1 2	Wintering in the sun: niche partitioning by three nonbreeding <i>Pterodroma</i> petrel species in the equatorial Pacific Ocean
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4	Running head: Niche partitioning by nonbreeding Pterodroma petrels
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41 42	Key words: Species distribution models, stable isotope analysis, niche, foraging ecology, seabirds, tropical Pacific

43 Abstract

44 Niche divergence is expected for species that compete for shared resources, including migrants that 45 occupy similar regions during the non-breeding season. Studies of temperate seabirds indicate that both spatial and behavioural segregation can be important mechanisms for reducing competition, but 46 47 there have been few investigations of resource partitioning by closely-related taxa in low productivity, 48 tropical environments. We investigated niche partitioning in three gadfly petrel taxa, Pterodroma 49 *leucoptera leucoptera* (n = 22), *P. leucoptera caledonica* (n = 7) and *P. pycrofti* (n = 12) during their 50 non-breeding season in the eastern tropical Pacific Ocean by combining tracking data from geolocator-immersion loggers with remotely-sensed environmental data in species distribution models 51 52 (SDMs), and by comparing feather stable isotope ratios. The three taxa showed spatial partitioning; 53 two foraged in the North Equatorial Counter Current and one in the South Equatorial Current. This 54 reflected differences in their realised habitat niches, with significant taxon-specific responses to thermocline depth, sea surface temperature and bathymetry. There were also differences among taxa 55 in activity patterns, and all birds spent a much larger proportion of time in flight at night than during 56 57 the day, suggesting predominance of nocturnal foraging behaviour. Comparison of stable isotope ratios in feathers suggests that P. leucoptera leucoptera and P. pycrofti mainly consume vertically-58 migrating mesopelagic fishes, whereas the diet of *P. leucoptera caledonica* also include some lower 59 60 trophic level including crustaceans and squid. Unique insights can be gained from studies of the 61 foraging ecology of tropical pelagic seabirds, in comparison with temperate and polar waters, and are 62 urgently required for understanding and protecting tropical avifauna in key marine habitats

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70 Introduction

71 The distribution of top predators in the marine environment is frequently linked to physical 72 oceanographic processes that govern spatial and temporal variation in primary productivity and associated prey availability (Block et al. 2011). In particular, physical forcing in marine boundary 73 74 currents creates convergence and upwelling zones known as some of the most productive marine 75 environments on earth, where predators target and compete for prey concentrations that can be 76 predictable at meso to large spatial scales (Belkin et al. 2009, Bost et al. 2009). Studies of seabird 77 distribution during the breeding season demonstrate that populations minimise interspecific competition in such foraging zones through a variety of mechanisms including segregation in space 78 79 and time, behaviour and diet (Phalan et al. 2007, Navarro et al. 2013, Moreno et al. in press). During 80 the nonbreeding season, when seabirds are free from central-place foraging constraints, studies highlight the importance of spatial and dietary segregation, frequently across ecological gradients 81 82 related to ocean temperature, reflecting the location of boundary systems (Spear et al. 2007, Thiebot et al. 2012, Cherel et al. 2013, Navarro et al. 2015, Quillfeldt et al. 2015). 83

84 Small gadfly *Pterodroma* petrels within the subgenus *cookilaria* (~ 10 species) are a group of seabirds 85 that usually make long distance migrations during the non-breeding season to productive boundary systems in the temperate northern and southern Pacific Ocean where they exhibit both spatial and 86 87 temporal resource partitioning (Rayner et al. 2011, Rayner et al. 2012, Priddel et al. 2014). However, not all *cookilaria* migrate to temperate habitats, with some species occupying tropical non-breeding 88 89 habitats such as the eastern tropical Pacific Ocean (ETPO) (Kessler 2006), where they are part of a diverse seabird community (Au & Pitman 1986, Ballance et al. 1997, Spear et al. 2007, Priddel et al. 90 2014). Both theoretical (Pianka 1974) and empirical data (Torres 2009, Young et al. 2010) suggest 91 that in such low productivity tropical habitats, competitors will increase niche separation to avoid 92 competition. Accordingly, the community of cookilaria in the ETPO represents a distinctive 93 94 ecological system for investigating niche partitioning in small Procellariiformes, providing a useful comparison with studies in temperate systems (Ainley et al. 1992, Navarro et al. 2015) 95

96 A number of gadfly petrels endemic to breeding sites in temperate Australasia are known, or 97 considered likely, to occupy the ETPO during part of their seasonal cycle. Gould's Petrel (Pterodroma leucoptera; ~200–250 g) occurs as two subspecies with low levels of gene flow between 98 99 populations breeding on islands off the coast of New South Wales, Australia (P. leucoptera 100 leucoptera, hereafter leucoptera) and on the main island of New Caledonia (P. leucoptera caledonica, 101 hereafter caledonica)(de Naurois 1978, Priddel et al. 1995, Gangloff 2010). During the nonbreeding season, both subspecies migrate to the ETPO (Priddel et al. 2014). Pycroft's petrel (Pterodroma 102 *pycrofti*, hereafter *pycrofti*) is endemic to islands off the northeast coast of New Zealand (~130 –200 g 103 (Marchant & Higgins 1990) but has a poorly known nonbreeding distribution that likely extends, in 104 part, to the tropical Pacific (Spear et al. 1992). Recent genetic research indicates that pvcrofti is a 105 sister taxon to *leucoptera* and *caledonica* (Steeves et al in prep) presenting an opportunity for 106 107 comparative analysis of foraging ecology in these morphologically and genetically similar taxa. 108 We investigated niche separation in *leucoptera*, *caledonica* and *pycrofti* during the nonbreeding 109 season by combining tracking data from geolocator-immersion loggers with remotely-sensed 110 environmental data in species distribution models (SDM), and by comparing stable isotope ratios in feathers grown by tracked individuals during the same period. Our aims were to present the first 111 detailed analysis of the nonbreeding movements of individual pycrofti and to highlight differences in 112 113 habitat use and foraging ecology indicative of niche partitioning among these three closely-related 114 taxa.

