Effects of blood parasite infections on spatiotemporal migration patterns and activity budgets in a long-distance migratory passerine

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Abstract

How blood parasite infections influence the migration of hosts remains a lively debated issue as past studies found negative, positive or no response to infections. This particularly applies to small birds, for which monitoring of detailed migration behaviour ovea whole annual cycle has been technically unachievable so far.

Here, we investigate how bird migration is influenced by parasite infections. To this end, we tracked great reed warblers (*Acrocephalus arundinaceus*) with multi-sensor loggers, characterized general migration patterns as well as detailed flight bout durations, resting times and flight heights and related these to the genus and intensity of their avian haemosporidian infections.

We found migration distances to be shorter and the onset of autumn migration to be delayed with increasing intensity of blood parasite infection, in particular for birds with *Plasmodium* and mixed-genus infections. Additionally, the durations of migratory flight bout were prolonged for infected compared to uninfected birds. But since severely infected birds and particularly birds with mixed genus infections had shorter resting times, initial delays seemed to be compensated for and the timing in other periods of the annual cycle was not compromised by infection.

Overall, our multi-sensor logger approach revealed that avian blood parasites have mostly subtle effects on migratory performance and that effects can occur in specific periods of the year only.

Keywords: activity, biologging, bird migration, flight height, great reed warbler, *Haemoproteus*, migration timing, parasites, *Plasmodium*, resting

Introduction

Infection-related deviations in host migration performance have already been described for a variety of parasite and host taxa: Monarch butterflies migrated slower when infected by ectoparasites (Bradley & Altizer, 2005), reindeer herds with more ectoparasites performed shorter migrations (Folstad, Nilssen, Halvorsen, & Andersen, 1991) and the timing of arrival at the breeding site was delayed for blood parasite infected compared to uninfected passerines (Rätti, Dufva, & Alatalo, 1993). A recent meta-analysis suggested movement capacity and phenology of migratory animals to be moderately affected by the status and intensity of infections, while body condition and survival were only weakly associated with infection (Risely, Klaassen, & Hoye, 2018). Nevertheless, the magnitude of parasite effects on migratory hosts and, particularly, how effects depend on the taxon and severity of infection, remain unresolved.

In migratory birds, parasites might affect host migration via several mechanistic pathways: Parasites could (i) directly reduce physiological performance and thereby lead to slower flight. Early seminal works indicated that *Plasmodium* infections decreased the oxygen consumption rates of experimentally infected canaries during nocturnal rest (Hayworth, Riper, & Weathers, 1987) and naturally infected lizards during exercise (Schall, Bennett, & Putnam, 1982). In contrast, the metabolic rates of a migratory passerine were not significantly influenced by infection neither during resting nor during exercise both for natural haemosporidian infections and experimental *Plasmodium* infections (Hahn et al., 2018).

Alternatively, parasites could (ii) indirectly impair host migration e.g. by slowing down fuelling, prolonging stopovers or increasing the need for resting after endurance flights. For instance, benign avian influenza infections lowered the feeding rates and prolonged stop-over durations in

migratory swans (van Gils et al., 2007) and infections with multiple intestinal parasites were correlated with delayed spring migration timing in passerines (López, Muñoz, Soriguer, & Figuerola, 2013).

Yet, regardless of the mechanisms behind parasite-effects on host migration, such changes are thought to cascade through to breeding, with delayed arrivals and late onset of breeding lowering individual fitness (Kokko, 199). However, so far, the insight into infection-related changes of host migration had been restricted to i) basic migration parameters measurable by standard methods (e.g. geolocation, telemetry and satellite tracking) and ii) snapshots of certain periods of the annual cycle. A full-migration approach relating detailed migration behaviour to individual infection parameters is still missing for wild birds.

Therefore, we used multisensor loggers to describe detailed migration patterns of great reed warblers (*Acrocephalus arundinaceus*) and related these patterns to the hosts' blood parasite infections (genera *Plasmodium* and *Haemoproteus* within Haemosporida). Multisensor loggers do not only enable geolocation by recording light, but also allow compiling individual activity and behaviour by recording accelero- and barometric data (Liechti et al., 2018). The recorded patterns were then related to the parasite genus and the individual intensities of avian **haemosporidian** parasite infections assessed by real-time quantitative PCR. Specifically, we related these infection parameters measured at the deployment and retrieval of the loggers to several key migration traits: distance, duration and speed of migration, the timing of autumn and spring migration, the duration of migratory flight bouts and resting periods, and flight height. If infections affect migration throughout on a broad scale, we expect infected individuals to migrate shorter distances or to take longer for the same distance. If migration distance would vary for infections with different parasite genera, this could both signify a parasite effect on migration performance and a differential probability of getting infected with different parasites in various regions of the non-breeding range. Depending on how infections affects migration timing, we may expect the following outcomes: If blood parasites hamper the preparation for migration, we expect that infected individuals to depart with delays. If, in contrast, blood parasites affect the progression of migration, we rather expect arrivals to be delayed. By using multisensor loggers, we also expect insights into the mechanisms behind parasite on migration patterns on a finer scale: If blood parasites impair flight, we expect infected birds to fly slower and their migratory flight bouts to be shorter (if fuel is limiting) or longer (if infected birds compensate for slower flight). If, in turn, blood parasites impair fuelling rates or increase energy expenditure, infected individuals are expected to need longer resting times. Finally, if partial oxygen pressures experienced aloft are limiting for migrating birds, we expect infected individuals to fly at lower altitudes. All these potential effects are expected to increase with growing intensity of infection and are to some degree expected to differ between parasite genera, as Haemoproteus and *Plasmodium* infections are known to differ in their average pathogenicity and co-infections with several genera are known to be most virulent (Valkiūnas, 2005).

