

1 **Consistency in migration strategies and habitat preferences of brown skuas over two**  
2 **different years, a decade apart**

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4 Running page head: Consistency in skua migration

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24 ABSTRACT: At-sea movements and activity patterns of brown skuas *Stercorarius*  
25 *antarcticus lonnbergi* from South Georgia were analysed in two winters, a decade apart, to  
26 examine the degree of consistency in migration strategies and habitat preferences during the  
27 non-breeding and pre-laying exodus periods. Oceanographic habitat preferences of tracked  
28 skuas were determined using a robust model accounting for availability. At the population  
29 level, brown skuas were broadly consistent in their choice of wintering areas and habitat  
30 preferences, although the distribution extended further east in 2012 than in 2002. Skuas  
31 preferred areas associated with static oceanography (bathymetric features) both during the  
32 non-breeding and pre-laying periods, which may explain the consistency between years in  
33 habitat use. There was no significant effect of year on departure dates from South Georgia,  
34 but birds returned earlier to the colony in 2002. Migration schedules varied according to  
35 breeding status, with failed birds departing earlier than birds that bred successfully. Although  
36 failed birds travelled further from the colony, there was little variation in dates of return. In  
37 general the timing of movements was similar between sexes, but females were more likely  
38 than males to engage in a pre-laying exodus. Brown skuas spent a much higher proportion  
39 of time sitting on the water than other seabirds during both the non-breeding and pre-laying  
40 exodus periods, and the number of flight bouts per day was surprisingly low. The selection of  
41 static features by brown skuas may indicate that skuas may have less flexibility to track  
42 environmental changes than species that use dynamic cues.

43 KEY WORDS: Annual variability, biologging, GAMs, geolocation, GLS, habitat selection,  
44 modelling, seabird

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## 54 INTRODUCTION

55 Seabirds spend most of their lives at sea, only returning to land to breed. Understanding the  
56 factors that influence their marine distribution and habitat preferences is therefore essential  
57 for mitigating threats at sea, including the establishment of an effective, evolving network of  
58 protected areas that will safeguard against future biodiversity loss (Block et al. 2011,  
59 Frederiksen et al. 2012). The open ocean is a dynamic environment, and seabirds frequently  
60 target areas of higher prey availability (Pinaud & Weimerskirch 2005, Nur et al. 2011,  
61 Quillfeldt et al. 2013). These regions are often characterised by local physical features or  
62 processes, including eddies, frontal systems, upwelling zones and shelf breaks, that  
63 increase primary production or serve to aggregate animals at higher trophic levels  
64 (Wakefield et al. 2009, Kappes et al. 2010, Pinet et al. 2011, Louzao et al. 2011a).  
65 Reductions in size, and improvements in battery-life and functionality of tracking devices  
66 since the 1990s have greatly improved our knowledge of at-sea distribution patterns and  
67 behaviour of marine predators (Phillips et al. 2008). More recently, studies that combine  
68 tracking data with remotely sensed environmental data in habitat models have provided key  
69 insights into the oceanographic factors that drive the distribution of seabirds (Péron et al.  
70 2010, Louzao et al. 2011b, Wakefield et al. 2011, Quillfeldt et al. 2015).

71 A limitation of many biologging studies is that data are available from relatively few  
72 individuals tracked over short periods of time (Žydelis et al. 2011). In addition, as most  
73 physical characteristics of the ocean are dynamic at varying temporal and spatial scales,  
74 prey, and hence predator, distributions are expected to vary accordingly (Pinaud et al. 2005,  
75 Žydelis et al. 2011, Quillfeldt et al. 2013). Although there is evidence in some species that  
76 individuals can change their non-breeding destination (e.g. Dias et al. 2011), the general  
77 trend appears to be for high regional site fidelity among migrant seabirds (Croxall et al. 2005,  
78 Phillips et al. 2005, 2006, Thiebot et al. 2011, Guilford et al. 2011, Yamamoto et al. 2014).  
79 There is also good evidence for high year-to-year consistency in migration schedules (timing  
80 and duration of events) for birds that are faithful to their winter destinations (Croxall et al.  
81 2005, Phillips et al. 2005, Dias et al. 2011, Yamamoto et al. 2014). It is important to note,  
82 however, that in all these studies, individuals were tracked for successive years, when  
83 conditions may be more similar than after longer intervals.

84 Individuals from the same breeding population frequently use multiple regions or habitats  
85 during their non-breeding period (Phillips et al. 2005, Kopp et al. 2011, Dias et al. 2011).  
86 These alternative strategies are often associated with differences in age, breeding status,  
87 sex, or individual preferences (Phillips et al. 2005, Ramírez et al. 2013, Quillfeldt et al. 2015).  
88 Migration schedules can also vary, especially in relation to sex or prior breeding outcome

89 (Phillips et al. 2005, Bogdanova et al. 2011, Catry et al. 2013a). In addition, some non-  
90 breeding regions may be used by a small minority of animals: only two of 34 black-browed  
91 albatrosses *Thalassarche melanophris* tracked from South Georgia did not migrate to the  
92 Benguela Upwelling but instead spent the non-breeding period on the Patagonian Shelf or in  
93 Australian waters (Phillips et al. 2005); and, only two of 39 south polar skuas *Stercorarius*  
94 *maccormicki* tracked from King George Island wintered in the southern hemisphere, one off  
95 Peru and the other west of Gough Island (Kopp et al. 2011). Thus, to better identify core  
96 areas for the entire population, and also to improve the accuracy of predicted distributions  
97 based on observed habitat preferences accounting for future environmental changes,  
98 tracking studies should ideally involve large numbers of individuals in multiple years (Hindell  
99 et al. 2003, Soanes et al. 2013).

100 The brown skua *Stercorarius antarcticus lonnbergi* breeds on islands from the subantarctic  
101 to the Antarctic continent, and is one of the main terrestrial consumers, primarily feeding on  
102 other seabirds, and seal carrion (Phillips et al. 2004a, Carneiro et al. 2014, 2015). To date,  
103 non-breeding ranges of brown skuas have been inferred from the distribution of just six birds  
104 tracked from South Georgia during the winter of 2002 using global location sensors (GLS);  
105 these birds dispersed over deep, subantarctic and mixed subantarctic-subtropical waters  
106 within the Argentine Basin (Phillips et al. 2007). However, stable isotope analyses of  
107 feathers from a larger sample suggested that some individuals may use continental shelf or  
108 shelf-slope waters (Phillips et al. 2009). In the present study, GLS-immersion loggers were  
109 deployed on adult brown skuas from the same population, but different individuals, in order  
110 to: (1) assess if distribution or other aspects of migration strategies have changed in the ten  
111 years since the first study, (2) identify key habitat preferences, (3) better characterise the  
112 migration period in terms of timings, durations and travel distances, (4) investigate  
113 differences in distribution and migration characteristics in relation to sex and recent breeding  
114 performance, (5) compare at-sea activity patterns between male and female non-breeders,  
115 and (6) describe the pre-laying exodus.