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117 Materials and Methods

118 Tracking data

119 Tracking methods for *leucoptera* and *caledonica* are provided by Priddel *et al.* (2014). In summary, combined geolocator-immersion loggers (MK14, British Antarctic Survey, Cambridge, UK and 120 121 NanoLAT2900, Lotek, Ontario, Canada) were deployed on adult leucoptera (subspecies population size 800 – 1000 breeding pairs (Priddel & Carlile 2007)) at Cabbage Tree Island (Australia) on March 122 123 2010 and *caledonica* (subspecies population size 1000 – 10000 breeding pairs (Brooke 2004)) at 124 Grande Terre (the main island of New Caledonia) in January 2010. Loggers were retrieved at both sites between November 2010 and January 2011 providing data on the entire nonbreeding period for 125 126 22 leucoptera and 7 caledonica respectively. Similar loggers (MK18, British Antarctic Survey, Cambridge, UK, 2g) were attached to twelve pycrofti (subspecies population size 2500 – 4000 127 128 breeding pairs (Brooke 2004)) at Red Mercury Island (New Zealand) in December 2009, and 10 129 (83%) were retrieved in January 2011, of which two failed to provide data. All loggers were attached 130 using the methods described by Rayner et al. (2008) and weighed < 1.5% of adult body mass. 131 Light data from the loggers were processed following the methods of Rayner et al. (2012). The 132 nonbreeding phase was defined according to Priddel et al. (2014) based on the first location outside or 133 134 inside a 1000 km buffer (an arbitrary but standardised measure) around the colony at the end or start, respectively, of the breeding season. The loggers tested for saltwater immersion data every 3 s, with 135 the data binned into ten minute intervals, resulting in values ranging from 0 (dry) to 200 (immersed 136 for whole period). These data were analysed using the methods of Rayner et al. (2012), providing 137 temporal data on percent of time immersed, flight bouts greater than 10 minutes, and duration of flight 138 139 bouts during daylight and darkness (based on the timing of civil twilight) for each species.

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141 Construction of species distribution models (SDM)

142 Post-processed geolocation data from each species during the nonbreeding season were included in 143 binomial species distribution models (SDM) following methods presented in Torres et al. (2015). In brief, locations from tracked individuals of each species (*leucoptera* n = 22, *caledonica* n = 7, *pycrofti* 144 n = 8) were used to construct monthly kernel density estimates (search radius 200 km) during the non-145 146 breeding season for all three species between April and October. Presence data for each SDM were those locations that fell within the 50% density contour for each of those months, which was 147 148 considered to represent core habitat. Background data (otherwise known as pseudo-absences) implemented in the SDM for each month were uniformly-spaced locations (every 100 km²) falling 149 within the 90% density contour for all species locations (April through October; n = 6245). This 150 method was based on the assumption that all habitat was equally available to the three taxa within the 151 152 timeframe of their non-breeding phase. Background locations were randomly assigned dates between 153 the date of arrival and departure for each species, and bird identity in proportion to the number of presence points from each bird. Using these background and presence data we produced binomial 154 155 boosted regression tree (BRT)(Freidman 2001) models of presence-availability (Boyce et al. 2002, 156 Torres et al. 2015) that describe the distribution of each *cookilaria* petrel relative to the available 157 habitat across the entire tropical pacific region exploited by all species.

For each species model of presence-availability, a range of static and remotely-sensed environmental 158 159 data were used to describe habitat use. Depth values at each presence and background point were 160 extracted from the 30 arc-second General Bathymetric Chart of the World (http://www.gebco.net/; GEBCO). Seabed slope angle and minimum distance to land were derived from the GEBCO 161 bathymetry data using the package 'raster' in R 3.1.1 (R Core Team 2014). Several dynamic 162 oceanographic variables (Table 1) were obtained from NOAA ERDDAP web servers using the 163 164 Xtractomatic routines in R (http://coastwatch.pfel.noaa.gov/xtracto/): Chlorophyll-a concentration (CHL; 8-day), Sea surface temperature (SST; 8-day), sea surface height deviation (SSH; 1-day), 165 Ekman upwelling (EKM; 3-day) and wind speed (WIND; 3-day). We obtained gridded, annual 166 167 climatology data (12 month) for top of thermocline depth from IFREMER (http://www.ifremer.fr/cerweb/deboyer/mld); selecting the appropriate monthly grids for the non-168

breeding period. The median values obtained for these environmental variables were those within a 200km radius of each location. This extraction technique accounted for the potential 200+ km spatial error in geolocation data (Phillips et al. 2004). The absolute deviation of the dynamic variables were also included in the models to assess how environmental variability may influence petrel habitat and log transformations of slope, WIND, CHL and EKM were implemented in the models to account for skewed distributions.

