1	Changes in surface water drive the movements of Shoebills
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20 Abstract

21 Animal movement is mainly determined by spatial and temporal changes in resource 22 availability. For wetland specialists, the seasonal availability of surface water may be a major determinant of their movement patterns. This study is the first to examine the movements of 23 24 Shoebills (Balaeniceps rex), an iconic and vulnerable bird species. Using GPS transmitters 25 deployed on 6 immature and 1 adult Shoebills over a 5-year period, during which 4 immatures matured into adults, we analyse their home ranges and distances moved in the Bangweulu 26 27 Wetlands, Zambia. We relate their movements at the start of the rainy season (October to December) to changes in Normalised Difference Water Index (NDWI), a proxy for surface 28 water. We show that Shoebills stay in the Bangweulu Wetlands all year round, moving less than 29 3 km per day on 81% of days. However, average annual home ranges were large, with high 30 31 individual variability, but were similar between age classes. Immature and adult Shoebills responded differently to changes in surface water; sites that adults abandoned became drier, 32 33 while sites abandoned by immatures became wetter. However, there were no differences in NDWI of areas used by Shoebills before abandonment and newly selected sites, suggesting 34 that Shoebills select areas with similar surface water. We hypothesise that the different 35 responses to changes in surface water by immature and adult Shoebills are related to age-36 specific optimal foraging conditions and fishing techniques. Our study highlights the need to 37 understand the movements of Shoebills throughout their life cycle to design successful 38 39 conservation actions for this emblematic, yet poorly known, species.

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41 Keywords (3-10): Balaeniceps rex, Bangweulu Wetlands, GPS tracking, Waterbird ecology,
42 Surface water, Home range, Age classes, Animal movement, Conservation

43 Introduction

One of the key challenges in ecology is to understand how environmental fluctuations drive 44 45 animal movements. Changes in the environment can alter resource distribution, which consequently determines animal migratory ^{1–3} and, local, movements ^{4,5}. In wetlands, the 46 distribution of surface water is one of the main determinants of species' spatial distribution ^{6–} 47 ⁸ and individual movements ^{9,10}. In tropical systems with strongly seasonal environments, 48 prolonged periods of drought followed by extreme floods can lead to striking changes in 49 50 habitat suitability ^{9,11} and drive the large-scale movements of waterfowl ¹², due to fluctuations in the abundance and availability of foraging resources ¹³. 51

The way individuals explore the environment can change as they age ^{14,15}, and recent 52 advances in GPS tracking technology and increases in device longevity, have enabled the 53 detailed study of individual movements for several years or even throughout lifetimes. This has 54 unravelled differences between adults and juveniles in space use ^{15,16}, habitat selection ^{14,17}, 55 and timing ^{18–20} and efficiency of movements ^{20–22}. Understanding the drivers of movement of 56 long-lived birds relies on information on the spatial and temporal dynamics of movement at 57 different ages in relation to environmental variables. Such information has only been available 58 relatively recently, through the integration of data from GPS trackers with remotely sensed 59 environmental data ^{23–25}. Indices based on satellite imagery have been increasingly used to 60 interpret environmental conditions and infer ecological processes ^{23,25}. The Normalised 61 Difference Water Index (NDWI) proposed by McFeeters ²⁶ is an index that uses remotely 62 sensed imagery to map surface water. The NDWI delineates and highlights open water by 63 distinguishing it from vegetation and bare soil, and has mostly been used to map waterscapes 64 in urban settings ^{27,28}. More recently, this index has been used to map surface water for animal 65

66 movement studies ¹², and to identify suitable habitat and inform area protection for shorebird
67 species ²⁹.

The Shoebill (Balaeniceps rex) is an iconic wetland specialist, with a patchy distribution 68 in central-eastern Africa, from South Sudan to Zambia ^{30,31}. The Shoebill is a large long-lived 69 70 species, categorised as Vulnerable by the IUCN. Shoebills have a declining population trend, due to habitat degradation and loss, illegal bird trade and disturbance by humans ^{31,32}. The 71 global population estimate for the species is 5,000-8,000 individuals, although large 72 73 uncertainty around this estimate exists, given that this species is cryptic and found in inaccessible areas ³¹. Shoebills inhabit permanent swampy wetlands with seasonal flooded 74 grasslands, where they prey on fish in shallow waters or use floating vegetation as fishing sites 75 ^{33,34}. Despite being a highly emblematic species, there are very few studies on Shoebill ecology, 76 and existing studies have focused on deriving local population estimates ^{30,35}, and better 77 understanding their foraging ^{33,34} and breeding ecology ^{36,37}. This species is believed to be 78 sedentary, staying in the same region all year long ^{32,38}; however, to date, the movement 79 ecology of the Shoebill is completely unknown, which is unsurprising given that very few birds 80 have ever been ringed and no previous tracking studies have occurred on this species. Being a 81 species of high conservation concern, as well as an important source of tourism revenue ³⁹, it 82 is critical to improve our knowledge of Shoebill ecology and habitat requirements to implement 83 effective conservation measures ³². 84

Using GPS tracking data collected over 5 years, we characterise the movements of immature and adult Shoebills in the Bangweulu Wetlands, Zambia. In common with many other areas occupied by Shoebills, the Bangweulu Wetlands undergoes dramatic changes in water levels between the dry (breeding season) and the wet season. We therefore hypothesise that changes in surface water drive the movements of Shoebills, and that their selected areas have