116

## 117 **MATERIALS AND METHODS**

118 This study was carried out during two non-breeding seasons a decade apart on brown skuas  
119 at Bird Island, South Georgia (54° 00' S, 38° 03' W). Bird Island is probably the most  
120 densely-populated colony of brown skuas in the world, with 132 nesting pairs per km<sup>2</sup>  
121 (Phillips et al. 2004a). A GLS-only logger (weight 7 g) or a combined GLS-immersion logger  
122 (weight 9 g) was deployed in austral summer 2001/02 on 28 breeding adults. Combined

123 GLS-immersion loggers (weight 1.5 g) were deployed on 25 breeding adults, none of which  
124 were in the earlier sample, in the same study area in austral summer 2011/12. All loggers  
125 were attached with a cable-tie to a British Trust for Ornithology (BTO) metal ring on the  
126 tarsus and recovered after a year. Skuas were captured on the ground by hand or using a  
127 hand net or noose pole. Only data encompassing the non-breeding and pre-laying exodus  
128 periods are analysed here. The non-breeding period was defined as the time from the start  
129 of the outward migration to the return to the breeding grounds the following season, and was  
130 assessed for each individual using both location and activity data. Departure date was  
131 identified as that immediately prior to a directed movement away from South Georgia and  
132 consecutive nights thereafter that were spent largely on the water, whereas return date was  
133 that preceding the first night spent dry (on land). As skuas with eggs or chicks at South  
134 Georgia never forage at sea, after the initial return from migration, subsequent periods of  
135 consecutive days with intervening nights spent largely on the water were classified as pre-  
136 laying trips (Phillips et al. 2007, Carneiro et al. 2014).

137 Light data were analysed using the BASTrak suite based on times of sunset and sunrise  
138 estimated from thresholds in light curves; latitude was derived from day (night) length, and  
139 longitude from the time of local midday (midnight) in relation to Greenwich Mean Time and  
140 day of the year, providing two locations per day with an average accuracy of  $186 \pm 114$  km  
141 (Phillips et al. 2004b). Locations were excluded for 2 to 4 weeks around the equinoxes,  
142 when latitudes were unreliable. All other analyses were carried out using R software (R Core  
143 Team 2014). A hierarchical state-space model fitted to data from multiple animals was used  
144 to filter and correct observed locations for logger error (Jonsen et al. 2013, Gutowsky et al.  
145 2014). Estimates of uncertainty for each latitude and longitude are required in order to fit  
146 state-space models (Winship et al. 2012). As there are no relevant published data for skuas  
147 from double-tagging experiments, a fixed geolocation error (SD of latitudinal and longitudinal  
148 error:  $1.66^\circ$  and  $1.82^\circ$ , respectively; Phillips et al. (2004b)) derived from concurrent  
149 deployment of satellite-transmitters and GLS loggers on albatrosses was used as an  
150 estimate of uncertainty. The state-space model was fitted using Markov Chain Monte Carlo  
151 (MCMC) sampling in the bsam package (Jonsen et al. 2015). Two chains of 5,000 samples  
152 from the joint posterior probability distribution were obtained after discarding the initial burn-  
153 in period of 100,000 samples, and retaining every 20<sup>th</sup> of the remaining samples.  
154 Convergence was assessed visually by checking trace, density and autocorrelation plots  
155 (Pollet et al. 2014). After processing, locations derived from curves with apparent  
156 interruptions around sunset and sunrise which were not filtered by the state-space model  
157 were removed after visualisation in ArcGIS v. 10.2.2.

158 Changes in the spatial distribution of brown skuas between non-breeding and pre-laying  
159 periods in 2002 and 2012 were investigated by producing 95 % (general use) and 40 %  
160 (core) utilization distributions (UDs) for each individual using kernel analysis with a cell size  
161 of 50 km and a fixed smoothing parameter (h) of 200 km (Phillips et al. 2006). The 40 %  
162 density contour was selected by noting where the relationship began to increase  
163 exponentially when incremental changes in UD estimates were plotted against the size of the total  
164 area identified, which is considered to be a reliable approach for defining core areas of  
165 activity (Lascelles et al. 2016). Population core and general use areas were created by  
166 merging individual UD estimates, assigning them equal weighting. Individual overlaps were quantified  
167 using Bhattacharyya's affinity, which ranges from 0 (no overlap) to 1 (identical UD estimates) and is  
168 considered to be the most appropriate index for quantifying the degree of similarity among  
169 UD estimates (Fieberg & Kochanny 2005). Kernel analysis and individual overlap  
170 calculations were carried out using the adehabitatHR package (Calenge 2006). Analyses of  
171 similarity (ANOSIM) were used to test for significant differences in the 95 % and 40 %  
172 overlaps by year, sex, and breeding status (failed or successful birds in the year of  
173 deployment) using the vegan package (Oksanen et al. 2015). The dissimilarities between the  
174 UD estimates were included in a distance matrix (also known as a dissimilarity matrix). ANOSIM uses  
175 a bootstrap randomization procedure to test for differences between groups; if two groups  
176 are different, then dissimilarities between the groups should be greater than those within the  
177 groups (Oksanen et al. 2015). R values from ANOSIM are equivalent to a correlation  
178 coefficient and range from -1 to 1, with 0 indicating completely random grouping.

179 Oceanographic habitat preferences of the tracked skuas during the non-breeding and pre-  
180 laying periods were determined using a case-control approach in binomial generalized  
181 additive models (GAMs). For each animal location, we simulated 50 temporally-matched  
182 pseudo-absences (controls) representing accessible areas where animals could potentially  
183 have travelled (see e.g. Wakefield et al. 2011, Žydelis et al. 2011, Raymond et al. 2015).  
184 These were created using correlated random walks (CRWs), based on the distribution of  
185 observed turning angles and distances between successive locations, using the  
186 adehabitatLT package (Calenge 2006). A constraint function defined by the minimum convex  
187 polygon (MCP) of all observed locations, increased by a 200 km buffer to account for GLS  
188 logger error was used to restrict the CRWs (Žydelis et al. 2011). The number of simulated  
189 locations was determined by measuring changes in  $x^2$  values of each parameter by running  
190 global models with all observed locations and varying numbers of simulated pseudo-  
191 absences (Žydelis et al. 2011). Fifty CRWs were sufficient for most parameters, even though  
192 a few would have ideally required even more CRWs. However, and given the computational

193 requirements associated with large numbers of CRWs, we used 50 for all our models (see  
194 electronic supplementary material, S1).

195 A set of oceanographic variables were included as candidate predictors in the habitat  
196 models on the basis of their biological relevance, i.e., their potential influence on skua  
197 distribution. To account for geolocation error, a buffer of 200 km was created around each  
198 observed and simulated location. A spatially weighed average of all oceanographic data was  
199 calculated for each of those locations. Potential predictors were: (1) depth (a proxy for  
200 coastal or pelagic domains), extracted from ETOPO1 at a spatial resolution of 0.01°,  
201 downloaded from <https://www.ngdc.noaa.gov/mgg/global/global.html>, and, for 2002 and  
202 2012, (2) chlorophyll *a* (CHL *A*, a surrogate of marine productivity), 8-day composites  
203 derived from SeaWiFS and Aqua MODIS, respectively, with a 0.09° spatial resolution, (3)  
204 sea surface temperature (SST, a proxy for water mass), a product from NOAA Pathfinder  
205 Advanced Very High Resolution Radiometer (AVHRR) v. 5.2 at 0.04° spatial and daily  
206 temporal resolutions, (4) eddy kinetic energy (EKE) and (5) sea level anomaly (SLA), which  
207 are indices of mesoscale activity, were daily delayed time AVISO DUACS products with a  
208 0.3° spatial resolution. All remotely-sensed variables were extracted using the Marine  
209 Geospatial Ecology Tools (MGET) for ArcGIS v. 10.2.2 (Roberts et al. 2010). As the volume  
210 of missing data prevented the use of daily composites, an average of three 8-day or 24 1-  
211 day composites centred on the date of the location was created for the dynamic predictors.  
212 After creating the composites, values for CHL *A* were still missing for ~ 24 % of locations.  
213 Such a large proportion is likely to affect inference because of the systematic pattern in the  
214 occurrence of missing values (Wakefield et al. 2011), therefore, CHL *A* was discarded as a  
215 predictor in the analyses. All composites were created from customized functions and the  
216 raster package in R. Additionally, spatial gradients of SST (SST gradient, an indicator of the  
217 presence of frontal systems) and depth (depth slope, a proxy for topographic features) were  
218 calculated as the standard deviation of the mean.

