

1 **Addressing tagging location bias to assess space use by marine**  
2 **animals**

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11 **Abstract:**

- 12 1. Estimates of space use derived from animal tracking studies are often biased by where  
13 animals are tagged, with areas distant to the tagging site, in both space and time, being  
14 under-represented.
- 15 2. We develop an approach to overcome this tagging bias by quantifying the likely movements  
16 of animals after tags have failed.
- 17 3. We illustrate the approach using high accuracy Fastloc-GPS tracking data for 35 adult female  
18 green turtles (*Chelonia mydas*) equipped with satellite tags within one of the world's largest  
19 marine protected areas (MPAs), the British Indian Ocean Territory MPA.
- 20 4. Individuals migrated up to 5127 km from the tagging site, breaking migration distance  
21 records for this species. For 28 of 35 individuals travelling to foraging locations well outside  
22 the MPA, we estimated that they spent, on average, 9.8% of their adult lives within the  
23 British Indian Ocean Territory MPA.
- 24 5. Synthesis and applications. The importance of the British Indian Ocean Territory MPA as a  
25 nesting sanctuary for individuals from across an ocean basin is highlighted. The general  
26 approach we outline can be applied to a broad range of taxa, including marine mammals,

27 fish and sea turtles and will allow unbiased estimates of how important areas, such as MPAs,  
28 are used.

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31 Keywords. Migration, Argos, space use, megafauna, telemetry, biologging, Chagos Archipelago

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33 Running head. Tagging location bias and space use

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## 36 **1. INTRODUCTION**

37 It is now commonplace to track animals to assess their patterns of movement and space use  
38 (Costa et al. 2012; Hussey et al. 2015; Kays et al. 2015), with tracking data often providing  
39 information that has translated into conservation management and policy (Hays et al. 2019). In  
40 many studies, however, space use estimates are influenced by the tagging locations. Put simply,  
41 space use estimates will often show tagging sites as high use areas simply because all of the  
42 tracks radiate from that location and not because animals necessarily spend lots of time at that  
43 tagging site. So, for example, if an animal had instead been tagged in another part of its range,  
44 then a different pattern of space use would be generated. Examples of where animals tend to  
45 be tagged in only part of their range are widespread across studies and include tagging animals  
46 at breeding sites, e.g. turtles, seabirds and pinnipeds, or for fish where they are captured in  
47 commercial fisheries (e.g. McMahon et al. 2008; Bailey et al. 2012). Some studies across marine  
48 taxa, including sharks, fish, cetaceans, pinnipeds, sea birds and turtles, have tried to correct for  
49 tagging location biases, with tracking locations being inversely weighted by when they were  
50 recorded in relation to the tagging date, so that locations obtained a long time after tagging are  
51 given more weight and vice versa (Block et al. 2011; Queiroz et al. 2019). This elegant approach  
52 partly ameliorates tagging location biases, but is not a perfect solution to the problem since  
53 animals may continue to move far away from the tagging locations well after the tags fail. Given  
54 the widespread issue with tagging location bias, here we conceptualize a new approach to

55 tackle this issue. Our approach will be broadly applicable across taxa, allowing unbiased  
56 estimates of how species use key areas such as Marine Protected Areas (MPAs). As a case  
57 study, we implement the approach using long-term satellite tracking data obtained from a  
58 species that can migrate 1000s of km across an ocean basin between breeding and foraging  
59 sites.

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## 61 **2. METHODS**

62 We first developed a conceptual framework for estimating space use in a way that is not biased  
63 by the release location of tagged animals. To develop this framework we used three examples  
64 from the literature of animals where empirical tracking results of individuals generally only  
65 cover part of an individual's overall movements. For illustrative purposes, we selected one fish,  
66 one marine mammal and one sea turtle to show the broad applicability of our framework. In  
67 each example, we also show how the overall pattern of movements may be broadly known  
68 using a range of data-sets that supplement the information provided by tracks of individuals.