90 similar surface water. Using the NDWI as a proxy for surface water, we compare 1) the NDWI of areas while Shoebills were present with the NDWI of the same areas the week after the birds 91 left (to examine how these abandoned areas change), and 2) the NDWI of areas used by 92 Shoebills the last week before abandonment with the NDWI of areas the first week after 93 94 Shoebill arrival (to examine whether they select for similar habitats in relation to surface water). We explore these questions for both adult and immature birds. By analysing the 95 96 movements of Shoebills in different life stages, and how these movements relate to available 97 surface water, a key environmental factor for wetlands, our goal is to improve our ecological understanding of Shoebills, to ultimately inform the conservation of this mostly unknown and 98 emblematic species. 99

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101 Results

We tracked 11 Shoebills in the Bangweulu Wetlands, Zambia, between December 2011 and 102 October 2018 and collected 119,321 valid GPS positions (Table 1). We obtained 47,134 GPS 103 104 positions for 6 Shoebills tracked as immatures and 44,985 GPS positions of 5 Shoebills tracked 105 as adults. All other GPS positions were from juveniles (n=4), which died or disappeared before they became immatures and were thus excluded from this research, also because they 106 107 remained near the nest for a long period after fledging. From the adult GPS positions, 28,057 locations were from 4 immature Shoebills that matured into adults during the tracking period, 108 109 and 16,928 GPS locations from the one individual tagged as a breeding adult.

Table 1 – Information for the tracked Shoebills: age of the individual at the time of logger
deployment, start and end dates of tracking, total number of valid GPS positions, excluding
outliers, and total number of tracking days as immature and adult.

				Number of	Immature	Adult
Bird ID	Age	Start of tracking	End of tracking	GPS	tracking	tracking
				positions	days	days
521		03/12/2011	28/10/2014	10,763	617	-
514		09/09/2012	26/11/2013	2,187	-	-
518		15/09/2012	21/04/2013	2,537	-	-
520		03/08/2013	02/05/2018	19,335	725	623
509		26/08/2013	05/03/2018	18,998	724	634
510	Juvenile	30/08/2013	14/11/2013	962	-	-
515		02/09/2013	15/05/2017	16,138	728	347
512		10/09/2013	29/10/2018	21,224	724	863
516		15/11/2013	05/01/2015	3,480	-	-
511		28/10/2014	08/06/2016	6,769	366	-
517	Adult	29/07/2013	15/08/2017	16,928	-	1,444

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115 Spatial analysis

Shoebill annual home range was similar in size for adults and immatures (mean 95% kernel =
1,514 km² (± 1,172) and 1,547 km² (± 1.296) for adults and immatures, respectively; Figure 1,
Table 2). There was large individual variation in home range size, both for immatures (range
95% kernel: 233 km² and 2,628 km²) and adults (range 95% kernel: 304 km² and 3,375 km²)
(Table 2).

Table 2 – Individual average annual home range area (in $km^2 \pm standard$ deviation) and total average home range of immature and adult Shoebills, estimated as the 95% and 50% kernel, based on the GPS tracking periods indicated in Table 1.

	Immat	ure	Adu	ılt	
ID	95% kernel	50% kernel	95% kernel	50% kernel	
	(km²)	(km²)	(km²)	(km²)	
521	233 (± 318)	46 (± 63)	-	-	
520	1,094 (± 585)	212 (± 129)	1,039 (± 866)	145 (± 124	
509	2,458 (± 617)	431 (± 59)	2,167 (± 241)	389 (± 85)	
515	2,628 (± 2,309)	403 (± 359)	3,375	652	
512	1,585 (± 360)	257 (± 56)	2,167 (± 474)	343 (± 133	
511	981	200	-	-	
517	-	-	304 (± 108)	54 (± 15)	
All ndividuals	1,547 (± 1,296)	263 (± 204)	1,514 (± 1,172)	247 (± 204	





Figure 1 – Cumulative 95% and 50% kernel density estimations for all tracked (A) immature
and (B) adult Shoebills, and cumulative 50% kernel density estimation for each (C) immature
and (D) adult individual, based on the GPS tracking periods indicated in Table 1. The dashed
line indicates the border of the Bangweulu Wetlands Game Management Area.

For both adults and immatures, the distribution of the maximum daily distance moved
was highly skewed (Figure 2). On most days both age classes moved relatively short distances
(median values; adults: 0.84 km/day, immatures: 0.73 km/day). For both age classes, on 81%
of days, birds moved less than 3 km (Figure 2).



Figure 2 – Frequency of the maximum daily distance (in Km's) moved by (A) immature and (B)
adult Shoebills. The dashed red line indicates the threshold that captures 80% of movements,
used to define Shoebills' Moving Days.

The mean maximum daily distances moved varied throughout the year, particularly for adult Shoebills. During the breeding season, from June until October, adults performed shorter movements, with the mean maximum daily distance moved being the lowest in August (1.1 km per day). In October, towards the end of the breeding season, adult mean maximum daily distance started to increase, peaking in December (10.5 km per day). Immature Shoebills show less variation in movement distances over the year. Birds moved least in September (1.9 km per day), while movement distances peaked in May to 5.5 km per day (Figure 3).



