 301 principal components account for 65.7% of the variance in the data (PC1 = 49.2%, PC2 = 16.5%). 302 and for PCA of the same covariates within the Lord Howe Island climatology dataset, the first 2 303 principal components account for 77.7% of the variance in the data (PC1 = 56.8%, PC2 = 20.9%; 304 Fig. 2). In both ordinations PC1 represents a positive relationship with latitude: higher SST and 305 tropical tuna biomass in the Coral Sea is associated with negative PC1 values where-as higher PRO 306 and BET\_ADU biomass in the Tasman Sea is associated with positive PC1 values. In the Heron 307 Island ordination PC2 represents an inverse relationship with proximity to a central seamount 308 region and in the Lord Howe Island ordination PC2 represents a positive relationship with 309 longitude. In both ordinations adult skipjack tuna (joined by adult vellowfin tuna in the Heron 310 Island PCA) was clustered with micronektonic skipjack and yellowfin tunas (joined by 311 micronektonic bigeve tuna in the Lord Howe Island PCA), hereafter termed the 'major-tuna cluster'. 312 In each PCA the major-tuna cluster was highly correlated with PC1 (Heron Island Pearson's  $R^2$  = 313 0.88 - 0.96, Lord Howe Island Pearson's  $R^2 = 0.85 - 0.95$ ) and SST (Fig. 2). 314

315

PCA of 13 covariates within the Heron Island 2015 hi-res dataset revealed that the first 2 principal components account for 56.3% of the variance in the data (PC1 = 33.9%, PC2 = 22.4%), PCA of the same covariates within the Lord Howe Island 2015 hi-res dataset revealed that the first 2 principal components account for 67.5% of the variance in the data (PC1 = 55.9%, PC2 = 11.6%) and PCA of the same covariates within the Lord Howe Island 2014 hi-res dataset revealed that the first 2 principal components account for 64.4% of the variance in the data (PC1 = 48.8%, PC2 =

15.6%; Fig. 3). The hi-res ordinations again show consistent clustering of tuna covariates, the Lord Howe Island 2014 and Lord Howe Island 2015 PCAs show all tuna covariates correlated with PC1 (2014 Pearson's  $R^2 = 0.68 - 0.98$ , 2015 Pearson's  $R^2 = 0.7 - 0.94$ ,), the Heron Island 2015 PCA shows lower correlation of all tuna with PC1 (Pearson's  $R^2 = 0.52 - 0.79$ ) as PC2 appears to split tuna covariates into two clusters (Fig. 3).

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#### 328 Climatology models

At broad scales, wedge-tailed shearwaters from both colonies selected core-areas, within their 329 foraging range, that were closer to seamounts (Heron Island:  $\chi^2 = 54.49$ , p = <0.001, Lord Howe 330 Island:  $\chi^2 = 53.71$ , p = <0.001; Table 3) and in deeper waters (Heron Island:  $\chi^2 = 17.08$ , p =331 <0.001, Lord Howe Island:  $\chi^2_1 = 24.29$ , p = <0.001). Primary productivity was negatively associated 332 with the selection of core-area location for wedge-tailed shearwaters breeding on Heron Island, the 333 model predicted that for every 100 mg C m<sup>-2</sup> day<sup>-1</sup> increase in primary productivity, an area is 12.16 334 times less likely to be selected as a core-area by Heron Island shearwaters ( $\chi^2 = 263.6$ , p = <0.001; 335 Fig. 4B). Primary productivity was not significant to selection of core-area location for wedge-336 tailed shearwaters breeding on Lord Howe Island ( $\chi^2 = 2.46$ , p = 0.117). Each climatology model 337 included two tuna covariates, one as a proxy for the major-tuna cluster (selected to have minimal 338 correlation with non-tuna covariates) and the other which was uncorrelated with the major-tuna 339 cluster. Micronektonic yellowfin (YFT\_MIC) was the proxy and micronektonic bigeye (BET\_MIC) 340 the uncorrelated covariate in the Heron Island model, while micronektonic bigeye was the proxy 341 and adult yellowfin (YFT\_ADU) the uncorrelated covariate in the Lord Howe Island model (Fig. 2). 342 The Heron Island climatology model predicted that for each additional 100 g m<sup>-2</sup> of tuna biomass an 343 area contains, it is 1.04 times more likely to be selected as a core-area by wedge-tailed shearwaters 344 if the tuna are micronektonic yellowfin ( $\chi^2 = 14.81$ , p = 0.001), and 2.06 times more likely to be 345