Material & Methods

Study species and field sites

We investigated great reed warblers (*Acrocephalus arundinaceus*) breeding at three study sites in Bulgaria (BG, Kalimok Biological Station, 44.00°N 26.45°E), in the Czech Republic (CZ, Mutěnice, 48.90°N, 17.05°E) and in western Russia (RU, Rybachy, 55.15°N, 20.85°E). Great reed warblers breed in reed beds along fresh or brackish standing water and spend the non-breeding season in sub-Saharan Africa (Koleček et al., 2016 and Figure 1a). They often harbour haemosporidian parasites – a widespread group of blood parasites which are transmitted by dipteran vectors – and within the two genera *Plasmodium* and *Haemoproteus* 29 genetic lineages have been recorded for great reed warblers (MalAvi database; Bensch, Hellgren, & Pérez-Tris, 2009; accessed on the 13.06.2019). The two parasite genera and their lineages can greatly vary in their virulence. While many infections are benign, some infected hosts can show fatigue and lose appetite during the acute phase or rarely even die (Lapointe, Atkinson, & Samuel, 2012). Most avian blood parasites infections become chronic and the minor effects of chronic infections are known to accumulate, e.g. resulting in reduced life-time reproductive success in great reed warblers (Asghar, Hasselquist, & Bensch, 2011).

Additionally, we found the maximum, but not the mean, resting durations to be reduced by infection, suggesting that, regardless of infection status, migrating birds need a certain minimum recovery period between consecutive migratory flight bouts. But extensive stop-over periods can be shortened to compensate for delays – may they be caused by infections or other factors. This also implies that focussing on a single migration variable (e.g. the arrival at a stopover or breeding site) may be insufficient, as parasite effects on timing in one phase of the annual cycle can be compensated with adjustments in other phases. Whether these adjustments come with the drawback of being more exposed to predators due to compensatory feeding activity or of continuing migration in worse condition as shorter resting times cannot be compensated by extended feeding, cannot be tested with our data set.

In conclusion, using multi-sensor loggers enabled to describe detailed patterns in individual migration behaviours and how infections with parasites affect these. We found that chronic

infections with avian blood parasites have diverse, but relatively weak effects on the migration performance of great reed warblers. Our results also indicate that birds can compensate some effects of parasites arising in certain periods of the annual cycle, which thus might not be detectable anymore in subsequent periods. This indicates that hosts can cope with a broad range of chronic infection intensities, maintaining most of their migratory capacity and thus also their potential for spreading parasites (Bauer & Hoye, 2014).

Authors' contributions

TE, SH and SBa developed the conceptual framework of this study. DK, PP and PZ organized the field work and took care of permissions for their study sites. TE, DK, PP and SH carried out the field work. TE and SBe planned and conducted the lab work. TE and SBa analysed the data and drafted the manuscript. All co-authors contrbuted to writing of the manuscript, approved the final version and declared not to have any competing interests.

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Data accessibility statement

The datasets analysed and generated for this study are available on Zenodo (raw logger data:

10.5281/zenodo.4017739; migration variables and infection data: 10.5281/zenodo.4022516).

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Figures



Figure 1: The breeding sites and site-specific non-breeding ranges of the tracked great reed warblers (a) as well as the individual genus and intensity of haemosporidian infections prior to and after the tracked migrations (b). Birds breeding in Bulgaria (closed square), Czech Republic (closed circle) and Russia (closed triangle) spent the non-breeding period in sub-Saharan Africa (open symbols = site-specific median, \pm 25-75% quantiles of the individual non-breeding sites). The haemosporidian infections were generally more intense after than prior to the tracked migrations, and *Plasmodium* infections (blue) tended to be less intense than *Haemoproteus* infections (red) and mixed-genus infections (purple). Note that for some samples parasite genus is unknown (grey) and some points are unconnected as the pre-migration blood sample was missing (n = 3).