69 We then develop this conceptual framework in a case study to estimate use of a large  
70 MPA by a long-distance ocean migrant. We use satellite tracking results for green sea turtles  
71 (*Chelonia mydas*) equipped while nesting on the island of Diego Garcia in the Chagos  
72 Archipelago, Indian Ocean (7.428° S, 72.458° E). This nesting area lies at the heart of one of the  
73 world's largest protected areas: the British Indian Ocean Territory (BIOT) MPA, that extends  
74 generally 200 nautical miles seaward from the outermost atolls to the limit of the UK territorial  
75 waters. During the nesting seasons in 2012, 2015, 2017 and 2018, female turtles were located  
76 while they were nesting ashore at night. Once turtles were returning to the sea they were  
77 restrained in a large open-topped and bottomless wooden box and a Fastloc-GPS Argos tag  
78 attached using quick setting epoxy (see Esteban et al., 2017 for details). In 2012, we used two  
79 models of satellite tag (SPLASH10-BF, Wildlife Computers, Seattle, Washington (n = 4) and  
80 model F4G 291A, Sirtrack, Havelock North, New Zealand (n = 4). In other years we only used  
81 SPLASH10-BF units (n= 10, 5 and 12 in 2015, 2017 and 2018). Transmitters relayed data via the  
82 Argos system (<http://www.argos-system.org/>) that allowed Fastloc-GPS positions to be

83 determined. Only Fastloc-GPS positions obtained with a minimum of four satellites and a  
84 residual error value of less than 35 were used, producing locations that were generally within a  
85 few tens of meters of the true location (Dujon et al. 2014).

86 To estimate the proportion of time adult female turtles spend within the BIOT MPA in a  
87 way that was free of bias introduced by where and when each animal was tagged or the  
88 behaviour of the animal, we supplemented the direct tracking data with other information  
89 about the scheduling of migration derived from other sources. The time an adult green turtle  
90 spends inside the BIOT MPA can be partitioned across: (i) time spent travelling from the edge of  
91 the MPA to and from the nesting beaches at the start and end of the breeding season  
92 respectively; (ii) time spent mating, which usually occurs close the nesting beaches; (iii) time  
93 spent laying several clutches of eggs. Each of these three periods can be estimated for both  
94 male and female turtles. We estimated “(i)” for male and female turtles using our tracking data  
95 for turtles leaving the MPA, assuming that the reciprocal journey from the boundary of the  
96 MPA to the nesting beaches took the same amount of time. (ii) The amount of time that  
97 females spend mating has been estimated at 30 days by using visual observations of green  
98 turtles at mating areas (Godley et al. 2002). Male turtles arrive to breed before females and  
99 their length of residency at the breeding grounds has been measured at 75 days (Schofield et al.  
100 2013). (iii) The length of time females spend laying clutches has been estimated from the  
101 attachment of satellite tags to females at the start of the nesting season (Esteban et al. 2017).  
102 Green turtles in the Chagos Archipelago lay a mean of 6 clutches with a modal inter-nesting  
103 interval of 13 days (Esteban et al. 2017), so that, on average, they spend a total of 65 days  
104 nesting.

105 The interval between breeding seasons (remigration interval) has been recorded for  
106 female green turtles in the Indian Ocean using numbered flipper tags that allow a turtle to be  
107 identified when they are encountered nesting across different years. At the island of Mayotte,  
108 an isolated green turtle rookery in the SW Indian Ocean, the modal remigration interval for  
109 females is 3 years (Bourjea et al. 2007). Given differences in their breeding biology, the likely  
110 corresponding modal remigration interval for adult male turtles is 2 years (Hays et al 2014a).

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## 113 **3. RESULTS**

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### 115 **3.1 Conceptual framework**

116 The issue of tagging site bias is illustrated schematically in Figure 1, which shows the movement  
117 patterns that may potentially be performed by animals before and after tags fail, i.e. recorded  
118 tracks may only cover a fraction of an animal's overall movements. To overcome tagging  
119 location biased estimates of space use, one approach is to augment individual tracking data  
120 with additional information on animal movement obtained by other means. For example, for  
121 some taxa, seasonal occurrence in certain locations may be known from visual direct  
122 observations (e.g. cetaceans); the scheduling of movement may be known from mark-recapture  
123 studies (e.g. seabirds, sea turtles, pinnipeds and some fish); different individuals may have been  
124 tracked in different parts of a species range. Blending these types of data, with the tracks of  
125 individuals tagged in focal areas, may help to complete the picture of an animal's likely  
126 movements while it was not being directly tracked.