Figure 3 – Boxplots of the mean maximum daily distance per month, for individual immature 152 153 and adult Shoebills, between 2011 and 2018. Data is organised to start at the beginning of the breeding season (May). The boxes represent the 25th, 50th and 75th percentiles of the mean 154 maximum daily distance. Whiskers the 1.5 times the value of inter-quantile range, with values 155 outside this range plotted as black dots. The dashed line above the plot indicates the dry season 156 (May to October) and the wet season (November to April). The shaded area highlights the 157 158 period between October and December, with an increase of adult mean maximum daily 159 distances.

161 Influence of NDWI on Shoebill movements

162 On over 80% of the days, Shoebills moved less than 3 km, thus 'Moving Days' were defined as 163 days when Shoebills moved more than 3 km (Figure 2) and the regions where birds stayed for 164 a minimum of two days between Moving Days were classified as 'Areas' (further details in the Methods section). Between October and December of 2013-2017, across the 5 adults we located 39 different Areas, and, in 2014 and 2015, across the 6 immatures, we identified 33 Areas. Immature birds stayed on average 14±19 days in Areas, whereas adult birds spent 17±24 days in Areas before moving to another location. These locations always had negative daily mean NDWI values, indicating that Shoebills were not in open water areas and selected relatively dry regions.

171 We found that the NDWI of the Areas used by Shoebills between October and December was statistically different from the NDWI of the same Areas the week after the birds 172 173 abandoned (Table 3), both for immature and adult Shoebills. However, these relationships 174 differed between the age classes. Adult Shoebills used Areas with an average NDWI value of -0.52, varying from -0.68 to -0.10. The week after adults left the Area, it became drier with the 175 176 NDWI decreasing to an average of -0.57 (range -0.79 to -0.18). In contrast, for immatures, the 177 mean NDWI of Areas used was -0.53, with a minimum NDWI value of -0.76 and maximum of -178 0.10.After abandonment, the average NDWI of these Areas increased to -0.43 (range -0.66 to -0.07), indicating that the Areas became wetter (Figure 4 A and B). The variance explained by 179 180 the immature model was higher (marginal R-squared 0.27) than by the adult model (marginal R-squared 0.14), and in both cases the random factors slightly increased the R-squared 181 182 (immature conditional R-squared 0.32; adult conditional R-squared 0.15) (Table 3).

We did not find a statistical difference between the NDWI of Areas used by Shoebills the week before abandonment, and the NDWI of the newly colonised Areas the first week after Shoebill arrival; this was the case for both adults and immatures (Table 3). The mean NDWI of Areas used by adult Shoebills the week before abandonment and the week after arrival was -0.54 (ranging from -0.76 to -0.16). For immatures, the NDWI of Areas before abandonment

188 was -0.49 (varying from -0.76 to -0.07), compared to -0.50 (from -0.66 to -0.09) the week after
189 arrival (Figure 4 C and D).

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191 Table 3 – Results of the GLMM models comparing the values of: a) daily mean NDWI of Areas 192 before Shoebill abandonment, with the values of NDWI after the birds abandoned the Area, 193 using year and Area nested within bird ID as random effects. Daily mean NDWI was 194 transformed as a second-degree polynomial (poly 1 and poly 2). b) daily mean NDWI of used 195 Areas the last week before abandonment, with the daily mean NDWI of used Areas the first 196 week after arrival, using year and Area nested within bird ID as random effects.

		A = -	Demonster		7	Duralius	Marginal	Conditional
	Model	Age	Parameter	Estimate (SE)	Z-value	P-value	R-squared	R-squared
			Intercept	0.17 (0.19)	0.87	0.384		
- \		Immature	NDWI (poly1)	33.87 (3.38)	10.02	<0.001	0.270	0.320
a)	Comparison of NDWI of		NDWI (poly2)	-7.35 (2.70)	-2.72	0.007		
	Areas before and after Shoebill abandonment		Intercept	0.06 (0.08)	0.76	0.445		
	Shoediii adandonment	6 -l l+	NDWI (poly1)	-23.12 (2.65)	-8.72	<0.001	0.144	0.151
		Adult	NDWI (poly2)	15.03 (2.73)	5.50	<0.001		
b)	Comparison of NDWI of	Immature	Intercept	-0.03 (0.54)	-0.05	0.963	<0.001	<0.001
	used Areas the last week	IIIIIIature	NDWI	-0.05 (1.01)	-0.05	0.957	<0.001	<0.001
	before abandonment and		Intercept	-0.50 (0.46)	-1.10	0.272	-0.001	-0.001
	the first week after arrival	Adult	NDWI	-0.95 (0.85)	-1.11	0.268	<0.001	<0.001
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209 Discussion

We described for the first time the annual home range sizes and variation in distances movedover the year for adult and immature Shoebills, providing evidence of age-related differences

in their movement ecology. Furthermore, we show that movement patterns of Shoebills were associated with changes in surface water, but these changes contrasted between age classes, with adult abandoning sites that became drier, whereas immatures abandoned sites that became wetter. Despite the small number of tracked Shoebills, which can make the generalisation of our results to other Shoebill populations challenging, this species inhabits similar habitats throughout their narrow distribution range ³¹, and thus their movement ecology is likely influenced by analogous environmental drivers.