219 Prior to GAM analysis, predictors were tested for collinearity by calculating all pairwise  
220 Spearman rank correlations ( $r_s$ ), and when  $|r_s| > 0.5$ , predictors were not included in the  
221 same model. Habitat selection was examined using GAMs because data exploration  
222 indicated potentially non-linear relationships between response and predictors. GAMs take  
223 into account non-linear relationships using non-parametric smoothers to fit flexible curves to  
224 data (Aarts et al. 2008), and were implemented within the mgcv package (Wood 2006).  
225 Smooth functions for model predictors were specified using cubic regression splines with  
226 shrinkage to avoid overparameterization and to identify the most parsimonious number of  
227 knots (Wood 2006). The number of knots, representing maximum degrees of freedom of

228 each smooth, was manually limited by  $k = 4$  to avoid excessive flexibility and model  
229 overfitting that would have no ecological meaning (Mannocci et al. 2014). Due to the  
230 dynamic nature of oceanographic predictors and possible changes in habitat preferences,  
231 year, but not sex or breeding status, was included as an interaction in the models.

232 The best minimum models were determined by forward selection using  $k$ -fold cross-  
233 validation (Wakefield et al. 2011), testing the goodness of fit of each individual, in turn,  
234 against the prediction based on the other 24 individuals. Cross-validation was preferred over  
235 information criteria, e.g. the Akaike Information Criterion (AIC), as the latter can lead to  
236 overparameterized models because of violation of the independence assumption inherent  
237 within tracking datasets (Aarts et al. 2008, Cleasby et al. 2015). Cross-validation prevents  
238 overparameterization by using a set of data for fitting the model and another set for  
239 assessing predictive performance (Aarts et al. 2008). Each model was trained on 24 folds  
240 and tested, in turn, on the remaining, withheld fold which represented an individual. Model  
241 selection was based on the predictive ability of the model using the area under the receiver  
242 operating curve (AUC) averaged across the 25 sets of results (i.e., individuals) using the  
243 pROC package (see Raymond et al. 2015). AUC values between 0.5-0.7 indicate poor  
244 model predictive ability, values from 0.7-0.9 indicate reasonable model discrimination ability,  
245 and values higher than 0.9 indicate very good model discrimination (Pearce & Ferrier 2000).  
246 The forward selection procedure consisted initially of fitting all possible single predictor  
247 models with and without the year-interaction and ranking those models according to AUC.  
248 The best ranking model was chosen, and then, each of the remaining predictors was added  
249 in turn; the best model among this new set was then retained if the AUC increased  
250 significantly. This process was continued until there was no further increase in the AUC. The  
251 significance of the increase in AUC between two models was tested with a paired  $t$ -test.  
252 Using cross-validation and paired  $t$ -tests to compare AUCs is somewhat conservative;  
253 however, we considered this approach to be the best solution to avoid overparameterization,  
254 as it prevents adding parameters that only contribute trivially to increasing AUC. Habitat  
255 preference models were fitted separately for the non-breeding and pre-laying periods.

256 Timing (departure and return dates, and duration), the size of the core and general use  
257 areas (40 % and 95 % UDs), travel speeds and distances (summed great circle distances  
258 between the locations) were compared between birds of different sex and breeding status  
259 during the pre-laying exodus and migration periods. The effect of year was included in  
260 models for the migration but not pre-laying period because of the reduced sample size (only  
261 two birds engaged in a pre-laying exodus in 2002). All main effects and interactions were  
262 tested in linear models. AIC values were used to rank all possible models according to their



263 degree of parsimony; the best model was the one with the lowest AIC. If two or more models  
264 were within 2 AIC units from the best-supported model, differed by one parameter and had a  
265 similar maximized log-likelihood value, only the most parsimonious model was selected. The  
266 latter suggests that improvements in model fit have not been enough to overcome the  
267 penalty of an additional parameter (Burnham & Anderson 2002, Arnold 2010). Unless  
268 indicated otherwise, all data are presented as mean  $\pm$  SD.

269

## 270 **At-sea activity analysis**

271 The combined GLS-immersion loggers tested for saltwater immersion every three seconds.  
272 The number of positive tests was recorded for each 10-min period, providing a value that  
273 ranged from 0 (continuously dry) to 200 (continuously wet). Each 10-min period was  
274 categorised as daylight or darkness using sunset and sunrise times estimated from the  
275 thresholds in light curves recorded by the same loggers. The activity data were used to  
276 determine the proportion of time spent flying during daylight and darkness, and daily (each  
277 consecutive light and dark period). The proportion of the total time spent on water, and time  
278 spent in flight in each day that occurred during darkness, as well as the duration and number  
279 of flight bouts per day, were also calculated. A flight bout was defined as any continuous  
280 sequence of 10-min periods of completely dry (0) values. This method is likely to  
281 underestimate the number of flight bouts as birds may engage in multiple short periods of  
282 flight within 10 minutes. However, Phalan et al. (2007) found a close correlation between  
283 bouts derived from lower and higher resolution loggers (which record immersion every 10  
284 seconds), indicating that bouts are adequate as proxies of activity. At-sea activity  
285 characteristics were compared between sexes, and between daylight vs. darkness using  
286 linear mixed-effect models with individual identity included as a random effect. Year was not  
287 included in the models because of the limited sample size for 2001/02 (only three birds for  
288 the non-breeding period and two birds for the pre-laying exodus). Proportion data were  
289 arcsine transformed. Models including all combinations of variables as well as their  
290 interactions were tested using AIC values following the approach described above.

291

## 292 **RESULTS**

293 Eight out of 28 breeding adults (29 %) fitted with loggers in summer 2001/02 returned to  
294 breed in 2002; seven loggers were retrieved (one bird was not recaptured), of which one  
295 failed to download. Immersion data were only available for three of the loggers. The low

306 return rate of instrumented birds was attributed to the accretion of goose barnacles *Lepas*  
307 spp. to the self-amalgamating tape wrapped around the logger causing an increase in load  
308 on the leg, and is discussed in Phillips et al. (2007). Of 25 birds fitted with a logger in  
309 2011/12, 22 returned to breed in the following season, from which 19 loggers were retrieved  
300 (three loggers were lost). Return rates were higher for birds with devices (22 of 25; 88 %)   
301 than for control birds (60 of 82; 73 %), all fitted with a plastic colour-ring with a unique alpha-  
302 numeric sequence in the first season, but the difference was not statistically significant  
303 (Fisher's exact test,  $P = 0.18$ ).