175 SDMs of each petrel species were generated using BRT models, which has demonstrated strong 176 predictive performance and model parsimony, including for another pelagic petrel species (Elith et al. 2006, Buston & Elith 2011, Torres et al. 2013, Torres et al. 2015). BRT is a machine learning method 177 that can handle correlated, interacting and non-linear data, all of which are common in ecological 178 179 studies (Leathwick et al. 2006, Elith et al. 2008). Two algorithms are applied in BRT modelling: the first partitions the predictor space into homogeneous response groups using decision trees, and the 180 second boosts this process to iteratively optimize the predictive performance of the model (Elith et al. 181 2008). The process combines a large number of individual decision trees to generate a BRT model. 182

183 The contribution of each predictor variable to a BRT model is determined by the number of times it is used to split a tree branch. If a predictor variable contributed less than 5% to the model, the model 184 was re-run without that variable (Buston & Elith 2011). The learning rate (lr) of a BRT determines the 185 186 contribution of each fitted tree to the final model and was set at 0.0025, while the bag fraction, which is the proportion of samples used at each tree from the whole dataset, was set at 0.5. Tree complexity 187 (tc), which represents the number of nodes on each tree and determines the number of interactions 188 between predictors, was allowed to vary between 1 and 4 so that the number of boosting iterations, 189 known as the number of trees (nt), was greater than 1000, as recommended by Elith et al. (2008). 190 191 Each model dataset included multiple presence and background points from the same bird track, and with varying sample sizes; to account for this internal structure in the datasets, CV-folds were 192 193 specified (De'ath 2007) as all presence and background locations from an individual bird. CV-folds 194 withhold subsets of data from the model at each tree that are subsequently used to test model fit.

195 While generating each BRT, 10% of presence and 10% of background locations were withheld from 196 model calibration for external validation to assess predictive performance and select optimal model parameters. Four metrics were used to select the optimal model and evaluate predictive performance: 197 cross-validation deviance explained and area under the receiver operator curve (AUC) calculated 198 199 during the modelling procedure, and by external validation metrics of deviance between observed and predicted values (validation deviance) and AUC (validation AUC) calculated using the withheld data. 200 201 Due to the presence vs. availability design of the BRT models, k-fold cross validations (Boyce et al. 202 2002) were also conducted on the optimal models to assess the predictive capacity of 'used' locations, 203 while ignoring the predictability of absence locations because these are less certain when working 204 with background or pseudo-absence data (Torres et al. 2015). The k-fold cross validation binned the 205 predicted habitat suitability of each presence and absence location into equal-interval groups between 206 0 and 1 (0-0.1, 0.1-0.2, 0.2-0.3, etc.), and the proportion of presence locations in each bin was determined. A Spearman-rank correlation (rs) was calculated between bin rank (0, 0.1, 0.2, 0.3, etc.) 207 208 and the proportion of presence locations to assess whether the latter increased with increasing 209 suitability of predicted habitat, indicating good predictive performance (Torres et al. 2015).

210 Stable isotope ratios

Observations of our study taxa indicate worn and fresh plumage at the end and beginning of each 211 212 breeding season, consistent with previous research indicating that *cookilaria* moult occurs during nonbreeding when dietary signals are encorporated into new plumage (Marchant & Higgins 1990, 213 Spear et al. 1992) (Hobson 1999). Stable isotope ratios of N (δ^{15} N) and C (δ^{13} C) provide an indication 214 of both the trophic level and carbon source (benthic versus pelagic, inshore versus offshore, and 215 information on water mass) of prey ingested during the time of tissue formation, which in the case of 216 feathers from adult seabirds typically allows comparisons between trophic level and geographic 217 218 segregation during the nonbreeding period (Hobson 1999, Phillips et al. 2009). To enable a 219 comparison between stable isotope ratios and distribution from tracking data, a single body feather 220 was collected from each *leucoptera* (n = 10), *caledonica* (n = 8) and *pycrofti* (n = 10) upon geolocator 221 retrieval. Feathers were stored in plastic bags in the field. Once in the lab, feathers were cleaned with