219 Shoebills in the Bangweulu Wetlands were largely sedentary, moving less than 3km on over 80% of days. The main prey of Shoebills in the Bangweulu Wetlands are catfish, which 220 they catch mainly through the tactics of stand and wait on top of floating vegetation ^{33,34,40}. 221 Indeed, field studies in the Bangweulu Wetlands found that they spent 85% of the time 222 performing low-energy activities, such as standing, sitting and preening ³⁴. Walking and flying 223 224 behaviours may also be associated with foraging, given that a Shoebill strike may disturb the prey and require a move to a different location ⁴⁰. Therefore, much of these Shoebill's daily 225 226 movements were likely related to foraging events or searching for suitable foraging habitat.

The average annual home range of Shoebills was around 1,500 km² which is larger than 227 for similar species, such as Abdim's Storks (Ciconia abdimii) in Niger (10–120 km^{2 41}). However, 228 there was large individual variation in home range size, both for adult $(304 - 3,375 \text{ km}^2)$ and 229 immature Shoebills (233 – 2,628 km²). Other studies have shown large individual variation in 230 231 home range size of similar wetland species, such as Wattled Cranes (Bugeranus carunculatus), with 95% kernel density estimates varying between 0.4 and 110.4 km² ⁴², Mauritanian 232 Spoonbills (*Platalea leucorodia balsaci*), with home ranges varying from 23 to 101 km^{2 43}, or 233 234 American White Pelicans (*Pelecanus erythrorhynchos*) summer home range varying between 177 and 4,710 km^{2 44}. These variations in individual home range size in the same habitat and 235

within the same species show that animal movement is more complex than a simple reflection
of underlying resource distribution ^{11,45}, and other factors (e.g. social attraction/repulsion) may
also influence individual distribution ^{46,47}.

Several factors can influence the home range size in birds, such as age ^{14,15,44,48}, sex 239 ^{14,44,49} and degree of individual specialisation in particular foraging areas ⁵⁰. Shoebills do not 240 exhibit strong sexual dimorphism, and the birds in this study were not genetically sexed, thus 241 242 it was not possible to investigate possible sex differences in home range size. We did not find 243 age-related differences in annual home range size, and although 2 individuals slightly decreased their home range size by an average of 173 km² as they aged from immatures to 244 adults, 2 other individuals increased their home range size by an average of 665 km² as they 245 matured. However, there is a suggestion of individual consistency, since the individuals with 246 smallest and largest home ranges as immatures maintained smaller and larger home ranges as 247 248 adults (Table 2). In many situations, breeding adults have smaller home ranges than non-249 breeders during part of the year because their movements are constrained by the location of their nest site ^{15,16}. Although animals in areas of higher productivity tend to have smaller home 250 ranges ^{11,51}, this might not be the case in swampy areas. In the wet season, with an increase in 251 water levels, Shoebill prey species occupy larger areas of the swamps, forcing birds to increase 252 their home ranges. Adult birds, with more experience, may build up knowledge of the 253 254 landscape, occupying the most suitable foraging locations and outcompeting less experienced birds ⁵². Consequently, a possible increase in adult home range size during the wet period may 255 be counterbalanced by the seasonal constrain of the nest site location, resulting in 256 approximately the same average home range size for adult and immatures. 257

Indeed, immature Shoebills moved consistent distances throughout the year, whileadults moved smaller distances during the breeding season (May-October), particularly during

the incubation and chick-rearing period (June-September). During the breeding season, adults 260 forage close to the nest, moving smaller distances and occupying smaller home ranges ^{15,16}. 261 Shoebills chicks hatch in June-July, and until the chicks are about 40 days old, at least one adult 262 is constantly on the nest ³⁷. Later in the breeding season (September and October), adult daily 263 264 distances moved started to increase. Shoebills build their nests on top of floating vegetation ^{36,53}, but as the breeding season progresses, the water levels recede to the point that by the 265 end of the breeding season, the nests are resting on solid ground ³⁷. This might also decrease 266 267 the suitability of the foraging areas surrounding the nest, forcing adult birds to increase their 268 daily moved distances as the breeding season progresses to find suitable foraging sites.

Environmental factors can also determine movement and home range size in birds, 269 and, for water-dependent species, the spatial and temporal distribution of surface water is one 270 of the main drivers of movement ^{9,10,12}. Bird species respond differently to changes in water 271 availability, with some functional groups responding to sequences of flooding and drying 272 patterns, while others respond immediately to changes in flooded area ⁵⁴. For example, Black 273 Storks wintering in West Africa move as the rivers begin to dry ⁵⁵, Mallard (*Anas platyrhynchos*) 274 movements are highly predictable and strongly linked to the presence of surface water ¹⁰ and 275 Grey Teal (Anas gracilis) fly hundreds of kilometres directly towards temporary water sources 276 ⁵⁴. In Southern Africa, the patterns of rainfall and primary productivity are the main drivers of 277 large-scale movements of Egyptian Geese (Alopochen aegyptiaca) and Red-billed Teal (Anas 278 erythrorhyncha) ¹². Here, we show that drying and flooding patterns of the Bangweulu 279 280 Wetlands at the start of the rainy season are important drivers in the movement of Shoebills. In the Bangweulu Wetlands, November marks the start of the rainy season, being the 281

282 month with lowest water levels in this region ⁵⁶. Our results show that between October and
 283 December, Shoebills occupied areas of low surface water availability (low NDWI values), which

is likely the most available habitat. There were, however, differences in how adult and 284 immature birds responded to changes in surface water. While adults seemed to abandon areas 285 that became drier, immatures abandoned areas that became wetter, suggesting age-related 286 differences in habitat use or foraging strategies. Moreover, the areas selected by Shoebills had 287 288 the same surface water as the areas they were previously occupying, which suggests a selection for an optimal surface water level by this species. Water-depth limits non-diving 289 waterbirds foraging ranges, by directly restricting the accessibility of the habitats due to birds 290 morphology (e.g. neck and metatarsus) ⁶. Consequently, Shoebills foraging locations are also 291 restricted to the water-depths suitable for foraging. 292