304

### 305 **Non-breeding period**

306 Tracked skuas started their outward migration on 4 March  $\pm$  19 days (range: 3 February to  
307 22 April,  $n = 24$ ). There was no significant effect of year on mean departure date (Table 1),  
308 but birds returned to South Georgia ca. 10 days earlier in 2002, on 5 October  $\pm$  8 days  
309 (range: 30 September to 15 October,  $n = 3$ ), compared with 16 October  $\pm$  10 days (range: 26  
310 September to 4 November,  $n = 19$ ) in 2012. The migration period lasted ca.  $225 \pm 20$  days  
311 (range: 192 to 270,  $n = 22$ ), and was similar between years (Table 1). Departure dates and  
312 durations of the migration period, but not return dates, varied according to breeding status  
313 and sex (Table 1). Excluding one bird that failed in early February and only started its  
314 outward migration on 4 April, failed birds departed earlier than birds that bred successfully,  
315 on 21 February  $\pm$  16 days (range: 3 February to 31 March,  $n = 12$ ) and 16 March  $\pm$  15 days  
316 (range: 1 March to 22 April,  $n = 12$ ), respectively (Table 1). Males departed 3 to 4 weeks  
317 later than females, on 17 March  $\pm$  18 days (range: 19 February to 22 April,  $n = 10$ ),  
318 compared with 23 February  $\pm$  14 days (range: 3 February to 22 March,  $n = 15$ ; Table 1);  
319 however, the effect of sex probably only reflects differences associated with breeding status,  
320 as the proportion of tracked males that bred successfully was higher than the proportion of  
321 females (60 % vs. 40 %). The mean distance travelled during migration was  $391.9 \pm 71.6$  km  
322 per day (range: 272.9 to 540.2 km,  $n = 24$ ), with no effect of year, sex or breeding status  
323 (Table 1). Maximum distances from the colony varied according to breeding status, ranging  
324 from 1,824 to 5,066 km ( $3,009 \pm 1,043$ ,  $n = 13$ ) for failed birds, and from 1,445 to 3,363 km  
325 ( $2,400 \pm 489$  km,  $n = 11$ ) for successful birds. There was no difference in the maximum  
326 distance travelled from the colony between 2002 and 2012, nor between males and females  
327 (Table 1).

328 Skuas were distributed over a wide area (95 % UDs) during the non-breeding period,  
329 ranging from 2,437,500 to 9,455,000 km<sup>2</sup> ( $4,442,500 \pm 1,470,068$  km<sup>2</sup>,  $n = 24$ ; Fig. 1). The

330 distribution was limited to waters between the northern extent of the Subtropical Front and  
331 the southern boundary of the Antarctic Circumpolar Current, and between the Argentine and  
332 Agulhas basins. Core areas were  $717,396 \pm 244,662 \text{ km}^2$  (range: 370,000 to 1,597,500  $\text{km}^2$ ,  
333  $n = 24$ ; Fig. 1). Although the non-breeding range appeared to be more extensive in 2012  
334 than 2002, there was no significant effect of year, sex or breeding status on distribution  
335 according to the results of the ANOSIM ( $R = 0.0 - 0.1$ ,  $P > 0.05$ ), nor did these factors affect  
336 the size of individual 95 % and 40 % UD's (Table 1).

337 The best model predicting skua distribution from oceanographic variables during the non-  
338 breeding period achieved an accuracy of  $\text{AUC} = 0.76$  (reasonable discrimination ability).  
339 However, the predictive accuracy when calculated separately for each individual varied from  
340  $\text{AUC} = 0.50$  (no discrimination) to  $\text{AUC} = 0.91$  (very good discrimination). The most  
341 important predictors of habitat use by the tracked skuas were, in order of decreasing  
342 importance, depth and depth slope. Model response curves indicated that probability of  
343 occurrence was higher in both shallow and deep water (i.e., lowest in mid-depths), and in  
344 areas with steeper depth slopes (Fig. 2). There was no effect of year on habitat preference  
345 during the non-breeding period, nor did the tracked birds prefer areas according to SST,  
346 SST gradient, EKE or SLA.

347

### 348 **Pre-laying exodus**

349 After the return to South Georgia, two out of three (67 %) and 10 out of 19 (53 %) brown  
350 skuas that returned to breed in 2002 and 2012, respectively, engaged in a pre-laying  
351 exodus. No characteristic of the pre-laying exodus (start and return dates, duration, distance  
352 travelled per day and maximum range from the colony) differed significantly between males  
353 and females, nor between birds that bred successfully or failed in the previous season  
354 (Table 1). However, the majority (75 %) of pre-laying trips were performed by females.  
355 Skuas departed to sea on 30 October  $\pm 9$  days (range: 18 October to 15 November,  $n = 12$ ),  
356 which was  $16 \pm 8$  days (range: 8 to 39 days,  $n = 12$ ) after their first return to the colony, for a  
357 pre-laying exodus that lasted six to nine days ( $7 \pm 1$  days,  $n = 12$ ). All trips were to the north  
358 of South Georgia, ranging from 772 to 2,636 km from the colony ( $1,553 \pm 580 \text{ km}$ ,  $n = 12$ ).  
359 Trips were to subantarctic and mixed subantarctic-subtropical waters (Fig. 1). The tracked  
360 birds covered  $455 \pm 212 \text{ km}$  per day (range: 195.5 to 953 km per day,  $n = 12$ ). Individual  
361 core areas and 95 % UD's during the pre-laying exodus ranged from 195,000 to 402,500  $\text{km}^2$   
362 ( $256,250 \pm 61,158 \text{ km}^2$ ,  $n = 12$ ) and from 1,105,000 to 2,360,000  $\text{km}^2$  ( $1,538,333 \pm 345,694$   
363  $\text{km}^2$ ,  $n = 12$ ), respectively. There were no significant effects of sex or breeding status on

364 distributions (ANOSIM results,  $R = -0.1 - 0.2$ ,  $P > 0.05$ ), nor on the size of individual core  
365 and general use areas according to linear models (Table 1).

366 Depth was the main predictor of skua distribution during the pre-laying period. The best  
367 model had reasonable discrimination ability (AUC = 0.77). Variation among individuals,  
368 however, was considerable, with predictive accuracies ranging from none (AUC = 0.50) to  
369 very high (AUC = 0.93). Probability of occurrence of brown skuas increased in areas with  
370 greater water depth (Fig. 3). Habitat selection by tracked skuas was not affected by year,  
371 SST, SST gradient, EKE, SLA, and depth slope.

372

### 373 **At-sea activity patterns**

374 Given the few birds sampled in 2002, immersion data from both years were pooled. Brown  
375 skuas from South Georgia spent a considerably smaller percentage of time in flight than on  
376 the water during both the non-breeding and pre-laying exodus periods (Tables 2 and 3).  
377 During the non-breeding period, females spent more time than males in flight during  
378 darkness, probably as a consequence of longer, but not more frequent, flight bouts (Tables 2  
379 and 3). Skuas were more active (i.e., performed more flight bouts), during daylight than  
380 darkness; however, the bouts in daylight were shorter. During the pre-laying exodus,  
381 although the proportion of time spent in flight appeared to be higher in daylight than  
382 darkness, this was not reflected in the best model, nor in this case, was there a significant  
383 difference between males and females (Tables 2 and 3). Flight bout duration, however, was  
384 longer during darkness, and females engaged in more flight bouts per day than males. There  
385 was no difference in the number of flight bouts per day between daylight and darkness, nor  
386 in the duration of flight bouts between males and females during the pre-laying exodus  
387 (Tables 2 and 3).

388

## 389 **DISCUSSION**

### 390 **Distribution and habitat use**

391 With the inclusion of the new, large sample from 2012, this study represents the most  
392 detailed published analysis of movements and activity patterns of brown skuas during the  
393 non-breeding and pre-laying exodus periods. Although habitat use of brown skuas was  
394 described previously in terms of differences in bathymetry, sea surface temperature and  
395 chlorophyll *a* concentrations between core and peripheral areas (Phillips et al. 2007), this is

396 the first study to present a robust model of habitat preference, accounting for availability. The  
397 loggers deployed in summer 2011/12 had no deleterious effect on adult survival; indeed,  
398 return rates were higher, but not statistically significant, for birds with devices than for control  
399 birds. Although some birds in the larger sample tracked in winter 2012 travelled to areas  
400 east of those used in winter 2002, there were no statistically significant differences in  
401 distribution, the size of the core and general use areas, nor in habitat preferences, despite  
402 the decade that had elapsed between the two series of deployments. Similarly, northern  
403 hemisphere skuas (great *Stercorarius skua* and long-tailed *S. longicaudus* skuas) tracked  
404 during two or three consecutive years migrated to the same general regions, and had similar  
405 migration schedules (Sittler et al. 2011, Magnúsdóttir et al. 2012, Gilg et al. 2013).