222	70% ethanol, rinsed in distilled water to remove contaminants, dried in at 50°C and cut into very fine
223	fragments. Stable isotope analyses of a subsample (~0.7 mg) of each homogenized feather were
224	carried out at National Institute of Water & Atmospheric Research (NIWA) using an AS200_LS
225	autosampler and NA 1500N (Fisons Instruments, Rodano, Italy) elemental analyser combustion
226	furnace connected to a Delta ^{Plus} continuous flow, IRMS (Thermo-Fischer Scientific, Bremen,
227	Germany). Operational details are outlined in (Rayner et al. 2008) with the exception that δ^{13} C values
228	were calibrated against CO ₂ reference gas, relative to the international standard Carrara Marble NSB-
229	19 (National Institute of Standards and Technology (NIST), Gaithersberg, MD, USA). This, in turn,
230	was calibrated against the original Pee Dee Belemnite (PDB) limestone standard and was then
231	corrected for ¹⁷ O. A two-point normalisation process using NIST 8573 (USGS40 L-glutamic acid;
232	certified $\delta^{15}N = -4.52 \pm 0.12$ ‰) and IAEA-N-2 (ammonium sulphate: certified $\delta^{15}N = +20.41 \pm 0.2$ ‰)
233	was applied to δ^{15} N data. Carbon isotope data were corrected via a two-point normalisation process
234	using NIST 8573 (USGS40 L-glutamic acid; certified $\delta^{13}C = -26.39 \pm 0.09$ ‰) and NIST 8542
235	(IAEA-CH-6 Sucrose; certified δ^{13} C = -10.45 ±0.07 ‰). DL-Leucine (DL-2-Amino-4-
236	methylpentanoic acid, C ₆ H ₁₃ NO ₂ , Lot 127H1084, Sigma, Australia) was run every 10 samples to
237	check analytical precision and enable drift corrections to be made if necessary. Additional
238	international standards NIST 8574 (USGS41 L-glutamic acid; certified $\delta^{13}C = +37.63 \pm 0.10$ ‰ and
239	δ^{15} N = +47.57 ±0.22 ‰), NIST 8547 (IAEA-N1 ammonium sulphate; certified δ^{15} N = +0.43 ±0.04)
240	were run daily to check isotopic accuracy. Repeat analysis of standards produced data accurate to
241	within 0.25 ‰ for both δ^{15} N and δ^{13} C, and a precision of better than 0.32 ‰, for δ^{15} N and 0.24 ‰
242	δ^{13} C.

243 Statistical analyses

Migration arrival and departure dates were compared between species using contingency analysis. Following tests for normality, a combination of non-parametric tests (Kruskall Wallace tests) and parametric tests (ANOVA) were used to test for differences among species in time spent within the non-breeding core range (50% kernel), activity parameters (based on the immersion data), and stable isotope ratios, respectively. Parametric and non-parametric multiple comparisons were used to test

- for pairwise differences (Wilcoxon matched pairs and Tukey's pairwise comparisons) between
- species. Analyses were conducted using JMP 11.2.0 (©SAS Institute 2013) with a threshold of
- significance at $\alpha = 0.05$. Unless indicated otherwise, data are presented as mean \pm SD. Geospatial
- processing of geolocation data was conducted using ArcGIS v10.3 (ESRI, CA, USA). Extraction of
- 253 remote-sensing data, creation of static environmental variables and BRT modelling were conducted in
- R 3.1.1 (R Development Core Team 2013) using the packages dismo (Hijmans et al. 2012), Raster
- 255 (Hijmans & van Etten 2012), Gbm, PresenceAbsence (Freeman 2007), and with custom code by Elith
- et al. (2008).

257 **Results**

258 Processing of light data from *leucoptera* (n = 22 adults), *caledonica* (n = 7 adults) and *pycrofti* (n = 8 259 adults), provided a total of 5287, 2330 and 2965 locations, respectively, for SDM analyses. Leucoptera, caledonica and pycrofti exhibited spatial segregation in their core distributions within the 260 central and eastern tropical Pacific Ocean, but there were no significant differences in migration 261 timetables, including the time spent in core areas (Table 2). The core region used by *leucoptera* was 262 263 from 10°N-5°S, 150-165°W, encompassing the eastern sector of the Republic of Kiribati (Line Islands), whereas that used by caledonica was 3000--6000 km southeast in the region of the East 264 Pacific Rise (0 - 15°S, 135°- 100°W) (Figs. 1 and 2). Core areas of pycrofti were from 0-10°N, 140°-265 135°W in the eastern equatorial Pacific, and situated between those of *leucoptera* and *caledonica* 266 (Fig. 1 and 2). The migration routes of pycrofti were similar to those of leucoptera and caledonica 267 268 (Priddel et al 2014); all birds first migrated east from New Zealand (at approximately 40°S) and then north to reach their nonbreeding range, and the return (pre-breeding) migration was southwest through 269 270 Melanesia to reach waters around their respective colony (Fig. 1).

271 Species distribution models

272 Optimal BRT models for each petrel species performed well according to internal validation metrics 273 and external measures of predictive performance using the withheld data (Table 3). Four predictor 274 variables were common to all three models: Thermocline, Depth, SST and Chl collectively 275 contributed 79%, 74%, 68%, and 39%, respectively, to all three models (total contribution of 87%: 276 261/300% for all three models). All models had a tree complexity of two, allowing one interaction 277 between terms. Species-environment response plots for these four variables indicate that each species used different habitats (Fig. 3). Intra-species comparisons showed that presence of leucoptera and 278 *caledonica* peaked in habitats where the thermocline was reached at > 100 m, and both species 279 exhibited a positive response to a deeper thermocline. Conversely, pycrofti showed a negative 280 response to a deeper thermocline, with presence peaking in habitats with the thermocline at ~ 25 m 281 (Fig. 3). Niche separation by depth was also evident, with *leucoptera* presence peaking in the deepest 282

regions (> 5000 m), *pycrofti* in habitats with water depths of 4-5,000 m, and *caledonica* in shallower
habitats (< 4000 m) (Fig. 3). *Leucoptera* presence peaked in habitats with the highest SST (27-29 °C), *pycrofti* at mid-temperatures (25-28 °C), and *caledonica* in cooler waters (20 - 26 °C) (Fig. 3). Overlap
in habitat preferences relative to Chl was apparent between *caledonica* and *pycrofti*, with both taxa
showing increased presence in waters with low Chl, whereas *leucoptera* avoided that habitat (Fig 3).