We hypothesise that the different movements in response to surface water between 293 age-classes might be related to prey availability and optimal foraging conditions. Immature 294 birds tend to be less efficient foragers ^{21,57} and occupy less optimal foraging locations ⁵². 295 296 Distributions of waterbirds are greatly influenced by the hydrology of the wetlands and distribution of food resources ⁵⁸, since different species have different foraging methods and 297 depend on particular water depths and prey vulnerability ⁶. Shoebills are typically solitary birds, 298 but they occasionally concentrate in drying pools of water, where fish may become highly 299 abundant ³³. Immature Shoebills may take advantage of this recession of the water level, which 300 301 promotes the availability of prey ¹³ and thus would be suitable areas for immatures to gain experience in capturing prey. Shoebills also forage in deep water, using floating vegetation as 302 fishing sites and then diving forward, described by Guillet ³³ as a "peculiar and complicated 303 technique called collapsing". Although birds using this technique have lower foraging success 304 than on flooded grassland, the catfish caught in deeper waters are on average larger than on 305 flooded grasslands ³⁴, as larger catfish prefer deeper waters ^{34,59}. Therefore, immature birds 306 307 might prefer drier areas with higher abundance of relatively smaller prey, whereas adults

having already mastered the highly specialised deep-water foraging technique, might prefer 308 309 flooded areas with larger prey, and thus greater rewards per capture. Nonetheless, our interpretations are based on, as yet, unverified validation of the NDWI in swampy areas, 310 particularly the areas used by Shoebills that are typically densely vegetated and have water 311 with low oxygen content ³³, which can pose constraints on the identification of water features 312 using satellite imagery ⁶⁰. In future research, newly available satellite imagery (e.g. Sentinel-2, 313 launched in 2015) and recently created indexes (e.g. Xu 2006⁶⁰) may provide further detail on 314 315 how surface water influences the movements of wetland species. However, these indexes need to be validated in swampy wetlands, which may have their own unique characteristics ⁶¹. 316 317 Moreover, changes in water surface may not be the only environmental variable driving the movement of Shoebills. Henry et al. ¹² explored the main environmental variables 318 influencing the movement decisions of Egyptian Geese and Red-billed Teal in Southern Africa, 319 320 and although changes in surface water appeared in several of their models, suggesting that the 321 flooding and drying patterns of wetlands have some predictive power, rainfall and primary productivity were found to be more important in explaining movement patterns in these 322 323 species. The relatively small variance explained by our models also suggest that other nonmeasured variables likely play a role in driving the movements of Shoebills; we therefore 324 suggest for future research to complement the use of NDWI with high temporal and spatial 325 326 records of rainfall and NDVI to further explore the drivers of movement patterns and spatial 327 distribution of Shoebills in the Bangweulu Wetlands.

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329 Methods

330 Study area and data collection

This study was conducted in the Bangweulu Wetlands, a Game Management Area (GMA) 331 332 located in the Muchinga province in north-eastern Zambia (approximately between 11°40' to 12°34'S and 29°78' to 30°87'E). The Bangweulu Wetlands consist of miombo (*Brachystegia* sp.) 333 woodlands, grasslands, floodplains, seasonal swamps, and permanent wetlands ⁶². This reserve 334 335 is classified as an Important Bird Area and the area of Chikuni is classified as a Ramsar Site ⁶³. The climate is characterised by a heavy rainfall season from November to April, with a total 336 annual precipitation of 1,200 mm to 1,400 mm ^{56,64}. The lowest water levels occurs in 337 November, and the mean annual water level difference is 1.4 m^{65,66}. This area harbours the 338 southernmost population of Shoebills ^{31,38}, however the size of this population is largely 339 unknown. In 1984, the first Shoebill census in the Bangweulu Wetlands estimated the 340 population at 200-300 individuals ⁶⁷. Nevertheless, a large area of the wetland remained un-341 surveyed ⁶² and, in a more recent survey, Roxburgh and Buchanan ³⁵ provided an estimated 342 343 population size of 1,296 individuals, although this estimate was based on very few sightings, 344 and there was considerable uncertainty around this estimate (95% confidence interval: 477-2,372). 345

Between August and September of 2011 to 2014, 10 juvenile and 1 adult Shoebills were 346 fitted with 70g satellite-based GPS-trackers (Solar Argos/GPS PTTs, Microwave Telemetry) 347 (Table 1). The transmitters were fitted using the body-loop attachment method, with a Teflon-348 349 tube harness. Eight pre-fledging juveniles were tagged on their nests when they were on average 84 days old (range 80 - 89 days). Shoebills fledge at approximately 95-105 days ³⁸. Two 350 juveniles (511 and 521) were raised in a recovery centre after being confiscated from the illegal 351 bird trade, and fitted with the GPS transmitter before being released at unknown ages, but 352 likely older than 80-89 days of the other birds. Only one Shoebill (517) was tagged as an adult, 353 354 which was caught at its nest site. Tracking devices, including harness, weighted 80g,

representing 1.3-1.6% of the body mass of birds at the time of deployment (4,900-6,300g). Licences to catch and deploy the tracking devices were provided by the Zambia Wildlife Authority (now Department of National Parks and Wildlife (DNPW)), and the work was approved by the University of Cape Town Science Faculty Animal Ethics Committee.