406 The brown skuas tracked in this study were not limited to any specific water mass; however,  
407 most of their core areas were in subantarctic and mixed subantarctic subtropical waters  
408 during both the non-breeding season and the pre-laying exodus. They preferred regions with  
409 either shallow or deep waters, and increasing depth slopes, during the migration period,  
410 whereas in the pre-laying exodus, they selected habitat based solely on water depth. This  
411 confirms the inference from stable isotope analysis that a minority of brown skuas winter in  
412 shelf-slope waters (Phillips et al. 2007, 2009). Depth gradients are steeper at seamounts,  
413 and where shelf-edge fronts form, which are zones of intense mixing and enhanced primary  
414 production (Bost et al. 2009, Louzao et al. 2011b, Wakefield et al. 2011, Scales et al. 2014).  
415 Also, several studies have shown associations of seabirds with frontal systems influenced by  
416 bathymetric features (Phillips et al. 2005, 2006, Paiva et al. 2010a, Paiva et al. 2010b,  
417 Louzao et al. 2011b, Rayner et al. 2012).

418 In the present study, the core areas of several individuals (as well as the composite 40 %  
419 UDs for the population) overlapped extensively with regions of steeper slopes around the  
420 Falkland Escarpment, which forms the boundary between the Falkland Plateau and the  
421 Argentine Basin (Fig. 1). The Falkland Escarpment is a region subject to high frontal  
422 probability during the austral autumn, winter and spring, mostly associated with the presence  
423 of eddies from the Brazil-Malvinas Confluence, where the warm Brazil Current and the cold  
424 Malvinas Current meet (Saraceno et al. 2004). Moreover, five birds targeted areas of  
425 shallower bathymetry around the Discovery Rise and its associated seamounts during their  
426 migration in 2012, whilst one bird travelled as far as the Agulhas Basin, reaching turbulence  
427 zones from the Agulhas Return Current, where primary productivity is enhanced.

428 Although the skuas showed a preference for oceanographic characteristics that are often  
429 found in areas with abundant and predictable prey, it is noteworthy that much of the  
430 Argentine Basin, which was one of the main wintering areas, is not particularly productive

431 based on chlorophyll a concentrations (Phillips et al. 2007). This suggests that bathymetric  
432 features alone cannot entirely explain the observed distribution of brown skuas. Quillfeldt et  
433 al. (2013) suggested that the occurrence of Antarctic prions *Pachyptila desolata* in deep  
434 waters to the east of the Patagonian Shelf may be to avoid competition with larger seabirds.  
435 The same may be true for brown skuas, as the Patagonian Shelf is used year-round by  
436 numerous albatross and petrel populations from Tristan da Cunha, Gough, the Falkland  
437 Islands, South Georgia, and New Zealand (Croxall & Wood 2002, Phillips et al. 2006, Catry  
438 et al. 2013b, Copello et al. 2013). The ranges of the tracked brown skuas still overlapped to  
439 some extent with non-breeding Falkland skuas *S. antarcticus antarcticus* (Phillips et al.  
440 2007), and even more with wandering albatrosses *Diomedea exulans*, year-round (Jiménez  
441 et al. in press). However, inter-specific competition between these taxa may be alleviated by  
442 trophic partitioning of resources or differences in behaviour. Whilst  $\delta^{13}\text{C}$  values are similar  
443 for wandering albatrosses and brown skuas,  $\delta^{15}\text{N}$  values in feathers from wandering  
444 albatrosses are much higher, and their activity patterns differ (Phalan et al. 2007, Phillips et  
445 al. 2009, Ceia et al. 2012). Therefore, although the tracking and isotope data indicate that  
446 they occur in the same general area, wandering albatrosses feed at substantially higher  
447 trophic levels. Although  $\delta^{15}\text{N}$  values in brown skuas are significantly higher than in non-  
448 breeding Antarctic prions, this is not by a full trophic level, suggesting there is some diet  
449 overlap (Phillips et al. 2007, 2009). Antarctic prions feed on a wider range of small  
450 zooplankton taxa, whereas brown skuas probably have a mixed diet that includes  
451 zooplankton, low trophic level squid or fish (Phillips et al. 2007, Grecian et al. 2015). Finally,  
452 skuas breeding at the Falkland Islands occur closer to the Patagonian shelf-break and to a  
453 lesser extent in open waters than brown skuas (Phillips et al. 2007).

454 As brown skuas select habitats based mainly on static oceanography (water depth), their  
455 return to the same general area each year is not surprising, assuming that prey preferences  
456 are also consistent. Similarly, the use of neritic waters and areas of steeper bathymetric  
457 relief by black-browed albatrosses varied very little over multiple breeding seasons (Pinaud  
458 & Weimerskirch 2005, Wakefield et al. 2011, Catry et al. 2013b). The recurrent use of similar  
459 areas may increase familiarity with feeding conditions, including fine-scale resource  
460 distribution, and potentially help reduce inter-specific competition (Quillfeldt et al. 2013,  
461 Ramírez et al. 2015). Although habitat preferences were consistent between years, the  
462 tracked skuas had multiple migration destinations in 2012, which should ensure that a  
463 proportion of the population would escape from detrimental changes during winter if those  
464 are constrained spatially (Phillips et al. 2009, Dias et al. 2011). At the population level, core  
465 and general use areas did not differ significantly between years, but a few individuals in  
466 2012 travelled to regions east of those exploited in 2002 (Fig. 1). Two birds in this latter

467 group remained exclusively in the Agulhas Basin or in waters around the Discovery Rise,  
468 whereas another four birds used areas around the Discovery Rise and in the Argentine  
469 Basin. The existence of several non-breeding destinations for the same population has been  
470 confirmed for many seabird species, including south polar and great skuas (Phillips et al.  
471 2005, Kopp et al. 2011, Dias et al. 2011, Magnúsdóttir et al. 2012, Tranquilla et al. 2013,  
472 Ramírez et al. 2015). That the minority strategies were not detected in brown skuas from  
473 South Georgia in the first set of deployments probably reflects the modest sample size rather  
474 than indicates that birds have explored new environments in the subsequent decade. This is  
475 supported by the stable isotope data which suggests that some birds sampled in 2001/02  
476 had exploited neritic or shelf-slope waters in the previous winter (see above), and by the lack  
477 of any substantial differences between years in distribution based on randomizations, or in  
478 habitat preferences (this study). Regardless, the two tracking datasets provide an adequate  
479 representation of the distribution of the majority of individuals. Indeed, in an analysis of 15  
480 years of tracking data, Bogdanova et al. (2014) suggested that one or two years of data is  
481 usually sufficient to identify a considerable proportion of the long-term foraging areas for  
482 species which feed on spatially stable habitats. Furthermore, the distributions of general (95  
483 % UD) and core areas (40 %) of the tracked skuas were not influenced by sex or breeding  
484 status, which are complicating factors in other studies (Bogdanova et al. 2011, 2014).