selected if the tuna are micronektonic bigeye ( $\chi^2$  = 235.24, p = <0.001). The Lord Howe Island 346 climatology model predicted that for each additional 100 g m<sup>-2</sup> of tuna biomass an area contains, it 347 is 1.42 times more likely to be selected as a core-area by wedge-tailed shearwaters if the tuna are 348 micronektonic bigeye ( $\chi^2_1$  = 70.35, *p* = 0.001; Fig. 4A), but 1.71 times less likely to be selected if 349 the tuna are adult yellowfin ( $\chi^2_1 = 15.39$ , p = <0.001). Validation of Heron Island climatology model 350 confirmed the final model fitted the data well (AUC = 0.97) and explained a good proportion of the 351 variance (McFadden's  $R^2 = 0.71$ ). Validation of Lord Howe Island climatology model confirmed the 352 final model also fitted the data well (AUC = 0.98) and explained a similar proportion of the 353 variance (McFadden's  $R^2 = 0.75$ ). 354

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#### 356 Hi-res models

The likelihood of wedge-tailed shearwater foraging was not significantly influenced by bathymetry, 357 and only shearwaters from Heron Island in 2015 were more likely to forage closer to seamounts 358 (Table 4; Fig. 5E). Shearwaters from Lord Howe Island in 2014 were more likely to forage at 359 higher Chlorophyll-a concentrations (Fig 5L) while sea surface height anomalies influenced 360 likelihood of foraging in shearwaters from Heron Island in 2015 only (Fig. 5D). Ekman upwelling 361 and sea surface temperature anomalies influenced likelihood of shearwater foraging in every 362 instance (Table 4; Fig. 5). The proxy selected to represent the major-tuna cluster was micronektonic 363 bigeye (BET\_MIC) in the Heron Island 2015 model, adult skipjack tuna (SKJ\_ADU) in the Lord 364 Howe Island 2015 model and micronektonic yellowfin tuna (YFT MIC) in the Lord Howe Island 365 2014 model (Fig. 3). Models predicted that for each additional 100 g m<sup>-2</sup> of tuna biomass an area 366 contains, shearwaters from Heron Island in 2015 were 1.3 times more likely to forage there (Table 367 4; Fig. 5C), shearwaters from Lord Howe Island in 2015 were 2.23 times more likely to forage there 368 (Fig. 5H) and shearwaters Lord Howe Island in 2014 were 1.13 times more likely to forage there 369

(Fig. 5K). Validation of the Heron Island 2015 model confirmed the final model fitted the data adequately (AUC = 0.79) and explained a good proportion of the variance (marginal  $R^2 = 0.31$ , conditional  $R^2 = 0.47$ ). Validation of the Lord Howe 2015 model confirmed the final model also fitted the data adequately (AUC = 0.73) and explained a similar proportion of the variance (marginal  $R^2 = 0.29$ , conditional  $R^2 = 0.51$ ). Validation of the Lord Howe 2014 model confirmed the final model fitted the data identically (AUC = 0.73) but explained a smaller proportion of the variance (marginal  $R^2 = 0.17$ , conditional  $R^2 = 0.30$ ).

377

### 378 **Discussion**

379

#### 380 Wedge-tailed Shearwater distributions and oceanography

Our results show that the long-trip destinations of Wedge-tailed shearwaters breeding on Heron 381 382 Island lie predominantly to the north-east of the colony in the Coral Sea, adding support for consistent use of this area over multiple years (McDuie et al. 2015). Our results also show, for the 383 first time, that during chick-rearing wedge-tailed shearwaters from Lord Howe Island undertake 384 long-trips to predominantly different regions in different years. In 2014 and 2016 wedge-tailed 385 shearwater long-trip destinations were distributed west of Lord Howe Island in the Tasman Sea over 386 the Tasmantid Seamounts that run parallel to the east Australian shelf, while in 2015 long-trip 387 destinations were predominately far to the north of Lord Howe Island in the Coral Sea, almost 388 overlapping with conspecifics foraging from Heron Island. Whether long-trips at Lord Howe Island 389 are undertaken as part of a coordinated dual-foraging strategy similar to that observed at Heron 390 Island (Congdon et al. 2005), or more opportunistically is currently unknown. However, these 391 results demonstrate that the unimodal foraging strategy observed by Peck & Congdon (2005) at 392 Lord Howe Island during early chick rearing appears to alter in the later stages of chick-rearing 393