Figure 2: Effects of parasite infection on basic parameters of great reed warblers' migration. While the total migration distance (a) and the total migration duration (b) of infected great reed warblers were significantly related to the intensity, but not the genus (open symbols: Plasmodium = blue, Haemoproteus = red, mixed-genus = purple) of their haemosporidian infections, the resulting average migration speed (c) was statistically unrelated to both infection parameters.



infection intensity [log-transformed]

Figure 3: The timing of departures (a+c) and arrivals (b+d) of great reed warblers in autumn (a+b) and spring (c+d) as dependent on the haemosporidian infection. Only the departure for autumn migration was significantly related to genus (LM fit ± CrI; closed symbols: Plasmodium = blue, Haemoproteus = red, mixed-genus = purple) and intensity of infections. All other timing events were statistically unrelated to infection parameters (open symbols).



infection intensity [log-transformed]

Figure 4: The interrelations between the haemosporidian infections of tagged great reed warbler with the mean (a) and the maximum (b) of individual flight bout durations as well as the median (c) and the maximum (d) individual resting durations. While mean flight times were merely related to infection intensity (LME fit ± CrI; open symbols: Plasmodium = blue, Haemoproteus = red, mixed-genus = purple), the maximal resting times were significantly related to both the intensity and the genus of the hosts' haemosporidian infections (LME fit ± CrI; closed symbols).

Appendix

Table S1: Overview on logger data and type of blood samples available from the tracked birds - sorted by logger ID.

general information			light data		multi-sensor data		blood smear		blood in SET		
logger id	ring id	study site	deployment year	autumn	spring	autumn	spring	before migration	after migration	before migration	after migration
14AZ	1-545764	bg	2015	1	1	1	1	1	0	0	0
14ES	1-545798	bg	2015	1	1	1	1	1	1	1	1
14FF	1-545865	bg	2015	1	1	1	1	1	1	1	1
14GD	1-545690	bg	2015	1	1	1	1	1	1	1	1
14GF	1-545842	bg	2015	1	1	1	1	1	1	1	1
14GG	1-545679	bg	2015	1	1	1	1	1	1	1	1
14GH	1-545683	bg	2015	1	1	1	1	1	1	1	1
14GS	1-545763	bg	2015	1	1	1	1	1	1	1	1
14GW	1-545859	bg	2015	0	0	1	1	1	1	1	1
14HA	1-545681	bg	2015	1	1	1	1	1	1	1	1
14HB	1-545703	bg	2015	1	0	1	0	1	1	1	1
14HC	1-545680	bg	2015	1	1	1	1	1	1	1	1
14HT	1-545817	bg	2015	1	0	1	0	1	1	1	1
14SD	za35988	CZ	2016	1	1	1	1	1	1	1	1
14SF	za60716	CZ	2016	1	1	1	1	1	1	1	1
14TC	za48338	CZ	2016	1	1	1	1	1	1	1	1
14TF	za32371	CZ	2016	1	0	1	0	1	1	1	1
14TS	1-545764	bg	2016	1	1	1	1	0	1	0	1
14TZ	1-545971	bg	2016	1	1	1	1	1	1	1	1
16BR	1-555140	bg	2016	1	1	1	1	1	1	1	1
16EA	fs27400	ru	2016	1	0	1	0	0	1	0	1
16ER	fs27378	ru	2016	1	1	1	1	0	1	0	1
16ET	fs27417	ru	2016	1	0	1	0	0	1	0	1
16EY	1-545791	bg	2016	1	1	1	1	1	1	1	1
16FB	1-545785	bg	2016	1	1	1	1	1	1	1	1
16FI	1-545777	bg	2016	1	1	1	1	1	1	1	1
16FT	1-545683	bg	2016	1	0	1	0	1	1	1	1
16GA	za59713	CZ	2016	1	1	1	1	1	1	1	1

16HB	1-545761	bg	2016	1	1	1	1	1	1	1	1
16HL	1-545679	bg	2016	1	1	1	1	1	1	1	1
16HM	1-555134	bg	2016	1	1	1	1	1	1	1	1
18HA	ZA76610	CZ	2017	1	0	1	0	1	1	1	1
18HY	ZA76621	CZ	2017	1	1	1	1	1	1	1	1
18IC	ZA48339	CZ	2017	1	1	1	1	1	1	1	1
18IJ	ZA69547	CZ	2017	1	1	1	1	1	1	1	1
18IQ	ZA76620	CZ	2017	1	0	1	0	1	1	1	1
18KA	ZA69517	CZ	2017	1	1	1	1	1	1	1	1
18LD	ZA44432	CZ	2017	1	1	1	1	1	1	1	1
18LX	ZA45656	CZ	2017	1	1	1	1	1	1	1	1
18LZ	ZA62192	CZ	2017	1	1	1	1	1	1	1	1

Table S2: The variables used in the models (also see Table S3). We list the abbreviations along with a full description, the units/levels and an annotation whether a variable was used as a response variable (RV), an explanatory variable (EV) or a random factor (RF).

Variable	Description	Unit/Levels	Used as?
am.start ¹	departure for autumn migration	day of the year	RV
$am.end^{\scriptscriptstyle 1}$	arrival from autumn migration	day of the year	RV
sm.star ¹	departure for spring migration	day of the year	RV
$sm.end^1$	arrival from spring migration	day of the year	RV
dist ¹	total migration distance	kilometres	RV
migdur ¹	total migration duration	days	RV
speed ¹	total migration speed	kilometers/day	RV
durFB ²	flight bout duration	hours	RV
durSO ²	duration of resting period	hours	RV
FH ²	flight height	meters a.s.l.	RV
int ^{3,4}	intensity of infection	relative qPCR intensity	EV
season	migration season	am = autumn sm = spring	EV
numSex	numerical sex of the host	1 = male 0 = female 0.5 = NA	EV
indi	individual geolocator id	40 different IDs	RF

¹ Unless stated otherwise, we centred this variable within study site by subtracting the site-specific mean from every individual value (indicated by the suffix '.pc').