485

#### 486 **Migration schedules**

487 Failed birds departed from the breeding grounds earlier than successful birds, in line with  
488 other studies (Phillips et al. 2005, Bogdanova et al. 2011, Quillfeldt et al. 2015). If migration  
489 commences earlier, more time may be allocated to undertake farther movements. Indeed,  
490 maximum distance from the colony was higher for failed than for successful breeders. Failed  
491 birds may also be in better condition to undertake longer migrations, as costs associated  
492 with breeding have been lower (Bogdanova et al. 2011). The latter is particularly likely for  
493 skuas at South Georgia, which are able to exploit abundant seal carrion only in December  
494 and January; thereafter, the availability of carrion declines steeply and breeding birds are  
495 forced to target other, less profitable prey (Phillips et al. 2004a, Anderson et al. 2009,  
496 Carneiro et al. 2014). Therefore, a movement away from the breeding grounds to areas with  
497 more favourable conditions should be expected for failed birds that are no longer under  
498 central-place constraints.

499 The mean date of return to South Georgia, however, was similar for all tracked individuals  
500 within the same year, suggesting that previous breeding outcome ceased to be important at

501 some stage during the migration period (Phillips et al. 2005). Similar arrival dates may  
502 decrease the chances of nest site usurpation by prospectors seeking to adopt a vacant site  
503 (Furness 1987). Indeed, it has been suggested that nesting habitat in the study area is  
504 saturated, with few opportunities for the establishment of new territories (Phillips et al.  
505 2004a). The explanation for the later return of tracked birds to the colony in 2012 than 2002  
506 is less clear. Shifts in phenology are expected as a response to climate change, and in the  
507 Southern Ocean, dates of arrival and of the first egg have been later for a number of species  
508 (Barbraud & Weimerskirch 2006). In summer 2002/03, brown skuas at South Georgia laid  
509 later, and had lower hatching success, fledging success and chick growth rates than in the  
510 mid-1980s (Phillips et al. 2004a). Later arrival in 2012 than 2002 may therefore be indicative  
511 of poorer environmental conditions during the preceding winter, potentially reflecting a  
512 longer-term trend, but this would need to be confirmed by more extensive tracking. There  
513 was no consistent effect of sex on timing of movements, in accordance with previous studies  
514 of arrival times of brown skuas at the South Orkney Islands, and of migration schedules of  
515 long-tailed skuas in the high Arctic (Burton 1968, Gilg et al. 2013).

516 Prior to egg-laying and after returning from migration, ca. 55 % of tracked skuas travelled to  
517 waters to the north of South Georgia in a pre-laying exodus. Although timings and distances  
518 travelled were unaffected by sex and previous breeding status, a pre-laying exodus was  
519 more commonly performed by females. At this time of year, food resources accessible on  
520 land for skuas are likely to be limited. There is very little carrion available from seals, and  
521 most burrow-nesting petrels have yet to return (Carneiro et al. 2014). Skuas arrive earlier at  
522 their breeding grounds to defend their territories and (re)establish pair bonds, but it is  
523 possible that to acquire enough resources for egg formation, females have to return to feed  
524 at sea. In most seabird species, females are generally away for longer periods or travel  
525 further to reach more productive waters (Guilford et al. 2009, Hedd et al. 2014, Quillfeldt et  
526 al. 2014, but see Pinet et al. 2012). Moreover, given the exceptionally high nesting density at  
527 South Georgia, loss of the territory to conspecifics is more likely if the breeding site is  
528 unattended for longer periods, suggesting it may be beneficial for one member of the pair to  
529 remain close to the colony. Indeed, high attendance rates and fewer joint nest absences  
530 have been reported for skuas at South Georgia during incubation and chick-rearing when  
531 compared to sites elsewhere (Pietz 1987, Catry & Furness 1999, Caldow & Furness 2000,  
532 Carneiro et al. 2014).

533

#### 534 **At-sea activity patterns**



535 Brown skuas spend much less time flying during the non-breeding and pre-laying periods  
536 than albatrosses and petrels, particularly during daylight (Phillips et al. 2007, this study). The  
537 number of flight bouts per day was surprisingly low. This suggests that birds may forage  
538 opportunistically using a sit-and-wait strategy as previously described for other species,  
539 including white-chinned petrels *Procellaria aequinoctialis*, grey-headed albatrosses  
540 *Thalassarche chrysostoma*, and Desertas petrels *Pterodroma deserta* (Catry et al. 2004,  
541 Mackley et al. 2010, Péron et al. 2010, Ramírez et al. 2013). Alternatively, brown skuas may  
542 feed on large items of carrion or moribund prey, as they do on land during the breeding  
543 season (Phillips et al. 2004a, Anderson et al. 2009, Carneiro et al. 2014). Amongst  
544 procellariids, the species with activity patterns during the non-breeding season that are most  
545 similar (i.e., low flight activity) to those of wintering brown skuas are wandering albatrosses  
546 and white-chinned petrels, which to some extent also rely on scavenging (Xavier et al. 2004,  
547 Mackley et al. 2010, 2011). These two species nonetheless spend considerably more time in  
548 flight, possibly because they use dynamic soaring to reduce flight costs, whereas skuas  
549 usually flap continuously in cruising flight and so would expend more energy if travelling long  
550 distances between prey patches (Pennycuik 1987, Catry et al. 2011, Gutowsky et al. 2014).

551 Although both sexes were more active during daylight, females spent proportionally more  
552 time flying during darkness than males in the non-breeding season. This reflects their longer,  
553 but not more frequent, flight bouts. As females have higher wing loading and less  
554 manoeuvrability than males (Phillips et al. 2002, Carneiro et al. 2014), they might land less  
555 frequently because of the greater cost of taking off (Shaffer et al. 2001, Phillips et al. 2004c).  
556 The greater manoeuvrability of males would also be an advantage during daylight when  
557 feeding by kleptoparasitism (Phillips et al. 2002), but the limited time spent flying suggests  
558 that either this foraging mode is rare, or that when used, it is highly successful. Although in  
559 theory, darkness should tend to limit movements, as aerial detection and capture of prey is  
560 more difficult (Phalan et al. 2007, Mackley et al. 2011), this seems not to affect females to  
561 the same degree as males, but it could explain their longer flight durations at night. The  
562 number of flight bouts per day during the pre-laying exodus was also higher for females,  
563 suggesting that they increase their foraging effort to acquire enough resources for egg  
564 formation.

565

## 566 **CONCLUSIONS**

567 Although seabird distributions are expected to vary temporally and spatially according to the  
568 dynamic nature of the marine environment, the population of brown skuas tracked from

569 South Georgia during migration and the pre-laying exodus visited the same general area and  
570 showed similar habitat preferences between 2002 and 2012. They preferred habitats  
571 associated with static features (i.e., bathymetry), which may explain the consistency found  
572 between years. Further studies comparing years of contrasting foraging conditions would  
573 reveal whether skuas are able to change their foraging strategies to compensate for  
574 environmental changes. In addition, this study highlighted the need for large sample sizes to  
575 detect strategies used by a small proportion of the population. The existence of multiple non-  
576 breeding destinations may be advantageous given increasing anthropogenic impacts on  
577 marine ecosystems.

578

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586

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Table 1: Summary of linear models for the effects of sex, breeding status (successful or failed) and year on characteristics of the non-breeding period and pre-laying exodus of brown skuas tracked from South Georgia in winter 2002 and 2012. + indicates predictors that were retained by the most parsimonious models, n/a (not applicable) indicates predictors that were not retained in the analysis. Interactions are not shown because they were not retained in the best models.