- 394 (Jakubas et al. 2014), and/or between breeding seasons (Granadeiro et al. 1998).
- 395

Our results show that both Wedge-tailed Shearwater populations selected core-areas in deep pelagic 396 waters that were close to seamounts, these findings mirror those of other studies into the species' 397 bathymetric preferences (Catry et al. 2009, McDuie et al. 2015, McDuie & Congdon 2016). At fine 398 scales, wedge-tailed shearwaters from Heron Island in 2015 were more likely to forage close to 399 seamounts but bathymetry and seamounts did not influence conspecifics foraging from Lord Howe 400 Island. These results indicate that marine topography may be used by wedge-tailed shearwaters to 401 locate profitable foraging areas at broad but not fine scales. Fine scale foraging behaviour is likely 402 to be triggered by sea surface temperature anomalies and Ekeman upwelling which influenced 403 likelihood of foraging in all models. Collectively, these two dynamic covariates identify frontal 404 areas of water mixing and associated up and down-welling. Our results are consistent with other 405 studies that indicate that these dynamic phenomena are major mechanisms of prey aggregation for 406 seabirds in lower latitude waters (Spear et al. 2001, Hyrenbach et al. 2006, Weimerskirch et al. 407 2010, McDuie & Congdon 2016). 408

409

#### 410 **Tuna relationships**

We found tuna distributions to influence Wedge-tailed Shearwater core-area location and likelihood of foraging in every instance, supporting both our broad scale and fine scale hypotheses. This means that wedge-tailed shearwaters sought out areas of increased tuna biomass, which at the regional scale are relatively stable between years, and that they home in on aggregations of tuna for foraging at fine scale. Our results describe the importance of facilitated foraging opportunities in two seabird populations and are expressed in an ecologically meaningful way: using tuna density increases of 100g m<sup>-2</sup>, interpretable as the weight of one micronektonic tuna (165 mm long; Harley

et al. 2014), or a large meal for a wedge-tailed shearwater. For example we predict that in 2015, for every additional micronektonic tuna encountered per m<sup>2</sup> of ocean, it was 2.23 times more likely that wedge-tailed shearwaters from Lord Howe Island would forage there and 1.3 times more likely that conspecifics from Heron Island would forage there. Such predictions demonstrate the potential of our approach to quantify facilitated foraging opportunities. However, they also come with the caveat that they are only as accurate as the underlying modelled tuna distributions.

424

The major-tuna cluster identified in each PCA, and represented in each model by a single covariate, 425 showed that densities of most tunas were positively associated with: each other, wedge-tailed 426 shearwater selection of core-areas, and shearwater likelihood of foraging. The major-tuna cluster 427 represents significant spatial overlap in the distribution of adult and micronektonic tunas. In 428 locations where this cluster of tunas co-occurs we envisage a scenario where micronektonic tuna 429 and similar sized micronekton (30-250 mm) prey upon each other and attract adult tunas through 430 cannibalism and inter-species predation (Allain et al. 2007, Allain 2010). Predation within the 431 cluster takes place in epipelagic waters because micronektonic tuna have a non-developed swim 432 bladder that confines them to surface waters (Magnuson 1973). As such, where the major-tuna 433 cluster occurs, micronekton of appropriate size to be wedge-tailed shearwater prey (up to 145 mm; 434 Harrison et al. 1983) are preved upon by tuna in surface waters, presenting clear facilitated foraging 435 opportunities. 436

437

Only in the Heron Island climatology model did shearwaters associate with a micronektonic tuna
(bigeye) outside of the major-tuna cluster. Even small micronektonic tuna are proficient swimmers
(Graham et al. 2007) and it is unlikely that wedge-tailed shearwaters from these colonies possess
the diving ability (maximum recorded dive depth of 12m; Peck & Congdon 2006) to capture them

without subsurface predator assistance during the day. This suggests that subsurface predators not
considered in our models, such as cetaceans (Au & Pitman 1986), billfish (family: Istiophoridae) or
dolphinfish *Coryphaena hippurus* (Young et al. 2010b), also facilitate wedge-tailed shearwater
foraging in this region.