127 The long-term patterns of movement of different taxa are shown schematically in Figure  
128 1. In each case, the overall scheduling of their movements may be broadly known, but only  
129 sections of their breeding journeys are recorded by direct tracking. For example, for southern  
130 right whales (*Eubalaena australis*), some individuals have been equipped with satellite tags in  
131 calving areas in southern Australia and New Zealand that, from photo-id records, they are  
132 known to return to seasonally, while their seasonal occurrence at foraging areas in the  
133 Southern Ocean has been recorded through whaling records (Mackay et al. 2020) (Figure 1a).  
134 Northern bluefin tuna (*Thunnus thynnus thynnus*) have been equipped with either satellite tags  
135 or archival loggers in different parts of their range, allowing their overall extent of movement to  
136 be pieced together (Block et al. 2005) (Figure 1b). For adult green turtles that nest on Ascension  
137 Island in the central Atlantic, both long-term fidelity to nesting beaches and breeding  
138 periodicity have been documented by mark-recapture studies (Mortimer and Carr 1987), while  
139 elements of their post-nesting migration routes have been recorded by satellite tracking (Luschi

140 et al. 1998) (Figure 1c). Viewed in these ways, direct tracking of individuals can be viewed as  
141 just one component of the information that can be used to assess their likely patterns of space  
142 use, including periods both before and after they were directly tracked.

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### 145 **3.2 Case Study: green turtles in the Indian Ocean**

146 For none of the tracked turtles was the round-trip migration from the nesting beaches to the  
147 breeding area and back to the nesting beaches recorded, i.e. tracks only covered part of the  
148 expected movements of individuals across a breeding cycle. This result is akin to those  
149 examples shown in Figure 1. For 33 of the 35 tags, individuals were tracked to their foraging  
150 grounds, as indicated by individuals traveling to localized, relatively shallow areas where they  
151 remained for several months before tags failed. In two cases tags failed while individuals were  
152 still travelling in the open ocean and had not reached their destination. Turtles travelled to a  
153 broad range of destinations after the nesting season, with foraging sites identified across the  
154 western Indian Ocean (Figure 2). For seven of the 35 tracked turtles (20% of all tracked  
155 individuals) the post-nesting migration was relatively short, with these individuals travelling  
156 around 100 km northwards to their foraging grounds on the Great Chagos Bank and hence  
157 these individuals did not leave the BIOT MPA. The other 28 tracked turtles travelled beyond the  
158 MPA boundary and for the 26 of these 28 individuals whose foraging area was identified, two  
159 travelled to the Maldives, 17 travelled to islands and submerged banks in parts of the  
160 Seychelles and Mascarene Plateau, three travelled to Somalia, two to Kenya, one to  
161 Madagascar and one to Mozambique (Figure 2a). The maximum migration distance was for an  
162 individual that travelled 5127 km to foraging grounds in southern Mozambique, but long  
163 migrations were widespread with 11 individuals travelling >3000 km to reach their foraging  
164 grounds and 20 travelling >2000 km. The earliest tag failure occurred 89 days after deployment  
165 and the longest lasting tag functioned for 554 days. After 230 days, 50% of the tags were still  
166 functioning.

167 For turtles that travelled beyond the boundary of the MPA, the mean time to travel  
168 from Diego Garcia to the limit of the MPA was 6.0 days (n=28, range = 3.8-9.6 days, SD = 1.5  
169 days). So for female green turtles migrating to foraging grounds outside the BIOT MPA, over a  
170 3-year (1095 days) breeding cycle they would be expected to spend on average, (i) 12 days  
171 travelling from the edge of the MPA to and from the nesting beaches, (ii) 30 days mating, (iii) 65  
172 days nesting. So the mean proportion of time spent in the MPA is  $107\text{d}/1095\text{d} = 0.098$  or 9.8%  
173 (Figure 2b). The corresponding values for male turtles leaving the MPA, is that over a 2-year  
174 breeding cycle (730 days) they spend, (i) 12 days travelling from the edge of the MPA to and  
175 from the nesting beaches, (ii) 75 days at the breeding grounds and so  $87\text{d}/730\text{d} = 11.9\%$  of their  
176 time inside the MPA per 2-year breeding cycle.