359 Data processing and spatial analysis

The trackers provided a GPS fix every 1-hour between 6 A.M. and 6 P.M., GMT+2, which corresponds to the activity period of Shoebills. The transmitters provided location (latitude and longitude) with a mean error of 18 m ⁶⁸. We considered all valid GPS locations until the transmitter failed or when there was no movement for several days, indicating death or loss of the GPS transmitter. GPS data was filtered for outliers based on unrealistic movements or speed (more than 150 km/h between two consecutive hourly locations) and visually inspecting the tracks.

Birds were classified as juveniles until the start of the following breeding season (1st of May), as immatures during the second and third year and as adults from the fourth year onwards, since Shoebills start to breed after three years ³⁸. For this study, we only considered the movements of immature and adult birds, since first year juveniles remained near the nest for a long period after fledging ³⁷. Six individuals provided more than 1 year of data, maturing from juvenile to immature birds, and four immature birds provided more than 3 years of data, becoming adults (Table 1).

We estimated the annual home range area of individual immature and adult birds using Kernel Density Estimation, with *h-ref* algorithm and grid size of 500 m, using R package adehabitatHR⁶⁹. The year was defined from the start of the breeding season (May) until the

following April. We also calculated cumulative home ranges of immature and adults, across allyears and individuals, to visualise the area used by this species in the Bangweulu Wetlands.

We quantified the maximum range of Shoebill individual daily movements by calculating the distance between all GPS locations each day and selecting the maximum value (hereafter maximum daily distance). All distances were calculated using R package *geosphere* 70. To understand how movements changed throughout the year for immature and adult birds, we calculated the mean maximum daily distance per month of each individual. All data processing and analysis were performed in R 3.6.1. ⁷¹.

385 Influence of NDWI on Shoebill movements

We analysed Shoebill movements between 2013 and 2017, in relation to changes in surface 386 water from October to December each year. During this period, the levels of surface water 387 change dramatically in the Bangweulu Wetlands, as the rainy season typically starts in 388 389 November. This period also encompasses the end of the Shoebill breeding season and birds 390 are less constrained by the location of the nests. We compared the NDWI of areas used by Shoebills prior to and after they abandoned them, and compared the NDWI of used areas the 391 last week before abandonment with the NDWI of the newly selected areas, the first week after 392 arrival. 393

To understand when Shoebills performed large movements, we analysed the frequency of the maximum daily distances. We defined a size threshold (in km's) which captured 80% of smaller scale movements and considered the remaining 20% as '*Moving Days'*. Here we also accounted for movements performed during the night, by calculating the distance between the first GPS location of the day and the last GPS location of the previous day. Movements performed during the night were allocated to the previous day. We classified as '*Areas*', the regions where birds stayed for a minimum of two days between Moving Days. 401 We computed the 95% minimum convex polygons (MCPs) of these Areas and, to understand 402 if birds moved to a different geographical area or remained in a similar location after a Moving 403 Day, we overlayed the MCPs of two consecutive Areas. If the two MCPs overlapped, we 404 considered the individual to have remained in the same Area; if they did not overlap, we 405 considered that the individual moved to a different Area. MCPs were calculated using R 406 packages *sp* 72,73 and *adehabitatHR* ⁶⁹.

We used the NDWI as a proxy for surface water and calculated this index for the 407 408 Bangweulu Wetlands for all weeks of October until January. When using satellite imagery there is a trade-off between temporal and spatial resolution. In this study, we favoured imagery with 409 410 higher temporal resolution, using satellite imagery from MODIS Terra Surface Reflectance with 8-days and 500 m resolution ⁷⁴, since the pixel size of 500 m was still smaller than the analysed 411 range of movements. All images had a minimum of 92% of pixels with good quality and a 412 413 maximum of 2% of pixels not classified due to cloud cover or other reasons. To calculate the NDWI, we used McFeeters ²⁶ formula: 414

415
$$NDWI = \frac{(Green - NIR)}{(Green + NIR)}$$

where *Green* is MODIS Band 4 (545-565 nm wavelength) and *NIR* (near infrared) is MODIS Band
2 (841-876 nm). The NDWI varies between 1, indicating open water features, and -1, indicating
a dry area, on a gradient of surface water. This index was interpreted comparatively, e.g. an
area of NDWI of -0.6 is drier than an area of NDWI -0.5 (Figure 5). All satellite imagery
manipulation was performed using R packages *raster* ⁷⁵ and *rgdal* ⁷⁶ and *rgeos* ⁷⁷.



422

Figure 5 – A) Location of the Bangweulu Wetlands Game Management Area (in red) within
Africa (light grey) and Zambia (dark grey); B) Changes in NDWI between the start of October
(week between 30/09/2015 and 07/10/2015), just prior to the start of the rainy season and
C) the end of December (week between 19/12/2015 and 26/12/2015), which is during the
rainy season. The red dashed line indicates the border of the Bangweulu Wetlands Game
Management Area.