446

At broad scales, our results suggest wedge-tailed shearwaters are consistent facilitated foraging 447 commensals of adult skipjack tuna, intermittent facilitated foraging commensals of adult yellowfin 448 tuna, and that adult bigeye tuna do not facilitate wedge-tailed shearwater foraging. Many tropical 449 seabirds are known to associate foraging with skipjack tuna (Au & Pitman 1986, Jaquemet et al. 450 2004, Hebshi et al. 2008). Frequent seabird association with skipjack could be due to their greater 451 biomass, relative to other tunas in tropical seas, fostering greater facilitated foraging opportunities. 452 Additionally, the smaller size of skipjack, relative to other tunas, means that they also target suitable 453 sized prey (mean prey length 42 mm; Roger 1994) for wedge-tailed shearwaters (mean prey length 454 57 mm; Harrison et al. 1983). However, larger yellowfin and bigeye also feed on very small prey 455 relative to their own size (Ménard et al. 2006) and could conceivably target shearwater-sized prey. 456 Tuna occupy different vertical niches based on their size (which limits thermal tolerance) and 457 biology (i.e. development of the swim bladder in yellowfin and bigeye); larger tunas can spend 458 more time foraging for prev in deep water within and below the thermocline. During the day, adult 459 bigeye are typically found deepest, followed by yellowfin and then skipjack (Schaefer et al. 2009, 460 Schaefer & Fuller 2013, Scutt Phillips et al. 2015). Consequently, the predominantly surface-461 dwelling skipjack tuna (and micronektonic tuna age-classes) are more often encountered by diurnal 462 seabirds and thus more likely to serve as facilitated foraging hosts than larger adult yellowfin or 463 bigeye. 464

Tuna biomass in the Coral and Tasman Seas changes throughout the year following spawning and 465 seasonal changes in water temperature and habitat. Regional tuna spawning times are poorly 466 known, however skipjack and yellowfin are thought to spawn year-round depending on water 467 temperature (Schaefer 1996, Schaefer 2001) and bigeye are thought to spawn from October to 468 December (Evans et al. 2008). Regional tuna movement broadly follows thermal boundaries (Evans 469 et al. 2011), this is especially true for less thermally tolerant skipjack and micronektonic tunas, 470 which SEAPODYM models to follow the southward advance of warm water in the austral summer 471 and subsequent retreat north in the winter. This regional spawning and movement pattern causes 472 annual densities of tunas in the major-tuna cluster to peak in the southern Coral Sea in the austral 473 summer. In fact, adult bigeye spawning in December would produce ~250mm micronektonic 474 offspring in March (Nicol et al. 2011). This means that skipjack and micronektonic tuna biomass 475 peaks within the foraging range of wedge-tailed shearwaters at a time when they have the highest 476 energy demands of chick-rearing. If facilitated foraging with tuna is as important for wedge-tailed 477 shearwater populations as we suggest, then tuna seasonality could have an important role in shaping 478 wedge-tailed shearwater breeding phenology, as proposed for productivity in Indian Ocean 479 conspecifics (Catry et al. 2009). 480

481

#### 482 Effect of productivity gradient on long-trip destinations

We found that the tropical wedge-tailed shearwater population did not adhere to the 'temperate long-trip model' by seeking out areas of high primary productivity, but instead targeted regions of high tuna biomass. The subtropical population at times adhered to the 'temperate long-trip model' but also targeted tuna. The Heron Island results are consistent with wedge-tailed shearwater nonbreeding preferences, where birds exploit warm, oligotrophic waters in the Indian (Catry, et al. 2009) and Pacific oceans (McDuie & Congdon 2016) when freed from the need to central-place

forage. Our results also support the suggestion of a temporal and spatial decoupling between
satellite surface-measured primary productivity and micronekton/tuna aggregation in tropical
oceans (Lehodey et al. 1998, McDuie & Congdon 2016).