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#### 179 **4. DISCUSSION**

180 The extent of tagging bias varies across taxa. For some groups this issue may be less of a  
181 problem. For example, for some birds small light-based geolocator tags as well as satellite tags  
182 have allowed annual roundtrip migrations to be recorded for many species in both marine and  
183 terrestrial species (e.g. Shaffer et al. 2006; Clay et al. 2017; Vardanis et al. 2016); while in some  
184 studies with fish, individuals have been tracked across multiple years (Lea et al. 2015), so that  
185 movement networks can be estimated to define space use (Jacoby et al. 2020). However in  
186 many studies, across multiple taxa, tagging bias issues will be important because tags fail before  
187 the full extent of an individual's movement has been captured. For example, often fish  
188 including sharks, rays and bony fish, are still travelling to new areas when tags fail (Queiroz et  
189 al. 2016; Sousa et al. 2016); keeping tags attached to cetaceans for long periods remains  
190 challenging (Fossette et al. 2014); pinnipeds shed external tags when they molt so that multi-  
191 year tracks are very difficult to obtain (McMahon et al. 2008); sea turtles generally only breed  
192 every few years so it is very difficult to track individuals throughout a complete breeding-  
193 foraging-breeding cycle (Hays & Hawkes 2018). For all of these taxa the challenge of our  
194 approach is to piece together where animals are most likely moving after tags fail. With sea

195 turtles, we have shown how the likely pattern of movement can be estimated by combining  
196 direct tracking with data derived from other techniques. Likewise for other taxa, including some  
197 fish and marine mammals, this same conceptual approach may be applicable, piecing together  
198 the likely overall pattern of movement from direct tracking used in conjunction with other  
199 sources of information that shed light on the extent and scheduling of movements.

200 In some cases, such as with pelagic fish, tagging individuals in different parts of their  
201 range as well as researchers sharing data from across tracking studies, offers great promise for  
202 assessing the overall pattern of species movements (Queiroz et al. 2019). For example, ocean  
203 sunfish (*Mola mola*) in the NE Atlantic have been satellite tagged at their more northerly  
204 summer feeding grounds off Ireland as well as more southerly feeding locations off Portugal  
205 (Sims et al. 2009, Sousa et al. 2016). So while individual tracks do not record the full extent of  
206 their north-south seasonal movements, these seasonal movements can be pieced together  
207 from the range of tracks obtained. For some taxa, tracking data can be supplemented with  
208 observations that show locations of seasonal residence, such as sightings data for many  
209 cetaceans or catch data for many fish. The likely movement of some taxa might also sometimes  
210 be assessed if the drivers of their movement patterns can be accurately described. For example,  
211 for wide-ranging oceanic foragers, movement models can now include a thermal constraint on  
212 movement (e.g. sea surface temperature for marine animals including seabirds) which may  
213 underpin seasonal north-south migrations, as well as the likelihood of animals stopping to feed  
214 in prey-rich oceanic patches (Lalire & Gaspar 2019, Pinaud et al. 2005). Empirical tracking data  
215 will allow such movement models to be better parameterized allowing, for example, thermal  
216 constraints on seasonal poleward movements to be better quantified (e.g. McMahon & Hays  
217 2006) as well as the probability of straight-line travel versus localized foraging (Bailey et al.  
218 2012; Humphries et al. 2012). In this way, if the pattern of movement can be accurately  
219 modelled, then individual space use can be projected across many years (Lalire & Gaspar 2019).

220 Our approach for estimating how green turtles use a large MPA may have broad  
221 conservation relevance. While the exact values for time spent in the BIOT MPA will not apply to  
222 other MPAs, the approach we used will still be applicable, i.e. combining direct tracking with  
223 other sources of information to estimate how much time individuals spend in protected areas.