² For this variable we calculated the mean, median, max and/or sum per individuum to be used in the models.

³ This variable was used as unitless log-transformed values.

⁴ We used 'int1' when we involved the intensity from the sample before to the tracked migration, 'int2' for the sample after or 'int' when we used either of the two samples.

Table S3: List of model formulas and parameter estimates from models with (A) parasite genus as the only focal explanatory variable, (B) parasite genus and infection intensity as two focal explanatory variables and (C) infection intensity as the only focal explanatory variable (from left to right), and the models with (1) basic migration parameters, (2) migration timing parameters, (3) flight times, (4) resting times and (5) flight heights as response variables (from top to the bottom). Significant estimates and the according variables are shown in bold. The suffix '.pc' indicates that a variable has been centred within each site by subtracting the site-specific mean from the individual values. See Table S2 for a glossary of the variables used in the models.

(1) Explor depreting parameters (2) Explor depreting parameters (3) Explor depreting parameters (3) Explor depreting parameters (4)	(A) GENUS ONLY	(B) LOGST(INT) + GENUS	(C) LOGST(INT) ONLY				
Initial product a status + status	(1) Basic migration parameters						
Project proc. space + saxe + sax + (1 1dd) Project (- logitht + space + saxe + sax + (1 1dd)) Project (- logitht + space + saxe + sax + (1 1dd)) Project (- logitht + space + saxe + sax + (1 1dd)) Project (- logitht + space + saxe + sax + (1 1dd)) Project (- logitht + space + saxe + sax + (1 1dd)) Project (- logitht + space +	dist.pc ~ genus + season + Sex + (1 indi) Fixed effects: Estimate Std. Error genus - 34,00 201.62 42.64 -34,00 201.62 42.64 -34,00	dist.pc ~ logstint + genus + season + Sex + (1 indi) Fixed effects: Estimate Std. Error df t value Pr(> t) (Intercept) -429.20 358.41 31.65 -1.198 0.240 logstint - 163.38 76.62 19.26 -2.103 0.240 - genush 283.03 242.50 31.64 1.410 0.168 - <td< td=""><td>dist.pc ~ logstint + season + Sex + (1 indi) Fixed effects: Estimate Std. Error df t value Pr(S t]) (Intercept) 17.17 158.51 47.27 0.108 0.218 logstint - 60.11 39.82 48.30 -1.509 0.138 seasons 97.22 114.88 26.37 0.846 0.409 Sex -240.66 170.52 56.74 -1.411 0.167</td></td<>	dist.pc ~ logstint + season + Sex + (1 indi) Fixed effects: Estimate Std. Error df t value Pr(S t]) (Intercept) 17.17 158.51 47.27 0.108 0.218 logstint - 60.11 39.82 48.30 -1.509 0.138 seasons 97.22 114.88 26.37 0.846 0.409 Sex -240.66 170.52 56.74 -1.411 0.167				
Superior	migdur.pc ~ genus + season + Sex + (1 indi) Fixed effects: Estimate Std. Error df t value Pr(> t) (Intercept) 2.750 5.628 53.000 0.489 0.627 genus P - 7.347 6.065 53.000 -1.212 0.231 genus P - 7.347 6.065 53.000 -1.212 0.231 genus P - 7.347 6.065 53.000 -1.214 0.231 genus P - 7.347 6.065 53.000 -1.426 0.141 seasons 2.834 4.819 53.000 0.361 0.719 Sex 1.932 5.347 53.000 0.361 0.719	migdur.pc ~ logstint + genus + season + Sex + (1 indi) Fixed effects: (Intercept) - 10.4884 13.2000 52.0000 - 0.795 0.430 logstint - 3.3509 3.0236 52.0000 - 11.08 0.2278 genush - 1.3344 12.4921 52.0000 - 0.1270 0.8211 genush - 0.1797 0.1328 0.226 52.0000 - 0.1270 0.8211 genush - 3.796 4.889 52.0000 -0.777 0.440 sex - 2.0243 5.3358 52.0000 0.777 0.440	migdur.pc ~ logstint + season + sex + (1 indi) Fixed effects: Estimate Std. Error df t value Pr(> t) (Intercept) -9.242 4.829 65.000 -1.914 0.0601 logstint -2.768 1.337 65.000 -2.070 0.0424 seasonsm 1.744 4.551 65.000 0.410 0.6829 sex 4.446 4.532 65.000 0.981 0.3302				