288 At-sea activity patterns

289 There were significant differences in flight activity between daylight and darkness in all three species; tracked birds spent less time on the water and made more, and longer flight bouts in darkness (Table 290 291 5; pairwise tests at 0.01). There was no significant difference in the daylight activity patterns between species; all three taxa spent a similar amount of time on the water ($F_{2,20} = 2.16$, P = 0.14), and the 292 293 number of flight bouts (P = 0.34), and flight bout duration (P = 0.42) were comparable (Fig. 4). In 294 contrast, during darkness, the time spent on the water by pycrofti was greater than in leucoptera and *caledonica* (Pairwise Tukey tests P < 0.05 and P < 0.001), and by *leucoptera* was greater than in 295 296 *caledonica* (Tukey HSD P < 0.05) (Table 4, Fig. 4). The trend of higher night time activity by 297 caledonica was consistent with the significantly higher number of flight bouts and longer duration of 298 flight bouts during darkness than in *leucoptera* and *pycrofti* ($F_{2,20} = 8.61$, P < 0.01; $F_{2,20} = 13.31$, P < (0.001); these last two taxa did not differ significantly in the number (P = 0.18) or duration of flight 299 300 bouts (P = 0.06).

301 *Leucoptera* and *pycrofti* exhibited higher flight activity during daylight and darkness at the beginning
302 and ending of the nonbreeding period, with a reduction in activity in June to August (Fig. 4).

303 *Caledonica* showed similar activity patterns across the nonbreeding season during daylight hours,

304 whereas night time activity remained high from April to October (Fig. 4).

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- 308 Isotopic niche
- 309 Nitrogen isotope ratios differed significantly between the three species (Fig. 5; $F_{2,27} = 19.52$, P <
- 310 0.001). δ^{15} N in feathers of pycrofti (16.91 ± 1.67) and leucoptera (15.22 ± 1.67) did not differ
- significantly (δ^{15} N 16.91 ± 1.67 and δ^{15} N 15.22 ± 1.67 respectively; pairwise tukey test, p = 0.09), but
- 312 were higher than that in *caledonica* (12.45 ± 0.93, Pairwise Tukey tests both p < 0.01) (Fig. 5). δ^{13} C
- 313 in feathers of *caledonica* (-15.72 \pm 0.55) were higher than in *leucoptera* (-16.43 \pm 0.60) but not
- 314 *pycrofti* (-16.04 \pm 0.28) (Fig. 5, $F_{2,27}$ = 4.93, P < 0.01, Pairwise Tukey tests: *caledonica leucoptera*
- 315 (p < 0.01), caledonica pycrofti (P = 0.38), leucoptera pycrofti (p = 0.18).

317 Discussion:

318 *Pterodroma* petrels are among the most wide-ranging of all birds, and capable of traversing >1,000 319 km within a single day (Pinet et al. 2011, Rayner et al. 2011, Rayner et al. 2012). Despite this 320 capacity for long-distance travel and thus shared habitat use, our three study taxa showed clear 321 differences in distribution and habitat use during the non-breeding season but not in the timing of 322 movements to and from these habitats. The core distributions of *leucoptera* and *pycrofti* were 323 separated longitudinally - west of ~ 158° W, and east of 133° W, respectively - within the North Equatorial Counter Current (NECC) (Fig. 2, C), extending south to the edge of the cooler South 324 Equatorial Current (SEC)(2-5°N) and north to the North Equatorial Current (NEC)(12-14°N), whereas 325 326 caledonica was distributed mainly in the South Equatorial Current (SEC) (0 - 15°S) between 135°W 327 and 95°W. The preference of *pycrofti* and *leucoptera* for the waters of the NECC is consistent with 328 at-sea observations of a range of procellariiform species in this region, including *Puffinus newelli*, 329 Puffinus pacificus, Pseudobulweria rostrate and other gadfly petrels, Pterodroma cervicalis, Pterodroma externa and Pterodroma sandwichensis, (Ballance et al. 1997, Spear et al. 2001). 330 331 Previous studies have suggested that *leucoptera* associates with the SEC (Ribic et al. 1997). However, this previous supposition is inconsistent with our data, which indicated that the tracked 332 *leucoptera* occupied strikingly different habitats to its sister taxon in the western NECC. This result 333 334 reaffirms the utility of tracking studies for revealing population-specific foraging ranges for taxa that 335 are similar morphologically, and therefore difficult to distinguish at sea (Rayner et al. 2011). Moreover, this utility is particularly relevant for *caledonica* and *leucoptera* with significantly different 336 populations sizes (caledonica ~ 10 000 breeding pairs, leucoptera ~ 1000 breeding pairs; (Brooke 337 2004, Priddel & Carlile 2007), making it particularly important to determine foraging areas and 338 339 ensure a balanced assessment of at-sea threats.