430 To test if Shoebills move due to changes in surface water, we extracted the daily mean NDWI of the GPS positions of Shoebills while they were in a particular Area. We then compared 431 the locations where birds were present, with the locations one week after the birds abandoned 432 the Area. We used binomials Generalised Linear Mixed Models (binomials GLMMs), with 433 434 presence (0) / abandonment (1) of Shoebills as the response variable, daily mean NDWI as the 435 fixed effect, and year and Area nested within bird ID as random effects, to account for lack of 436 independence of measures within years and within the Areas used by different Shoebills. Due to the non-linearity of the relationship between Shoebill presence/abandonment and NDWI 437 (as areas Shoebills abandoned could have become drier or wetter, *i.e.*, with lower or larger 438 NDWI values), we introduced the NDWI as a second-degree polynomial term in the GLMM. We 439

calculated the marginal and conditional R-squared, to assess the variance explained by the
fixed effect of the model (mean daily NDWI), and the fixed and random effects of the model,
respectively. We built two models, one for adults and another for immatures, to evaluate if the
two age groups responded differently to changes in surface water.

444 To understand if Shoebills select areas of similar surface water when they move, we 445 compared the Shoebill locations the first week after arrival (1) with the locations the last week before they abandoned an Area (0). We tested this hypothesis for immatures and adults. We 446 used binomials GLMMs, with newly selected area (1) / previously occupied area (0) as the 447 response variable, mean daily NDWI as a fixed effect, and year and Area nested within bird ID 448 as random factors. We assessed the variance explained by the model using marginal and 449 conditional R-squared. GLMMs were computed using R package *lme4*⁷⁸, and R-squared values 450 computed using the package MuMIn⁷⁹. 451

452

453 Declarations

454 Approval for animal experiments

Licences to catch and deploy the tracking devices were provided by the Zambia Wildlife Authority (now Department of National Parks and Wildlife (DNPW)). The work was carried out with approval from the University of Cape Town Science Faculty Animal Ethics Committee (permit number: 2011/V14/AA). Capture, handling and tagging procedures were carried out by RHEM, qualified in 2007 under the Article 9 of the Experiments on Animals Act in The Netherlands. No bird was injured by the capturing/handling procedure.

461 Consent for publication

462 Not applicable.

463 Availability of data and materials

464 The datasets used and analysed during the current study are available from the corresponding465 author on reasonable request.

466 Competing interests

467 The authors declare that they have no competing interests.

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476 Authors' contributions

AA, RHEM, FJW and MA conceptualised the study. AA, RHEM, MA, AMAF and FJW designed
the methodology. RHEM conducted the fieldwork and data collection. MA analysed the data
and wrote the manuscript. All authors read, edited, and approved the final manuscript.

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487		
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675 Tables

Table 1 – Information for the tracked Shoebills: age of the individual at the time of logger
deployment, start and end dates of tracking, total number of valid GPS positions, excluding
outliers, and total number of tracking days as immature and adult.

				Number of	Immature	Adult
Bird ID	Age	Start of tracking	End of tracking	GPS	tracking	tracking
				positions	days	days
521		03/12/2011	28/10/2014	10,763	617	-
514		09/09/2012	26/11/2013	2,187	-	-
518		15/09/2012	21/04/2013	2,537	-	-
520		03/08/2013	02/05/2018	19,335	725	623
509		26/08/2013	05/03/2018	18,998	724	634
510	Juvenile	30/08/2013	14/11/2013	962	-	-
515		02/09/2013	15/05/2017	16,138	728	347
512		10/09/2013	29/10/2018	21,224	724	863
516		15/11/2013	05/01/2015	3,480	-	-
511		28/10/2014	08/06/2016	6,769	366	-
517	Adult	29/07/2013	15/08/2017	16,928	-	1,444

Table 2 – Individual average annual home range area (in km² ± standard deviation) and total
average home range of immature and adult Shoebills, estimated as the 95% and 50% kernel,
based on the GPS tracking periods indicated in Table 1.

	Immat	ture	Adult			
ID	95% kernel	50% kernel	95% kernel	50% kernel		
	(km²)	(km²)	(km²)	(km²)		
521	233 (± 318)	46 (± 63)	-	-		
520	1,094 (± 585)	212 (± 129)	1,039 (± 866)	145 (± 124		
509	2,458 (± 617)	431 (± 59)	2,167 (± 241)	389 (± 85)		
515	2,628 (± 2,309)	403 (± 359)	3,375	652		
512	1,585 (± 360)	257 (± 56)	2,167 (± 474)	343 (± 133		
511	981	200	-	-		
517	-	-	304 (± 108)	54 (± 15)		
All	4 5 47 (+ 4 200)	262 (+ 204)	4 544 (1 4 472)	247/1204		
individuals	1,547 (± 1,296)	263 (± 204)	1,514 (± 1,172)	247 (± 204		

Table 3 – Results of the GLMM models comparing the values of: a) daily mean NDWI of Areas
before Shoebill abandonment, with the values of NDWI after the birds abandoned the Area,
using year and Area nested within bird ID as random effects. Daily mean NDWI was
transformed as a second-degree polynomial (poly 1 and poly 2). b) daily mean NDWI of used
Areas the last week before abandonment, with the daily mean NDWI of used Areas the first
week after arrival, using year and Area nested within bird ID as random effects.