(2) Nigration tising m.start.pc - genus1 + num5ex as.start.pc - logstint1 + genus1 + num5ex Coefficients: Coe	speed.pc ~ genus + season + Sex + (1 indi) Fixed effects: Estimate Std. Error (Intercept) -20.000 genusH 28.576 genusH 70.066 yeasonsm -9.820 Seasonsm -9.820 Sex -3.886 31.103 27.587 -0.125 0.052 0.0569 -0.055 0.0567 0.0557 0.0567 0.0557 0.0	speed.pc ~ logstint + genus + season + Sex + (1 indi) Fixed effects: 61 + 302 Cintercept) 58,9324 63,1837 Genus 28,2324 63,1837 Genus 28,2324 63,1837 Genus 29,8235 63,1837 Genus -43,6364 60,2706 Genus -43,6364 60,2706 Genus -43,6364 60,5153 Genus -43,6364 60,5153 Genus -43,6364 60,5153 Genus -43,6364 63,1513 Genus -43,6364 60,2706 Seasons -17,4979 22,6579 Seas -5,0208 31,0972 Sea -5,0208 31,0972	speed.pc ~ logstint + season + Sex + (1 indi) Fixed effects: Estimate Std. Error df t value Pr(> t) (Intercept) 28.398 24.595 55.000 1.157 0.2583 seasonsm -11.601 22.552 55.000 -0.698 0.6543 seasonsm -11.349 24.877 55.000 0.054 0.9569 Sex				
matrix pc - genust + numsex an.start.pc - logstint + numsex m.start.pc - logstint + numsex m.start.pc - logstint + numsex Coefficients: Contervol 5:1707 5:1516 0:0788 * Coefficients: <	(2) Migration timing						
am.end.pc - genus1 + numSex am.end.pc - logstint1 + genus1 + numSex am.end.pc - logstint1 + genus1 + numSex am.end.pc - logstint1 + numSex Coefficients: Coefficients: <td< td=""><td>am.start.pc ~ genus1 + numSex Coefficients: Estimate Std. Error t value Pr(S t) (Intercept) -2.700 4.390 -0.615 genus1 = 16.773 5.475 genus1 = 10.088 5.507 genus1 = 10.77 7.463 genus1 = 19.271 7.463 genus1 = 10.271 9.582 genus1 = 10.271 9.582 <!--</td--><td>am.start.pc ~ logstintl + genusl + numSex Coefficients: Estimate Std. Error t value Pr(> t) (Intercept) 4.093 12.250 0.334 0.7412 logstintl 1.719 2.888 0.595 0.5573 genuslP 13.821 7.441 1.857 0.0756 . genuslH 4.447 11.000 0.040 0.6896 genuslHP 13.437 12.380 1.085 0.2885 numsex ~ 8.042 4.640 -1.733 0.0959 .</td><td>am.start.pc ~ logstint1 + numSex Coefficients: (Intercept) 13.574 4.308 3.151 0.00339 ** logstint1 3.638 1.271 2.663 0.00713 ** numSex -8.040 4.041 -1.990 0.05471 .</td></td></td<>	am.start.pc ~ genus1 + numSex Coefficients: Estimate Std. Error t value Pr(S t) (Intercept) -2.700 4.390 -0.615 genus1 = 16.773 5.475 genus1 = 10.088 5.507 genus1 = 10.77 7.463 genus1 = 19.271 7.463 genus1 = 10.271 9.582 genus1 = 10.271 9.582 </td <td>am.start.pc ~ logstintl + genusl + numSex Coefficients: Estimate Std. Error t value Pr(> t) (Intercept) 4.093 12.250 0.334 0.7412 logstintl 1.719 2.888 0.595 0.5573 genuslP 13.821 7.441 1.857 0.0756 . genuslH 4.447 11.000 0.040 0.6896 genuslHP 13.437 12.380 1.085 0.2885 numsex ~ 8.042 4.640 -1.733 0.0959 .</td> <td>am.start.pc ~ logstint1 + numSex Coefficients: (Intercept) 13.574 4.308 3.151 0.00339 ** logstint1 3.638 1.271 2.663 0.00713 ** numSex -8.040 4.041 -1.990 0.05471 .</td>	am.start.pc ~ logstintl + genusl + numSex Coefficients: Estimate Std. Error t value Pr(> t) (Intercept) 4.093 12.250 0.334 0.7412 logstintl 1.719 2.888 0.595 0.5573 genuslP 13.821 7.441 1.857 0.0756 . genuslH 4.447 11.000 0.040 0.6896 genuslHP 13.437 12.380 1.085 0.2885 numsex ~ 8.042 4.640 -1.733 0.0959 .	am.start.pc ~ logstint1 + numSex Coefficients: (Intercept) 13.574 4.308 3.151 0.00339 ** logstint1 3.638 1.271 2.663 0.00713 ** numSex -8.040 4.041 -1.990 0.05471 .				
sm.start.pc - genus2 + numSex sm.start.pc - logstint2 + genus2 + numSex sm.start.pc - logstint2 + genus2 + numSex sm.start.pc - logstint2 + numSex Coefficients: Estimate Std. Error t value Pr(>[t]) (Intercept) - 0.141603 8.245046 - 0.017 0.986 Sm.start.pc - logstint2 + genus2 + numSex Coefficients: genus2P 8.245103 8.34878 0.965 0.9341 logstint2 0.1460 3.8174 0.116 0.908 regression 0.928 genus2P 8.245103 8.34878 0.965 0.344 genus2P - 8.5990 17.1090 - 0.755 0.457 logstint2 - 0.3460 - 0.017 0.986 genus2P 8.245103 8.54878 0.960 0.344 genus2P - 8.5990 17.1090 - 0.755 0.457 logstint2 - 0.988 genus2P 8.245103 8.54878 0.960 0.900 genus2P - 8.5990 17.1090 - 0.755 0.457 logstint2 - numSex