	Sex	Status	Year	d.f.	AICc	$\Delta$ AICc
<u>Non-breeding period</u>						
Departure dates	+	+		4	193.6	0.0
Return dates			+	3	169.2	0.0
Duration	+	+		4	175.3	0.0
Size of core area (40% UD)				2	667.2	0.0
Size of general use area (95% UD)				2	753.3	0.0
Distance travelled per day				2	-11.7	0.0
Max. distance from colony		+		3	10.8	0.0
<u>Pre-laying exodus</u>						
Departure dates			n/a	2	-46.6	0.0
Return dates			n/a	2	90.6	0.0
Duration			n/a	2	38.0	0.0
Size of core area (40% UD)			n/a	2	1.6	0.0
Size of general use area (95% UD)			n/a	2	1.3	0.0
Distance travelled per day			n/a	2	19.1	0.0
Max. distance from colony			n/a	2	85.5	0.0

Table 2: Summary of activity patterns of brown skuas tracked from South Georgia during the non-breeding period and pre-laying exodus in winter 2002 and 2012. Values are presented as mean  $\pm$  SD (range).

	Daylight		Darkness	
	Male	Female	Male	Female
<u>Non-breeding period</u>				
% time in flight				
2002	25.9 $\pm$ 0.0 (25.9 - 25.9)	19.8 $\pm$ 6.9 (14.9 - 24.7)	11.4 $\pm$ 0.0 (11.4 - 11.4)	15.1 $\pm$ 1.4 (14.1 - 16.1)
2012	24.8 $\pm$ 4.5 (18.8 - 33.0)	25.9 $\pm$ 5.4 (17.9 - 37.5)	16.9 $\pm$ 3.9 (12.1 - 22.7)	24.5 $\pm$ 7.9 (9.2 - 42.0)
Total time in flight (h per day)				
2002	3.0 $\pm$ 0.0 (3.0 - 3.0)	2.3 $\pm$ 0.8 (1.7 - 2.9)	1.4 $\pm$ 0.0 (1.4 - 1.4)	1.9 $\pm$ 0.2 (1.8 - 2.1)
2012	2.6 $\pm$ 0.5 (2.0 - 3.5)	2.9 $\pm$ 0.7 (1.9 - 4.1)	2.2 $\pm$ 0.5 (1.6 - 3.1)	3.2 $\pm$ 1.1 (1.2 - 5.6)
Total time on water (h per day)				
2002	8.5 $\pm$ 0.0 (8.5 - 8.5)	9.3 $\pm$ 0.8 (8.7 - 9.9)	11 $\pm$ 0.0 (11.0 - 11.0)	10.5 $\pm$ 0.2 (10.4 - 10.6)
2012	8.0 $\pm$ 0.5 (7.2 - 8.6)	8.3 $\pm$ 0.8 (6.9 - 9.5)	11.2 $\pm$ 0.5 (10.6 - 11.7)	9.7 $\pm$ 1.2 (7.7 - 11.7)
No. flight bouts per day				
2002	2.3 $\pm$ 0.0 (2.3 - 2.3)	2.0 $\pm$ 1.0 (1.3 - 2.7)	0.9 $\pm$ 0.0 (0.9 - 0.9)	1.3 $\pm$ 0.1 (1.2 - 1.4)
2012	1.8 $\pm$ 0.8 (0.9 - 2.7)	2.2 $\pm$ 0.7 (1.0 - 3.7)	1.3 $\pm$ 0.3 (1.0 - 1.9)	1.4 $\pm$ 0.4 (0.9 - 2.0)
Duration flight bouts (min)				
2002	29.1 $\pm$ 0.0 (29.1 - 29.1)	32.9 $\pm$ 4.4 (29.8 - 36.0)	85.9 $\pm$ 0.0 (85.9 - 85.9)	89.7 $\pm$ 1.5 (88.7 - 90.8)
2012	30.8 $\pm$ 2.2 (28.6 - 35.1)	46.1 $\pm$ 29.3 (33.4 - 135.7)	95.6 $\pm$ 15.4 (81.6 - 119.3)	119.7 $\pm$ 43.7 (57.6 - 188.9)
<u>Pre-laying exodus</u>				
% time in flight				
2002	23.4 $\pm$ 0.0 (23.4 - 23.4)	11.5 $\pm$ 0.0 (11.5 - 11.5)	1.4 $\pm$ 0.0 (1.4 - 1.4)	0.3 $\pm$ 0.0 (0.3 - 0.3)
2012	13.6 $\pm$ 3.3 (11.3 - 15.9)	19.1 $\pm$ 9.0 (12.6 - 40.0)	9.9 $\pm$ 6.1 (5.6 - 14.2)	17.8 $\pm$ 10.0 (5.2 - 33.0)
Total time in flight (h per day)				
2002	3.6 $\pm$ 0.0 (3.6 - 3.6)	1.7 $\pm$ 0.0 (1.7 - 1.7)	0.2 $\pm$ 0.0 (0.2 - 0.2)	0.0 $\pm$ 0.0 (0.0 - 0.0)
2012	2.0 $\pm$ 0.4 (1.7 - 2.3)	2.6 $\pm$ 1.4 (1.1 - 5.6)	0.9 $\pm$ 0.6 (0.5 - 1.3)	1.9 $\pm$ 1.2 (0.4 - 4.0)
Total time on water (h per day)				
2002	12.0 $\pm$ 0.0 (12.0 - 12.0)	13.6 $\pm$ 0.0 (13.6 - 13.6)	8.1 $\pm$ 0.0 (8.1 - 8.1)	8.6 $\pm$ 0.0 (8.6 - 8.6)
2012	12.4 $\pm$ 0.7 (11.9 - 12.9)	10.6 $\pm$ 2.2 (7.9 - 13.5)	8.6 $\pm$ 0.8 (8.1 - 9.1)	8.8 $\pm$ 2.4 (6.1 - 13.0)

N. flight bouts per day					
	2002	2.0 ± 0.0 (2.0 - 2.0)	1.1 ± 0.0 (1.1 - 1.1)	0.5 ± 0.0 (0.5 - 0.5)	0.1 ± 0.0 (0.1 - 0.1)
	2012	0.5 ± 0.2 (0.3 - 0.7)	2.0 ± 1.9 (0.7 - 6.1)	0.8 ± 0.2 (0.7 - 1.0)	1.4 ± 1.0 (0.3 - 3.4)
Duration flight bouts (min)					
	2002	46.0 ± 0.0 (46.0 - 46.0)	34.4 ± 0.0 (34.4 - 34.4)	63.3 ± 0.0 (63.3 - 63.3)	28.2 ± 0.0 (28.2 - 28.2)
	2012	23.7 ± 2.7 (21.8 - 25.6)	40.5 ± 18.4 (26.1 - 77.5)	62.3 ± 6.8 (57.5 - 67.1)	67.5 ± 40.8 (24.9 - 147.2)