492

At broad scales the selection of core-areas by wedge-tailed shearwaters from sub-tropical Lord 493 Howe Island appeared uninfluenced by primary production. However, this was due to variation in 494 long-trip destinations between years. In 2014 wedge-tailed shearwater foraging was positively 495 associated with increased chlorophyll-a concentration and the population exploited sub-tropical 496 waters west of Lord Howe Island. In 2015 shearwater foraging was negatively related to 497 chlorophyll-a concentration and the population exploited tropical waters north of Lord Howe 498 Island. In 2014 the Lord Howe Island result conforms to our prediction for a sub-tropical shearwater 499 colony, where long-trip foraging destinations target enhanced productivity driven by oceanic fronts 500 (Baduini & Hyrenbach 2003, Paiva et al. 2010), but in 2015 the result does not. In 2015 wedge-501 tailed shearwater likelihood of foraging was more strongly associated with tropical tuna densities. 502 Individuals clearly transited over waters of the Tasman Sea, which are usually high in productivity, 503 to reach oligotrophic waters with high tuna biomass in the Coral Sea. A potential explanation is that 504 505 in years like 2015, productivity in the Lord Howe region becomes reduced, through a distancing or reduction in strength of the Tasman Front (Mulhearn 1987, Przeslawski et al. 2011). Wedge-tailed 506 shearwaters remained in sub-tropical waters to the west of Lord Howe Island in 2014 and 2016, 507 indicating that the 2015 northward movement could mark a departure from normal conditions, 508 however additional years of data are needed to confirm this. 509

510

511 Although it is unclear what triggers wedge-tailed shearwaters from Lord Howe Island to switch 512 long-trip destinations, it is unequivocal that both shearwater populations preferentially target tuna in

some years. At the same time of year in 2015, individuals from both populations undertook long, 513 purpose-directed flights towards almost the exact same region of the southern Coral Sea, indicating 514 an *a priori* expectation of high resource availability at these sites. This is not the indirect, looping 515 flight that tropical seabirds use to exploit unpredictable resources (Weimerskirch 2007, 516 Weimerskirch et al. 2010). Rather, it suggests that the tropical tuna biomass targeted by these flights 517 was not patchily distributed or ephemeral at broad spatio-temporal scales. The apparent reliability 518 of this resource suggests that in tropical systems, facilitated foraging with tuna can act as a 519 consistently available 'productive' long-trip destination, analogous to chlorophyll-a concentration in 520 temperate systems. In terms of population-level reliance on tuna we suggest that facilitated foraging 521 with tuna is consistently important to sustain breeding in the Heron Island wedge-tailed shearwater 522 population. Primary productivity per se appears more important to the Lord Howe Island wedge-523 tailed shearwater population in most years, although facilitated foraging with tuna becomes an 524 important strategy under certain conditions. As such, the relative importance of facilitated foraging 525 for wedge-tailed shearwater populations appears to be dependent upon their access to reliable areas 526 of high primary productivity. 527

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## 801 Tables

802 **Table 1.** Summary of oceanographic and tuna covariates used in broad and fine scale models

Covariate (units)	Abbreviation	Temporal resolution	Spatial resolution	Data source (provider)	
High resolution					
dynamic covariates					
Chlorophyll- <i>a</i> concentration (mg m <sup>-3</sup> )	CHL	8 day	4 km	MODIS & VIIRS (NASA)	
Sea surface temperature (°C)	SST	8 & 1 day	$0.1^{\circ} \& 0.25^{\circ}$	POES & AVHRR (NOAA)	
Sea surface temperature anomaly (°C)	SSTA	8 & 1 day	$0.1^{\circ} \& 0.25^{\circ}$	POES AVHRR (NOAA)	
Sea surface height anomaly (m)	SSHA	1 day	0.083°	HYCOM & NCODA (NRL)	
Ekman upwelling (m day <sup>-1</sup> )	EKM	1 day	0.25°	Metop ASCAT	
Climatology and static					
Primary productivity (mg C $m^{-2}$ day <sup>-1</sup> )	PRO	Monthly (20 yr mean)	4.4 km	SeaWiFS & AVHRR (NASA & NOAA)	
Sea surface temperature (°C)	SST	Monthly (30 yr mean)	4.4 km	AVHRR (NOAA)	
Bathymetry (m)	BTY	Static	0.083°	GEBCO	
Distance to seamount (km)	SMT	Static	0.083°	Global seamount database	
Tuna distribution covariates					
Tuna weekly biomass	BET_ADU, BET_MIC,	7 day	0.25°	INDESO V2 Fished	
distribution (g m <sup>-2</sup> )	YFT_ADU, YFT_MIC,			(SEAPODYM) (unavailable in 2016)	
Tuna monthly biomass	RFT ADU RFT MIC	Monthly	10	INTERIM Fished	
distribution ( $\sigma$ m <sup>-2</sup> )	YFT ADU YFT MIC	(30 vr mean)	1	(SEAPODYM)	
distribution (5 m )	SKJ_ADU, SKJ_MIC	(50 yr mean)			

Table 2. Summary of Wedge-tailed Shearwater GPS and PTT\* tracked long-trips collected during
the study. The core-area is the 50% utilization distribution from kernel analysis of each tracking

<sup>806</sup> dataset.