coefficients: sm.end.pc ~ genus2 + numSex Sm.end.pc ~ logstint2 + genus2 + numSex sm.end.pc ~ logstint2 + genus2 + numSex coefficients: (Intercept) - 7.46937 7.469307 - 0.985 0.0467 logstint2 - 1.386 0.176 genus2P - 7.840 6.117 - 1.282 0.0212 genus2P - 7.46937 3.0196 0.6271 logstint2 - 1.985 genus2P - 6.380 1.1000 genus2P - 1.640 8.373 0.196 0.6271 logstint2 - 2.307 1.271 - 1.816 0.6797 genus2P - 6.3857 5.929 - 1.448 0.160 genus2P - 1.285 0.5	am.end.pc ~ genus1 + numSex Coefficients: (Intercept) -2.0701 4.3200 -0.479 0.636 genus1 3.3621 5.3878 0.624 0.538 genus1 4.0383 5.4190 0.745 0.463 genus1 3.8818 7.3441 0.529 0.602 numSex -0.4723 4.4721 -0.106 0.917	am.end.pc ~ logstint1 + genus1 + numSex Coefficients: [Intercept] -4,9882 12,1262 -0.411 0.684 logstint1 -0.7384 2.8590 -0.258 0.798 genus1P 4.6302 7.3663 0.629 0.536 genus1H 6.4619 10.8890 0.593 0.558 genus1H 6.3681 12.2556 0.521 0.607 numSex -0.3248 4.5936 -0.071 0.944	am.end.pc ~ logstint1 + numSex Coefficients: (Intercept) 2.4280 3.9642 0.612 0.544 logstint1 0.9252 1.1691 0.791 0.434 numSex -0.4663 3.7178 -0.125 0.901				
sm.end.pc - genus2 + numSex sm.end.pc - logstint2 + genus2 + numSex sm.end.pc - logstint2 + genus2 + numSex sm.end.pc - logstint2 + numSex Coefficients: Estimate Std. Error t value Pr(> t) Coefficients: Coefficients: Coefficients: genus2 + numSex -6.382 0.493 0.6213 0.6213 Coefficients: Coefficients: genus2 + numSex -6.382 6.791 -0.940 0.357 genus2 + numSex 2.439 -0.430 0.6213 Coefficients:	sm.start.pc ~ genus2 + numSex Coefficients: Estimate Std. Error t value Pr(> t) (Intercept) -0.141603 8.245946 -0.017 0.986 genus2h - 7.486973 7.699201 -0.972 0.341 genus2h 8.249150 8.548787 0.965 0.344 genus2h - 0.058269 9.687533 0.006 0.995 numSex -0.001944 7.463227 0.000 1.000	sm.start.pc ~ logstint2 + genus2 + numSex Coefficients: Estimate Std. Error t value Pr(> t) (Intercept) 1.5097 16.5168 0.091 0.928 logstint2 0.4460 3.8374 0.116 0.908 genus2P - 8.3990 11.1090 -0.756 0.457 genus2H -1.5728 17.1764 -0.092 0.928 numSex 0.0651 7.6420 0.008 0.933	sm.start.pc ~ logstint2 + numSex Coefficients: (Intercept) 7.4491 5.4190 1.375 0.180 logstint2 0.8006 1.8205 0.443 0.661 numSex -9.1303 6.5896 -1.386 0.176				
(3) Flight times durFB.mean.pc ~ genus + season + Sex + (1 indi) durFB.mean.pc ~ logstint + genus + season + Sex + (1 indi) durFB.mean.pc ~ logstint + season + Sex + (1 indi) Fixed effects: Fixed effects: Fixed effects: Fixed effects:	sm.end.pc ~ genus2 + num5ex Coefficients: Estimate Std. Error t value Pr(> t) (Intercept) 11.809 6.550 1.803 0.084 . genus2P - 7.840 6.117 -1.282 0.212 genus2H - 6.382 6.791 -0.940 0.357 genus2H - 8.740 7.156 0.267 num5ex - 8.587 5.929 -1.448 0.160	sm.end.pc ~ logstint2 + genus2 + numsex Coefficients: Estimate Std. Error t value Pr(> t) (Intercept) -5.355 12.449 -0.430 0.671 logstint2 -4.635 2.882 -1.603 0.123 genus2P 1.640 8.373 0.196 0.846 genus2P 1.12.850 0.863 0.387 genus2P 1.540 1.12.855 0.863 0.367 genus2P - 3.640 8.373 0.196 0.846 genus2P - 3.640 8.373 0.196 0.367 genus2P - 3.640 8.373 0.196 0.367 genus2P - 3.640 8.373 0.196 0.367 genus2P - 3.663 0.367 genus2P - 3.664 0.372 genus2P - 3.664 0.367 genus2P - 3.663 0.367 genus2P - 3.664 0.372	<pre>sm.end.pc ~ logstint2 + numSex Coefficients: stimate std. Error t value Pr(> t) (Intercept) 2.075 3.782 0.549 0.5875 (Intercept) 2.210 1.281 - 1.816 0.0797 . numSex -8.685 4.599 -1.888 0.0690 .</pre>				
durFB.mean.pc ~ genus + season + Sex + (1 indi) durFB.mean.pc ~ logstint + genus + season + Sex + (1 indi) durFB.mean.pc ~ logstint + season + Sex + (1 indi) Fixed effects: Fixed effects: Fixed effects:	(3) Flight times						
	durFB.mean.pc ~ genus + season + Sex + (1 indi) Fixed effects:	<pre>durFB.mean.pc ~ logstint + genus + season + Sex + (1 indi) Fixed effects:</pre>	durFB.mean.pc ~ logstint + season + Sex + (1 indi) Fixed effects:				