224 Given that many 1000s of sea turtles have been satellite tracked around the world (Hays &  
225 Hawkes, 2018) and often there is good information on breeding intervals derived from mark-  
226 recapture tagging, our approach will have very broad utility across sea turtle species and  
227 populations. With regards to the BIOT MPA, since this protected area was declared in 2010 by  
228 the UK Government, there has been clear evidence of the conservation benefits for a range of  
229 taxa (D'agata et al. 2016). Nevertheless, the legality of the MPA has been challenged (e.g.  
230 Appleby 2015 but also recent media articles such as [https://www.bbc.com/news/uk-  
231 48371388](https://www.bbc.com/news/uk-48371388)). Our results greatly extend previous preliminary observations from seven tracked  
232 green turtles (Hays et al 2014b). We show that most of breeding female green turtles spend the  
233 majority of their adult lives outside the BIOT MPA. In other words, the MPA provides a nesting  
234 sanctuary for turtles that spend most of their adult lives at foraging sites across the entire  
235 western Indian Ocean. This nesting sanctuary is important, as turtles are particularly susceptible  
236 to poaching when ashore nesting, as are their eggs in nests. Therefore, the protection of turtles  
237 and their nests in the BIOT MPA has likely been a major contributor to the large increases in  
238 nesting numbers observed inside the MPA in recent years (Mortimer et al. 2020). Further, the  
239 pan-oceanic migrations of green turtles highlight the value of the conservation efforts that are  
240 being implemented across very broad spatial scales such as the western Indian Ocean (e.g., .  
241 Likewise, the approach we outline could allow unbiased estimates of how a range of marine  
242 mammals and fish use important areas such as MPAs, territorial waters of specific countries or  
243 important fishing zones. In summary, we suggest that by using tracking data synergistically with  
244 other information about the pattern and scheduling of animal movement, space-use estimates  
245 that are free of tag bias impacts may be made. Where this approach can be implemented it will  
246 allow improved estimates of how animals use protected areas and are exposed to various  
247 threats such as fishing activities.

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256

#### 257 **AUTHORS' CONTRIBUTIONS**

258 G.C.H. conceived the study and led the writing. G.C.H. and N.E. conducted the fieldwork, G.C.H.  
259 and A.R. analysed the data. All authors contributed to manuscript editing.

260

#### 261 **DATA ACCESSIBILITY**

262 All the data used in the calculations to avoid tagging bias issues appear within the manuscript.

263

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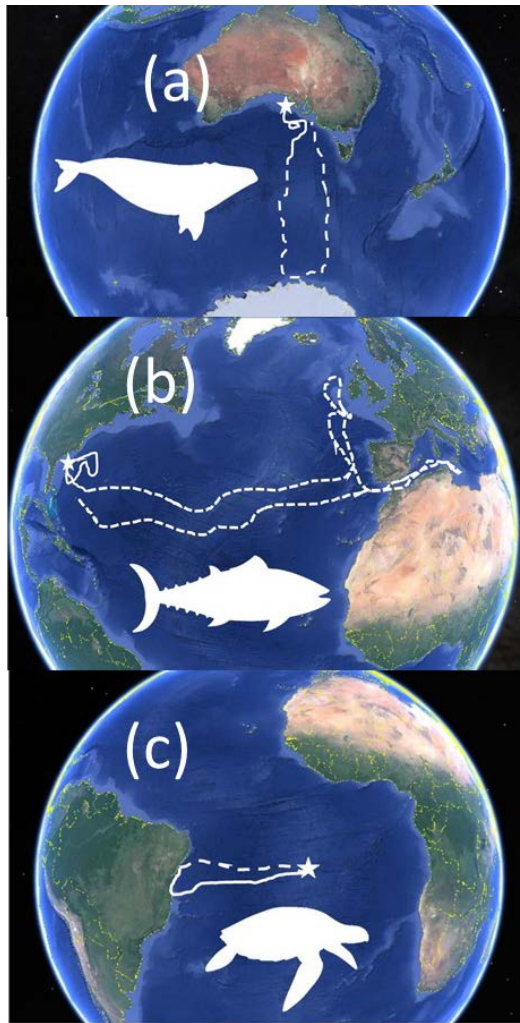