Model	A .co	Daramatar	Estimata (SE)	7 volue	P-value	Marginal	Conditional
Model	Age	Parameter	Estimate (SE)	Z-value	r-value	R-squared	R-squared
		Intercept	0.17 (0.19)	0.87	0.384		
c) Comparison of NDWI of Areas before and after Shoebill abandonment	Immature	NDWI (poly1)	33.87 (3.38)	10.02	<0.001	0.270	0.320
		NDWI (poly2)	-7.35 (2.70)	-2.72	0.007		
		Intercept	0.06 (0.08)	0.76	0.445		
	Adult	NDWI (poly1)	-23.12 (2.65)	-8.72	<0.001	0.144	0.151
	Auuit	NDWI (poly2)	15.03 (2.73)	5.50	<0.001		
d) Comparison of NDWI of	Immature	Intercept	-0.03 (0.54)	-0.05	0.963	<0.001	<0.001
used Areas the last week		NDWI	-0.05 (1.01)	-0.05	0.957		
before abandonment and		Intercept	-0.50 (0.46)	-1.10	0.272	<0.001	<0.001
the first week after arrival	Adult	NDWI	-0.95 (0.85)	-1.11	0.268		
701							

706 Figure legends

Figure 1 – Cumulative 95% and 50% kernel density estimations for all tracked (A) immature
and (B) adult Shoebills, and cumulative 50% kernel density estimation for each (C) immature
and (D) adult individual, based on the GPS tracking periods indicated in Table 1. The dashed
line indicates the border of the Bangweulu Wetlands Game Management Area.

Figure 2 – Frequency of the maximum daily distance (in Km's) moved by (A) immature and (B)
adult Shoebills. The dashed red line indicates the threshold that captures 80% of movements,
used to define Shoebills' Moving Days.

714 Figure 3 – Boxplots of the mean maximum daily distance per month, for individual immature 715 and adult Shoebills, between 2011 and 2018. Data is organised to start at the beginning of the breeding season (May). The boxes represent the 25th, 50th and 75th percentiles of the mean 716 maximum daily distance. Whiskers the 1.5 times the value of inter-quantile range, with values 717 outside this range plotted as black dots. The dashed line above the plot indicates the dry season 718 (May to October) and the wet season (November to April). The shaded area highlights the 719 720 period between October and December, with an increase of adult mean maximum daily distances. 721

Figure 4 – Diagram describing the analysed spatial and temporal relationships of Shoebill movements. The plots show the predicted mean daily NDWI values (solid line) and 95% confidence interval (shaded areas) for (A) adults and (B) immatures of 'Areas' before and after Shoebill abandonment, and for (C) adults and (D) immatures before Shoebill abandonment and after Shoebill arrival in a new 'Area'. The boxplots display the observed values of daily mean NDWI, with the boxes representing the 25th, 50th and 75th percentiles and the whiskers the 1.5

728	times the value of inter-quantile range. Values outside this range are plotted as grey dots.
729	Brown colours indicate adult data and blue colours indicate immature data.
730	Figure 5 – A) Location of the Bangweulu Wetlands Game Management Area (in red) within
731	Africa (light grey) and Zambia (dark grey); B) Changes in NDWI between the start of October
732	(week between 30/09/2015 and 07/10/2015), just prior to the start of the rainy season and
733	C) the end of December (week between 19/12/2015 and 26/12/2015), which is during the
734	rainy season. The red dashed line indicates the border of the Bangweulu Wetlands Game
735	Management Area.
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Figure 1 – Cumulative 95% and 50% kernel density estimations for all tracked (A) immature
and (B) adult Shoebills, and cumulative 50% kernel density estimation for each (C) immature
and (D) adult individual, based on the GPS tracking periods indicated in Table 1. The dashed
line indicates the border of the Bangweulu Wetlands Game Management Area.





Figure 2 – Frequency of the maximum daily distance (in Km's) moved by (A) immature and (B)
adult Shoebills. The dashed red line indicates the threshold that captures 80% of movements,
used to define Shoebills' Moving Days.



Figure 3 – Boxplots of the mean maximum daily distance per month, for individual immature and adult Shoebills, between 2011 and 2018. Data is organised to start at the beginning of the breeding season (May). The boxes represent the 25th, 50th and 75th percentiles of the mean maximum daily distance. Whiskers the 1.5 times the value of inter-quantile range, with values outside this range plotted as black dots. The dashed line above the plot indicates the dry season (May to October) and the wet season (November to April). The shaded area highlights the period between October and December, with an increase of adult mean maximum daily distances.



783 Figure 4 – Diagram describing the analysed spatial and temporal relationships of Shoebill movements. The plots show the predicted mean daily NDWI values (solid line) and 95% 784 confidence interval (shaded areas) for (A) adults and (B) immatures of 'Areas' before and after 785 Shoebill abandonment, and for (C) adults and (D) immatures before Shoebill abandonment and 786 after Shoebill arrival in a new 'Area'. The boxplots display the observed values of daily mean 787 NDWI, with the boxes representing the 25th, 50th and 75th percentiles and the whiskers the 1.5 788 times the value of inter-quantile range. Values outside this range are plotted as grey dots. 789 790 Brown colours indicate adult data and blue colours indicate immature data.





Figure 5 – A) Location of the Bangweulu Wetlands Game Management Area (in red) within
Africa (light grey) and Zambia (dark grey); B) Changes in NDWI between the start of October
(week between 30/09/2015 and 07/10/2015), just prior to the start of the rainy season and
C) the end of December (week between 19/12/2015 and 26/12/2015), which is during the
rainy season. The red dashed line indicates the border of the Bangweulu Wetlands Game
Management Area.