340 Large scale spatial segregation is an important component of resource partitioning by small

341 procellariform seabirds at high latitudes, as demonstrated by recent studies linking divergent

342 distributions with species-specific preferences for particular sea surface temperature regimes,

frequently partitioned across oceanic fronts (Navarro et al. 2013, Navarro et al. 2015, Quillfeldt et al.

344 2015). The ETPO lacks the strong latitudinal gradients in temperature typical of higher latitude
345 regions in the Southern Ocean. Instead, the SDMs for our study taxa indicate that spatial segregation
346 was paralleled by subtle taxon-specific differences in habitat niche, across a range of environmental
347 predictors.

Functional relationships with depth of the thermocline (reflecting a subsurface gradient in temperature 348 below the warm surface mixed layer), SST and depth indicates a division into two strategies of habitat 349 350 use by the three taxa; leucoptera and pycrofti foraged in deep and warm waters where the depth of 351 the thermocline was the strongest predictor of presence yet where both species had opposing functional relationships with thermocline depth: leucoptera preferring a deeper thermocline and 352 353 *pycrofti* a shallower one. There is evidence that in the ETPO, the thermocline depth is a strong 354 predictor of the abundance and distribution of other seabirds (Ballance et al. 1997, Spear et al. 2001, 355 Ballance et al. 2006). In the NECC, vertically-migrating prey aggregate at the top of the thermocline 356 and are frequently driven to the surface by feeding tuna and dolphins and thus exploited by a range of diurnally-feeding seabirds (Ballance et al. 2006). Leucoptera and pycrofti in these habitats are 357 358 solitary foragers and not considered parts of this tuna-dolphin-seabird feeding assemblage (Spear et al. 2007), yet their presence was nevertheless predicted strongly by thermocline depth, indicating that 359 similar ecological processes influence their distribution. Conversely, the presence of *caledonica* in 360 361 the SEC was associated with cooler SSTs and shallower waters, particularly over the east Pacific rise, 362 where thermocline depth was a weak predictor. Bathymetric features associated with the east Pacific rise in this region likely provide foraging opportunities as a result of upwelling that are targeted by 363 various Pterodroma species (Rayner et al. 2012). 364

In the ETPO low iron availability reduces primary production resulting in a low chla environment (Pennington et al. 2006). The responses of the tracked birds to chla gradients were weak and indicated the use of waters with generally uniform and low chla values, particularly by *leucoptera* and *pycrofti* which maintained core habitats in oligotrophic waters of the NECC. This result is inconsistent with the results of studies demonstrating spatial relationships between chla, prey and predators in the ETPO (Ballance et al. 2006) and at higher latitudes where chla is more abundant 371 (Weimerskirch 2007, Peron et al. 2010). The analysis of immersion data from the three study taxa 372 showed a striking pattern of reduced time resting on the surface, more numerous and longer flight 373 bouts and higher variance in activity parameters during darkness than daylight. These results suggest 374 that although our study taxa may forage opportunistically during daylight, their primary foraging 375 strategy in the ETPO is to exploit nocturnally available prey, which is consistent with previous 376 research on tropical (Spear et al. 2007, Pinet et al. 2011, Ramirez et al. 2013)) and temperate procellariiform seabirds (Imber 1973, Imber 1996, Rayner et al. 2012). In a nine-year study (1983-377 1991), Spear et al (2007) demonstrated that the diet of small procellariform species in the ETPO, 378 including *leucoptera*, was dominated (>90%) by vertically-migrating mesopelagic fishes that were not 379 associated with diurnally feeding surface predators. Although our comparison of nitrogen stable 380 381 isotope ratios in feathers grown during the nonbreeding period suggest little dietary segregation 382 between *leucoptera* and *pycrofti*, which occupy waters north of the equatorial front and NECC, the mean δ^{15} N values in these two taxa were higher than in *caledonica*, which occupy habitats in the 383 SEC. Values of 13 to 18% for δ^{15} N in *leucoptera* and *pycrofti* feathers indicates that the tracked birds 384 were foraging at a similar trophic level to tropical piscivores (Young et al. 2010), confirming a diet 385 386 likely dominated by mesopelagic fishes (myctophids, bregmacerotids, diretmids, and melamphaids)(Spear *et al.* 2007). In contrast, the lower δ^{15} N of *caledonica* (11.0 to 14.0%) suggests 387 the additional consumption of prey from lower trophic levels, such as cephalopods or marine 388 crustaceans (δ^{15} N of 8.1 to 10.2‰, and 3.6 to 6.5‰,)(Quillfeldt et al. 2005). This interpretation 389 assumes there are no differences in regional baselines for δ^{15} N, which is supported by the similarity in 390 δ^{13} C values for all three study taxa, which is consistent with foraging in deep tropical waters within a 391 392 narrow latitudinal range (Hobson et al. 1994, Young et al. 2010).