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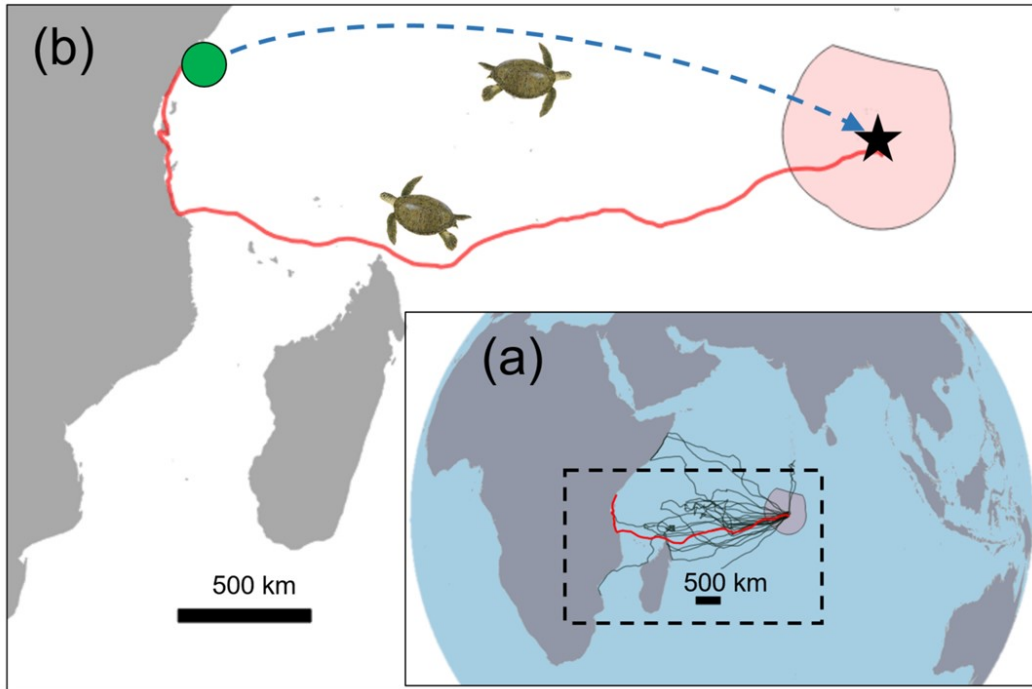
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FIGURE 1. Schematic representation of how tagging bias may impact estimates of space use in tracking studies. In each case the star represents an assumed tagging site, the solid line the track recorded by a tag and dashed line the possible animal movements after the tag has failed or detached. Icons represent example taxa. Overall patterns of movement, i.e. encompassing the time before, during and after individuals were directly tracked, may be informed by known patterns of seasonal occurrence (e.g. derived from visual observations), mark-recapture studies or by equipping individuals with tracking tags in different parts of a species range. (a) bluefin tuna in the North Atlantic, (b) green turtles nesting on Ascension Island, (c) southern right whales calving off southern Australia. See text for further details.



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433 FIGURE 2. Case study of estimated space use by migrating green turtles. (a) The tracks of 35  
 434 green turtles equipped with satellite tags while they were ashore nesting on Diego Garcia in the  
 435 Chagos Archipelago. The extent of the BIOT MPA is indicated by red shading. Of the 35 tracks,  
 436 seven travelled to foraging areas inside the MPA on the Great Chagos Bank, 28 travelled outside  
 437 the MPA of which 26 individuals were tracked all the way to their foraging area. (b) Breeding  
 438 migration cycle of a female green turtle tracked from a nesting beach on Diego Garcia (black  
 439 star) to foraging grounds on the coast of Kenya (green circle) from August to November 2015.  
 440 Fastloc GPS track (solid red line) shows the 105 day, 4835 km post-nesting migration, and the 3-  
 441 yearly (1095 days) return migration inferred from flipper tagging studies is represented by a  
 442 dashed blue line. Female green turtles spent on average 9.8% (107 days) of the breeding  
 443 migration cycle within the BIOT MPA.

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