Island	Year	n trips	Max colony	Trip	Tracked	Trip	Core-area
Colony			distance	length	days	length	centroid
			( <b>km</b> )	( <b>km</b> )		( <b>d</b> )	(Lat, Long)
Lord Howe	2016	19	$328 \pm 114$	$1165 \pm 381$	$4.4 \pm 1.3$	$6.9 \pm 2.3$	-31.8, 157.2
Lord Howe	2015	14	$661 \pm 291$	$1986 \pm 590$	$5.9 \pm 2.3$	$9.9 \pm 3.3$	-27.7,158.1
Lord Howe	2014	8	$498 \pm 292$	$1588 \pm 561$	$6 \pm 1.7$	$9.6 \pm 4$	-31.8, 156.2
Heron	2015	8	$625 \pm 223$	$1532 \pm 510$	$3.9 \pm 1.5$	$9.5 \pm 2$	-20.4, 156.9
Heron *	2013	9	$744 \pm 312$		$8.3 \pm 3.2$		-19.8, 154.2
Heron *	2011	3	$649 \pm 375$		$9.7 \pm 2.3$		-19.2, 155.7

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**Table 3.** Climatology logistical regression models of wedge-tailed shearwater presence-absence against broad scale oceanographic and tuna covariates. The following coefficients ( $\beta$ ) and std. errors (SE) are expressed in terms of a 100 unit change: tuna covariates (100 g m<sup>-2</sup>), seamount distance (100 km) and bathymetry (100 m). All covariates have significance p < 0.001. Core-area and foraging range covariate values are expressed as mean  $\pm$  standard deviation

Colony	Covariate	$\beta \pm SE$	Core-area	Foraging range
Heron Island				
	Intercept	$1.195 \pm 1.304$		
	Productivity (mg C m <sup>-2</sup> day <sup>-1</sup> )	$-0.025 \pm 0.002$	$425.53 \pm 42.62$	$534.80 \pm 154.28$
	Seamount distance (km)	$-0.698 \pm 0.099$	96.38 ± 83.85	$167.52 \pm 125.95$
	Bathymetry (m)	$0.034 \pm 0.008$	$2756 \pm 940$	$2709 \pm 1389$
	Micronektonic bigeye tuna biomass (g m <sup>-2</sup> )	$0.721 \pm 0.056$	$0.084 \pm 0.017$	$0.066 \pm 0.021$
	Micronektonic yellowfin tuna biomass (g m <sup>-2</sup> )	$0.040 \pm 0.010$	$0.334 \pm 0.060$	$0.268 \pm 0.150$
	Autocovariate	$0.129 \pm 0.005$		
Lord Howe Island				
	Intercept	$-4.681 \pm 0.706$		
	Seamount distance	$-1.169 \pm 0.176$	$84.56 \pm 45.07$	$168.10 \pm 144.98$

(km)			
Bathymetry (m)	$0.048 \pm 0.010$	3429 ± 1069	2772 ± 1359
Micronektonic bigeye tuna biomass (g m <sup>-2</sup> )	$0.350 \pm 0.048$	$0.067 \pm 0.010$	$0.045 \pm 0.030$
Adult yellowfin tuna biomass (g $m^{-2}$ )	$-0.534 \pm 0.137$	$0.027 \pm 0.008$	$0.029 \pm 0.007$
Autocovariate	$0.112 \pm 0.004$		

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**Table 4.** Hi-res logistical regression models of wedge-tailed shearwater probability of foraging against fine scale oceanographic and tuna covariates. The strength ( $\chi^2$ ) and effect direction of each covariate is given for each of the three colony-year models. Covariates with significance p < 0.01are shown in bold and the corresponding effect given, NA indicates that the covariate was not included in a model due to multicollinearity. Positive and negative effect directions are denoted by  $\uparrow$  and  $\downarrow$  respectively, and in the case of a polynomial relationship the value at which foraging is most or least (denoted by \*) likely is given. For coefficient ( $\beta$ ) and std. error values see Appendix 2