$ \begin{array}{c} \mbox{Estimate Std. Error} & \mbox{df} t value \mbox{Pr}(> t) \\ (Intercept) & 0.02466 & 0.37246 41.24017 & 0.0666 & 0.948 \\ \mbox{genusP} & 0.18642 & 0.39433 47.90401 & 0.473 & 0.639 \\ \mbox{genusH} & 0.19661 & 0.41069 & 50.58490 & 0.479 & 0.634 \\ \mbox{genusHP} & 0.08910 & 0.52795 & 44.05214 & 0.169 & 0.867 \\ \mbox{seasonsm} & -0.01743 & 0.27265 & 30.27173 & -0.0644 & 0.949 \\ \mbox{Sex} & -0.24174 & 0.36617 & 31.83505 & -0.660 & 0.514 \\ \end{array} $		Estimate Std. Error df t value pr(>(1t) (Intercept) 0.57640 0.035342 53.21645 1.631 0.109 logstint 0.13786 0.09454 64.87535 1.458 0.150 seasons -0.10354 0.26659 86.0886 0.700 Sex -0.39837 0.34961 35.73147 -1.139 0.262
durFB.max.pc ~ genus + season + Sex + (1 indi) Fixed effects: Estimate Std. Error df t value Pr(> t) (Intercept) - 1.944 1.2410 42.9742 - 1.117 0.042 genush 0.9023 1.9127 44.1383 0.471 0.640 genush - 1.6268 2.4671 36.7901 - 0.659 0.514 seasons 0.6623 1.2685 19.7022 0.522 0.607 Sex 1.6716 1.7133 21.3003 0.976 0.340	durFB.max.pc ~ logstint + genus + season + Sex + (1 indi) Fixed effects: Estimate Std. Error of t value Pr(S t] (Intercept) 1.3163 5.853 33.8132 0.345 0.322 genus - 1.3539 2.4088 31.8655 -0.552 0.577 genus - 2.0712 3.6407 51.2685 -0.569 0.572 genus - 4.5628 3.9306 1.2661 -1.161 0.251 seasons 0.4340 1.2904 13.9101 0.336 0.740 sex	durF8.max.pc ~ logstint + season + Sex + (1 indi) Fixed effects: Estimate Std. Error df t value Pr(> t)) (Intercept) -0.280 1.6733 63.1853 -0.136 0.482 seasonsm -0.3032 1.0928 35.1853 -0.136 0.482 seasonsm -0.3032 1.0928 35.097 -0.277 0.783 sex
(4) Resting times		
dµrS0.mean.pC ~ genus + SeeSon + Sex + (1 indi) Fixed effects: (Intercept) -1.1759 genusP 0.8055 1.0702 45.6124 0.753 0.4555 genusP 0.8055 1.0702 45.6124 0.753 0.4555 genusP 0.134 1.1170 48.6977 -0.102 0.9196 genusP -0.134 1.1170 48.6977 -0.102 0.9196 genusP 0.20587 0.27587 37.0921 2.713 0.0114 * Sex -1.3849 0.9839 27.9010 -1.408 0.1703	durs0.mean.pc ~ logstint + genus + Season + Sex + (1 indi) Fixed effects: Estimate Std. Error df t value Pr(s t) (Intercept) - 2.64016 2.24882 51.06715 - 1.174 0.24583 logstint - 0.36713 0.50526 43.14071 - 0.727 0.47121 genus 1 1.21676 2.14202 51.04039 0.5686 0.57268 genus 0.08963 2.30678 51.19049 0.039 0.969616 seasons 2.15952 0.77110 26.99632 2.801 0.09931 ** sex - 1.35880 0.99371 28.14858 - 1.367 0.18232	durs0.mean.pc ~ logstint + season + Sex + (1 indi) Fixed effects: Estimate Std. Error ff t value Pr(> t) (Intercept) -1.4658 0.8466 52.6913 -1.732 0.0891 - logstint -0.1456 0.2325 61.0186 -0.626 0.5335 seasonsm 1.9385 0.7202 35.7824 2.719 0.0309 * Sex -1.0992 0.8075 33.8499 -1.361 0.1825