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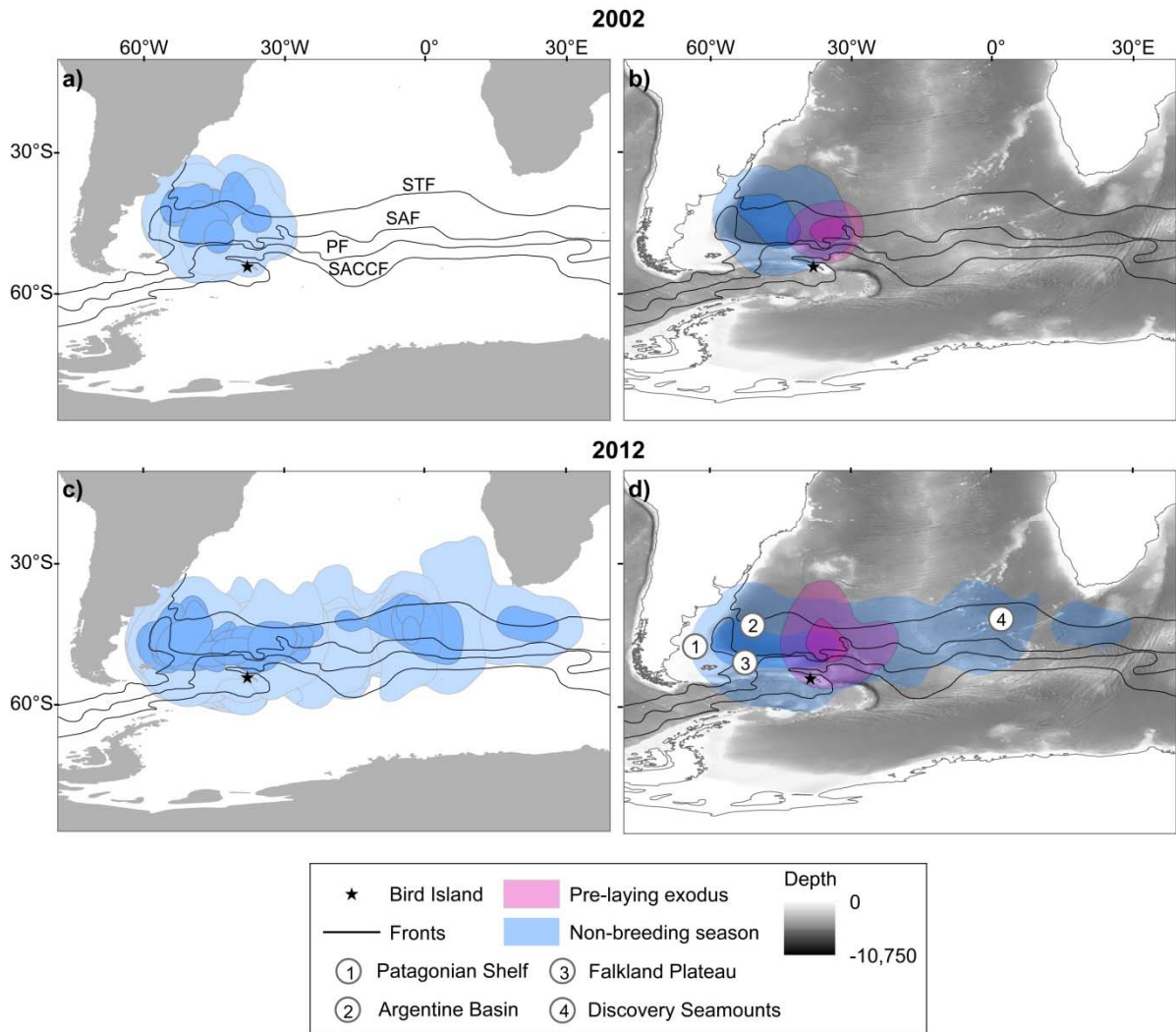
Table 3: Summary of linear mixed-effects models of activity patterns of brown skuas tracked from South Georgia in 2002 and 2012 during the non-breeding period and pre-laying exodus. + indicates predictors retained in the most parsimonious models.

<u>Non-breeding period</u>	LightDark	Sex	LightDark:Sex	d.f.	AICc	ΔAICc
% of time spent in flight	+	+	+	6	-1399.6	0.0
No. of flight bouts per day	+			3	-3.7	0.0
Duration of flight bouts	+	+		4	373.1	0.0
<hr/>						
<u>Pre-laying exodus</u>						
% of time spent in flight				3	-82.9	0.0
No. of flight bouts per day		+		3	50.0	0.0
Duration of flight bouts	+			3	32.0	0.0

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864 Fig. 1: Utilization distributions (40 % and 95 %) of individual brown skuas tracked from South  
 865 Georgia during the non-breeding season in 2002 [a] and 2012 [c]. Composite of individual  
 866 utilization distributions (40 % and 95 %) during non-breeding and pre-laying exodus in 2002  
 867 [b] and 2012 [d] overlaid on bathymetric map. Black lines correspond to STF (Subtropical  
 868 Front), SAF (Subantarctic Front), PF (Polar Front), and SACCF (Southern Antarctic  
 869 Circumpolar Current Front).  
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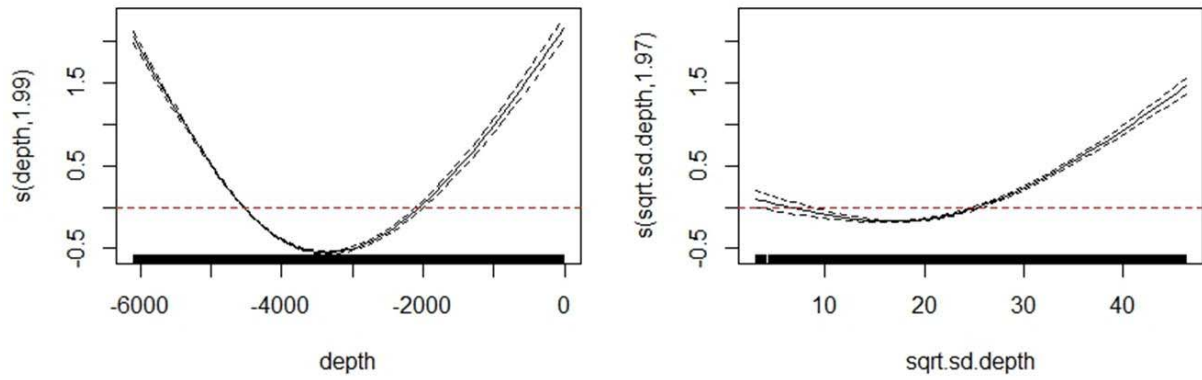
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876 Fig. 2: Response curves of predictors retained in the most parsimonious habitat model for  
877 brown skuas during the non-breeding period. Dashed lines indicate estimated 95 %  
878 confidence intervals, with covariate values as a rug along the bottom of the figure.



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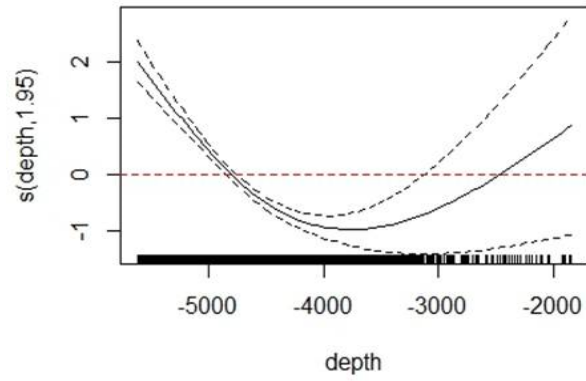
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893 Fig. 3: Response curves of predictors retained in the most parsimonious habitat model for  
894 brown skuas during the pre-laying exodus. Dashed lines indicate estimated 95 % confidence  
895 intervals, with covariate values as a rug along the bottom of the figure.



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## Supplement 1

Effects of different numbers of correlated random walks (CRWs) on model performance. Number of simulations was tested by running global models with all observed locations and varying numbers of simulated pseudo-absences. The number of simulations was determined by measuring changes in  $\chi^2$  values of each parameter, being set when  $\chi^2$  stabilized. SST refers to sea surface temperature, SLA to sea level anomaly, and EKE to eddy kinetic energy.