Covariate	Heron Island 2015		Lord Howe Island 2015		Lord Howe Island 2014	
	Effect	Strength $(\chi^2)$	Effect	Strength $(\chi^2)$	Effect	Strength ( $\chi^2$ )
Bathymetry (m)		0.40		3.80		NA
Seamount distance (km)	$\downarrow$	38.77		0.80		0.04
chlorophyll- $a$ concentration (mg m <sup>-3</sup> )		NA		NA	Ť	9.67
Ekman upwelling (m day <sup>-1</sup> )	0.22	16.83	-0.43*	65.61	0.67*	48.91
Sea surface temperature anomaly (°C)	1	29.46	-0.17	101.77	$\downarrow$	6.57
Sea surface height anomaly (m)	0.62	33.27		NA		3.27
Tuna biomass (g m <sup>-2</sup> )	↑	26.07	↑	230.21	1	66.14

## 825 Figures



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Fig. 1. Wedge-tailed Shearwater long-trip tracking data collected from birds rearing chicks on
Heron Island and Lord Howe Island between 2011 and 2016, overlaid with 50% kernel utilisation
distribution (UD) core-use areas for each colony





Fig. 2. Ordination from principal components analysis of broad scale oceanographic and tuna
covariates used in climatology models of Wedge-tailed Shearwaters breeding from (A) Heron
Island, and (B) Lord Howe Island

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**Fig. 3.** Ordination from principal components analysis of fine scale oceanographic and tuna covariates used in hi-res models of Wedge-tailed Shearwaters tracked with GPS from (A) Heron Island in 2015, (B) Lord Howe Island in 2015, and (C) Lord Howe Island in 2014



Fig. 4. GPS tracking data of breeding Wedge-tailed Shearwaters shown as (A) kernel utilisation
distributions (UDs) overlaying SEAPODYM-predicted long-term mean micronektonic bigeye tuna
biomass for march, (B) kernel UDs overlaying long-term mean primary productivity for march, and
(C) individual foraging trips from Heron Island overlaying SEAPODYM-predicted weekly
micronektonic skipjack tuna biomass.



Fig. 5. Mean (solid line) and 95% confidence interval (dotted line) predictions of Wedge-tailed Shearwater foraging probability in relation to fine scale tuna and oceanographic covariates. Density plots are overlaid showing the distributions of foraging (black line) and non-foraging (grey shade) samples for each covariate. 'Mic.' refers to micronektonic tuna age-classes

## 852 Appendix

Appendix 1. Flow chart showing input data and analyses steps taken for: A. hi-res models; and B.

climatology models (separated by dotted line).



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Appendix 2. Logistical regression coefficients for hi-res models of wedge-tailed shearwater probability of foraging against fine scale oceanographic and tuna covariates. The following coefficients ( $\beta$ ) and std. errors (SE) are expressed in terms of a 100 unit change: tuna covariates (100 g m<sup>-2</sup>), seamount distance (SMT) (100 km)

Colony – Year	Covariate	$\beta \pm SE$
Heron Island 2015		
	Intercept	$-11.829 \pm 3.764$
	poly(EKM)1	$0.900 \pm 0.392$
	poly(EKM)2	$-2.000 \pm 0.515$
	SSTA	$1.499 \pm 0.275$
	poly(SSHA)1	$3.773 \pm 1.086$
	poly(SSHA)2	$-3.019 \pm 0.775$
	SMT	$-0.907 \pm 0.157$

	BET_MIC	$0.265 \pm 0.052$				
Lord Howe Island 2015						
	Intercept	$-2.583 \pm 0.416$				
	poly(EKM)1	$1.064 \pm 0.163$				
	poly(EKM)2	$1.592 \pm 0.310$				
	poly(SSTA)1	$-0.201 \pm 0.230$				
	poly(SSTA)2	$-0.604 \pm 0.143$				
	SKJ_ADU	$0.804 \pm 0.055$				
Lord Howe Island 2014						
	Intercept	$-2.182 \pm 0.859$				
	poly(EKM)1	$-1.240 \pm 0.350$				
	poly(EKM)2	$1.454 \pm 0.621$				
	SSTA	$-0.430 \pm 0.167$				
	SSHA	$0.860 \pm 0.477$				
	YFT_MIC	$0.119 \pm 0.016$				
	Chlorophyll-a	$0.232 \pm 0.080$				