In conclusion, our study of three closely-related *Pterodroma* petrels in the ETPO is consistent with theoretical and empirical data on niche overlap, which predicts divergence in habitat use, diet or behaviour among competitors that are sympatric in low resource environments. The slight behavioural differences appear unlikely to reduce inter-specific competition given the high morphological similarities between the three taxa. Rather, competition appears to be avoided by 398 habitat segregation. The SDMs provide evidence that the spatial separation between these three taxa is driven by differences in realized habitat niches. However unlike temperate systems, where seabird 399 distributions can be predicted by strong surface temperature, wind or productivity gradients, two of 400 three of our species were present in oligotrophic waters north of the equator, and the overall 401 402 distribution of our study taxa was mediated by subtle horizontal and vertical temperature gradients, as well as depth. Thermocline depth in particular was a strong predictor of presence, and its role appears 403 linked to the unique foraging niche of the study taxa. All three taxa were highly active during 404 darkness, and thermocline depth likely plays a critical role by governing the proximity to the surface 405 of the key prey source, which are vertically-migrating mesopelagic fishes. The nocturnal foraging 406 niche of the *Pterodroma* petrels parallels the diurnal tuna-dolphin-seabird assemblages that make use 407 of shallow thermoclines in the same region (Spear et al. 2007). Our study highlights the unique 408 409 insights that can be gained from comparative studies of foraging ecology between pelagic seabirds in tropical systems and temperate or polar waters, and reinforces the importance of the ETPO as a 410 411 critical habitat for numerous small Procellariiformes that breed in temperate regions. Management 412 programs seeking to protect key marine habitats for the many highly threatened temperate and tropical 413 Pterodromas would benefit from further tracking to map foraging areas during the breeding and non-414 breeding seasons, and to identify oceanographic drivers and their impacts on distributions.

415

416 Acknowledgements:

We thank Todd Dennis, and New Zealand Department of Conservation Staff Graeme Taylor, Rodd Chappell and Shane McInnes for assistance with field work and Craig Donovan (Waters Edge Charters) for boat transportation. We thank Nick Sard for assisting with plots in R and Julie Brown and Anna Kilimnik at the National Institute of Water & Atmospheric Research New Zealand for analysing the stable isotope samples. MJR was supported by a Ministry of Research Science and Technology post-doctoral fellowship during the completion of this research and acknowledges the

- 423 support of Wendy Rayner during its completion. This paper is dedicated to the memory of "field
- 424 legend", Mr Vince Vaanders (23/03/1970 18/08/2013).

425 Table 1. Environmental variables used in construction of boosted regression tree models.

Variable (units)	Product code	Temporal	Spatial	Data source
		resolution	resolution (degrees)	
Sea Surface Temperature (°C)(SST)	TMHchla8day	8-day	0.05	MODIS AQUA www.oceancolor.org
Chlorophyll-a Concentration	TMHsstd8day	8-day	0.05	MODIS AQUA www.oceancolor.org
(mg/m^3) (Chl)				
Sea surface height deviation anomaly	TTAsshd1day	1-day	0.25	DUAACS AVISO www.aviso.oceanobs.com
(m)(SSHD)				
Wind speed (m/s)(Wind)	TQAumod3day	3-day	0.25	METOP ASCAT www.eumetsat.int
Ekman upwelling (m/day)	TQAwekm3day	3-day	0.25	METOP ASCAT www.eumetsat.int
Top of thermocline depth (m)		Monthly	2	IFREMER
		climatology		http://www.ifremer.fr/cerweb/deboyer/mld
Depth (m)		Static	0.0083	GEBCO
				www.gebco.net
Seabed slope angle (°)(Slope)		Static (derived)	0.0083	GEBCO
				www.gebco.net
Distance to nearest land (km)		Static (derived)	0.0083	GEBCO
				www.gebco.net

429 Table 2. Timing of arrival and departure in core nonbreeding habitats for *leucoptera*, *caledonica*, and *pycrofti* in the eastern tropical Pacific Ocean in 2010. P values indicate

- 430 significance of contingency analysis to assess differences between species in arrival and departure dates, and time spent in core areas. Data are shown as means ± standard
- 431 deviation in days, with range in parentheses.

	P. l. leucoptera	P.l. caledonica	P. pycrofti	Р
Arrival nonbreeding habitat	24 Apr ± 9.8 days (7 Apr – 12 May) n = 14	18 May ± 13.0 days (4 May – 13 Jun) n = 7	21 Apr ± 12.35 days (5 Apr – 14May) n = 8	0.20
Departure nonbreeding habitat	14 Sep ± 18.7 days (15 Aug – 16 Oct) n = 9	4 Oct ± 23.5 days (31 Aug - 28 Oct) n = 7	$5 \text{ Sep} \pm 4.24 \text{ days}$ $(2 \text{ Sept} - 13 \text{ Sept})$ $n = 5$	0.51
Time in nonbreeding core	$141.3 \pm 17.6 \text{ days}$ (113-161) n = 9	135.9 3 ± 31.1 days (78-174) n = 7	137.0 ± 16.6 days (110-152) n = 5	0.13

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443 Table 3.Boosted regression tree (BRT) model parameters and validation results for *leucoptera*, *caledonica* and *pycrofti*. Thermocline = thermocline depth; SST=sea surface

temperature; Slope = seabed slope angle; Chl = chlorophyll a concentration; SSTad = sea surface temperature absolute deviation around median; SSHD = sea surface height

445 deviation; Wind = wind speed. AUC varies from 0 to 1, with 1 indicating perfect model fit, 0.5 indicating random assignment. Cross validated deviance represents the mean

446 residual deviance per fold across the whole BRT model (lower values denotes better fit, but values cannot be compared between models). Validation deviance indicates the

447 mean residual deviance between the withheld presence and absence values (1 or 0) and model predicted values for those points (higher values denote better model fit). rS

448 indicates Spearman's rank correlation derived from k-fold cross validation of withheld presence points from each model.