durS0.max.pc ~ genus + season + Sex + (1 indi) Fixed effects: Estimate Std. Error df t value Pr(> t) (Intercept) 22.67 51.78 51.78 (enusrept) 22.67 52.67 51.78 (enusrept) 22.67 (enusrept) 22.67 (enusrept) 22.67 (enusrept) -116.58 (enusrept) -268.92 (enusrept) -268.92 (enusrept) -3.263 (enusrept) -3.263 <t< td=""><td>$\begin{array}{llllllllllllllllllllllllllllllllllll$</td><td>durs0.max.pc ~ logstint + season + Sex + (1 indi) Fixed effects: Estimate Std. Error df t value Pr(> t) (Trtercept) -131.79 49.14 52.97 -2.682 0.00947 logstint -39.24 13.38 63.57 -2.932 0.00467 seasonsm 23.23 39.58 35.68 0.0587 0.56089 Sex 63.98 47.57 34.40 1.345 0.18748</td></t<>	$ \begin{array}{llllllllllllllllllllllllllllllllllll$	durs0.max.pc ~ logstint + season + Sex + (1 indi) Fixed effects: Estimate Std. Error df t value Pr(> t) (Trtercept) -131.79 49.14 52.97 -2.682 0.00947 logstint -39.24 13.38 63.57 -2.932 0.00467 seasonsm 23.23 39.58 35.68 0.0587 0.56089 Sex 63.98 47.57 34.40 1.345 0.18748
(5) Flight height		
FH.mean.pc ~ genus + season + Sex + (1 indi) Fixed effects: Estimate Std. Error df t value Pr(> t) (Intercept) -100: 20 106: 20 51:48 -1.036 0.07 genus + 190:87 108:74 52:99 1.755 0.085 genus + 126:45 143:27 51:61 0.883 0.382 seasons -2.10 64:00 27:89 -0.038 0.974 Sex 26:08 105:28 33:16 0.248 0.806	$ \begin{array}{l} \mbox{FH.mean.pc} \sim logstint + genus + season + Sex + (1 \ \ indi) \\ \mbox{Fixed effects:} \\ \mbox{Constraints} & Std. Error & df t value Pr(S \ 1 \) \\ \mbox{Constraints} & 208.42 & 40.13 & 1.016 & 0.0769 \ . \\ \mbox{Genus} & -32.43 & 131.59 & 50.66 & -0.246 & 0.8063 \ \\ \mbox{genus} & -132.88 & 198.24 & 50.14 & -0.605 & 0.5481 \ \\ \mbox{genus} & -132.88 & 198.24 & 50.14 & -0.605 & 0.3993 \ \\ \mbox{seasons} & -24.89 & 66.48 & 26.43 & -0.374 & 0.1111 \ \\ \mbox{sex} & 20.67 & 9.9.51 & 30.65 & 0.274 & 0.7111 \ \\ \mbox{sex} & 20.67 & 9.9.51 & 30.85 & 0.207 & 0.737 \ \end{array} $	FH.mean.pc ~ logstint + season + Sex + (1 indi) Fixed effects: Estimate Std. Error df t value Pr(> t) (Intercept) 104.16 84.57 51.44 Jegstint 56.39 21.16 22.21 Seasonsm -25.94 54.03 33.43 -0.480 0.63427 Sex 40.98 87.88 35.13 0.466 0.64392
FH.max.pc ~ genus + season + Sex + (1 indi) Fixed effects: Estimate Std. Error df t value Pr(> t) (Intercept) 94.94 297.38 36.03 0.319 0.751 genus P 544.89 312.65 46.40 -1.74 0.088 - genus H -193.21 324.40 50.44 -0.596 0.554 - genus H -273.62 186.29 226.57 46.53 0.518 - seasonsm 204.22 38.23 22.25 0.160 0.874	FH.max.pc ~ logstint + genus + season + Sex + (1 indi) Fixed effects: Estimate Std. Error df t value Pr(>[t]) (Intercept) 1115.66 1095tint 236.06 9enusP -1142.93 400.39 51.31 9enusP -117.43 9enusP -117.63 9enusP -117.63 92.69.9 22.43 92.15 0.516 92.69.9 22.45 92.15 0.516 92.69.9 22.45 92.15 0.516 92.69.5 2.152 92.15 0.516 92.65.5 2.152	FH.max.pc ~ logstint + season + Sex + (1 indi) Fixed effects: (Intercept) -20.10 279.78 51.00 -0.072 0.943 logstint 13.01 70.60 62.73 0.184 0.854 seasonsm 42.11 181.96 32.71 0.231 0.818 Sex 75.81 289.30 34.21 0.262 0.795



Figure S1: A sample data set for the period 16-31 August 2015 from a great reed warbler with the multi-sensor logger ID 14AZ. The panels depict data from the different sensors (from top to the bottom): light intensity, air pressure, ambient temperature, locomotor activity and pitch (i.e. mean body position). The colours in the pressure and activity panels show behaviours categorised as 'resting' (black), 'active' (green), 'flight' (blue) and 'migratory flight' (red).