Species	Number of individuals tracked (total presence points)	Parameters (% contribution)	# of interactio ns	learning rate	# trees	Cross validated deviance (internal)	Cross validated AUC (internal)	Validation deviance (external)	Validation AUC (external)	rS (P- value) (external)
Leucoptera	22 (1963)	Thermocline (43.0) SST (22.0) Depth (14.3) log(Slope) (9.1) log(Chl) (8.6) SSHD (3.1)	2	0.0025	1350	0.151	0.765	0.938	0.841	0.9904 (<0.0001)
Caledonica	7 (1102)	SST (36.0) Depth (26.8) log(Chl) (15.7) Thermocline (9.4) SSTad (6.6) SSHD (5.5)	2	0.0025	1800	0.207	0.822	0.671	0.888	0.9880 (<0.0001)
Pycroft	8 (1659)	Depth (33.3) Thermocline (26.7) log(Chl) (14.8) SST (10.4) SSHD (9.6) Wind (5.2)	2	0.0025	4550	0.294	0.829	0.952	0.833	0.9893 (<0.0001)

- 452 Table 4. Activity patterns of *leucoptera*, *caledonica* and *pycrofti* tracked with geolocator-immersion loggers in
- 453 the tropical Pacific during the nonbreeding period. Flight bouts constitute periods where loggers were dry for ten
- 454 minutes or longer. Values are the mean \pm SD.

						465	
	Proportion of time spent on water (%)		Number of fl	ight bouts	455 Duration of flight bouts456 (min) 457		
	Daylight	Darkness	Daylight	Darkness	Daylight	Darkness 458	
leucoptera	75.6 ± 6.5	28.9 ± 17.4	2.0 ± 0.7	5.6 ± 1.4	25.1 ± 5.0	^{53.6 ± 1} 439	
caledonica	75.8 ± 4.6	7.9 ± 4.6	1.7 ± 0.5	7.2 ± 0.5	21.7 ± 5.0	67.0 ± 1 4.60	
pycrofti	83.1 ± 11.1	51.0 ± 18.3	2.4 ± 1.5	4.5 ± 1.3	25.8 ± 8.9	461 45.6 ± 7.9	
		•	•		•	462	

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467 Figure 1 option 1. Locations and general post-breeding (solid lines) and pre-breeding (dashed lines) migration

routes of (a) *leucoptera*, (b) *caledonica* and (c) *pycrofti* tracked with geolocators between March and November
2010. Locations shown in bold colour are those that were within monthly 50% kernels during the nonbreeding

470 season (Apr-Oct) and used as presence data in the species distribution models.



Figure 1 option 2. Locations and general post-breeding (solid lines) and pre-breeding (dashed lines) migration
routes of *leucoptera*, (green) *caledonica* (blue) and *pycrofti* (red) tracked with geolocators between March and
November 2010. Locations are those that were within monthly 50% kernels during the nonbreeding season
(Apr-Oct) and used as presence data in the species distribution models. Respective species colonies shown as
squares coloured as above.



482 Figure 2. Nonbreeding distribution of leucoptera, caledonica and pycrofti between April and October 2010 483 overlaid on averaged oceanographic climatologies for the month of July. The 90% (black dashed lines) kernel 484 contours of all species locations and the 50% (coloured solid lines) for each species from April to October: 485 leucoptera (green lines), caledonica (blue lines) and pycrofti (red lines). The environmental layers are ordered 486 by collective contribution to all three species models: A) Thermocline depth (m), B) Depth (m), C) Sea surface 487 temperature (C°) and D) Chlorophyll-a concentration (mg/m³). Dashed lines represent approximate locations of 488 the North Equatorial Current (NEC), North Equatorial Counter Current (NECC) and South Equatorial Current 489 (SEC) adapted from (Pennington et al. 2006).



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Figure 3.Comparison of fitted functions derived from presence-availability boosted regression tree models of *leucoptera* (green lines), *caledonica* (blue lines) and *pycrofti* (red lines) in relation to the four most influential predictor variables across all taxa. Y axes represent the relative effect of each predictor variable (x axes) on petrel habitat use while fixing all other variables to their mean value. Positive Y-axes values represent a positive contribution by the predictor variable to species presence, and negative values indicate a negative contribution. The percent contribution of each predictor variable to each species model is given by labels in plot, coloured as per species.

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Figure 4. Mean monthly activity metrics during daylight (clear symbols) and darkness (filled symbols) for
 leucoptera (triangles), *caledonica* (squares), and *pycrofti* (circles) tracked with geolocator-immersion loggers
 during the nonbreeding season including A) Percent of time on water, B) Number of flight bouts and C)
 Duration of flight bouts.



527 Figure 5. Feather stable isotope ratios of *leucoptera* (green square, n = 12), *caledonica* (blue triangle, n = 7), and

pycrofti (red circle, n = 10) tracked with geolocator-immersion loggers during the non-breeding season.

- 529 Coloured symbols and errors are the mean \pm s.d., and unfilled symbols correspond to individual values.

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