

### Flight behaviours and energy savings in adult and juvenile house martins (Delichon urbicum) foraging near their breeding colony

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martins (*Delichon urbicum*) foraging near their breeding colony
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### 11 Abstract

12 Foraging is an extremely important behaviour for birds, especially during the breeding 13 season, when they have to carry the cost of incubation and chick rearing, in addition to their 14 own energy needs. Aerial insectivores perform most of their foraging behaviours in flight, so they have evolved various adaptations to reduce energy output while increasing energy input 15 16 during this critical period. In this study, we recorded the 3D flight behaviours of 100 house 17 martins (Delichon urbicum) flying near their colony during the breeding season in Rennes, France. We give a first description of the distribution of several kinematic and biomechanical 18 19 variables (horizontal and vertical speed, rates of change in kinetic and potential energy, turning 20 radius of curvature and centripetal force), compare flapping and gliding flight, and describe 21 several strategies used by flying house martins to save energy, such as environmental energy 22 extraction (thermal soaring) and optimisation of flight speed according to wind speed and 23 direction. We also report an effect of temperature, solar radiation and humidity on the mean 24 vertical speed of gliding birds, highlighting the effect of weather on the availability of external 25 energy sources such as thermal updrafts. Finally, we compare the distribution of flight speed 26 and vertical speed between 5 juveniles identified using magnified photographs and 20 adults recorded during the same field sessions, and we show that during flapping flight, juveniles 27

exhibit higher, more variable airspeed than adults, suggesting that their flight behaviours arenot immediately fine-tuned after leaving the nest.

30

31 Keywords: energy, wind, kinematics, ontogeny

### 32 Significance statement

33 Aerial insectivores use various strategies to reduce the cost of foraging flight. Using an 34 optical tracking method, we recorded the 3D flight behaviours of house martins (Delichon 35 urbicum) flying near their colony during the breeding season. We describe the distribution of 36 several biomechanical variables and show that house martins use external energy sources such 37 as thermal updrafts and also adapt their airspeed to wind speed and direction, supporting the 38 predictions on optimal cost of transport in birds. Moreover, juveniles were also recorded, and 39 they show a greater variability in flight speed, possibly because they may not be as accurate as 40 adults in finely adjusting their speed and altitude. Our findings add to the existing literature 41 showing energy saving strategies in aerial insectivores, and also study an ontogenetical aspect 42 rarely explored.

43

### 44 Introduction

Foraging is a behaviour of crucial importance in the life cycle of birds, especially during the breeding season. During incubation, parents have to cope with various constraints and invest time and energy (Shaffer et al. 2003). When chicks hatch, parents still have to dedicate some time to warm or protect them in many species, and they additionally have to cover the food needs of an entire brood (Ydenberg 1994; Markman et al. 2002).

50 Aerial insectivores, like swifts, swallows and martins, have to fly continuously and to 51 perform flight manoeuvres while foraging (Bryant and Turner 1982; Kacelnik and Houston

52 1984). Swifts, swallows and martins feed their chicks with a food bolus constituted of tens to 53 hundreds of arthropods (Bryant and Turner 1982; Gory 2008), which avoids having to fly back 54 and forth between the nest and the foraging patches for each individual prey. During foraging, 55 maximisation of energy intake per unit time is obviously important, but energy consumed 56 during flight is considerable, and the foraging strategy must be a balance between the energy 57 output during flight and the energy intake from feeding (Rayner 1982). Thus, flying at a low 58 cost is of paramount importance for foraging aerial insectivores.

Various behavioural adaptations exist to reduce flight energy expenditures. For example, some aerial insectivores are able to extract energy from their environment during foraging. Common swifts (*Apus apus*) can glide in thermal updrafts and use wind gusts and wind gradients to save energy (de Margerie et al. 2018; Hedrick et al. 2018), while barn swallows (*Hirundo rustica*) also use wind gradients to gain potential and kinetic energy during turns (Warrick et al. 2016).

65 Additionally, wind speed and direction may also influence the flight behaviours of foraging aerial insectivores. Indeed, theory predicts that birds should adjust their airspeed when 66 67 flying upwind or downwind for energetically optimal cost of transport in the ground reference 68 frame (Pennycuick 1978). The maximum range speed of birds (the airspeed at which the 69 distance travelled for a given amount of energy consumed is maximised) is influenced by wind, 70 and birds optimising their energy expenditure per unit of distance should increase their airspeed when flying upwind and decrease it when flying downwind. This phenomenon has been 71 72 confirmed in migrating or commuting birds (Wakeling and Hodgson 1992; Hedenström et al. 73 2002; Kogure et al. 2016; Sinelschikova et al. 2019) and also in the common swift while 74 foraging on aerial insect prey (Hedrick et al. 2018), probably because of the presence of its nest 75 at a fixed ground position.

In addition to wind, other weather variables might have an effect on the flight behaviours
of aerial insectivores, such as temperature, solar radiation or humidity, because they influence
the availability and movement patterns of aerial insect prey (Lack and Owen 1955; Bryant
1973; Wainwright et al. 2017), and also the availability of external energy sources such as
thermal updrafts (Poessel et al. 2018).

Finally, very little is known about the ontogeny of foraging and energy-saving flight behaviours within an individual. Since foraging flight is a complex behaviour, it is possible to hypothesize that juvenile birds may not be as efficient as adults in all aspects immediately after fledging, as is the case in many species for various flight behaviours (see review in Ruaux et al., 2020).

86 The house martin (Delichon urbicum) is a socially monogamous, coursing insectivore 87 nesting in colonies, in which both sexes incubate and feed the chicks (Bryant 1979; 88 Whittingham and Lifjeld 1995; del Hoyo et al. 2020). They lay up to three clutches per year, 89 and each clutch is composed of one to seven (most often four to five) eggs (del Hoyo et al. 90 2020). Bryant and Westerterp (1980) studied the energy budget of breeding house martins and 91 calculated that each parent spent around 6 h per day away from the nest during incubation, and that a bird foraging at the highest observed rate in optimal conditions during this time would 92 93 gather energy only 6% in excess of its requirements, leaving little margin for other activities 94 and lower foraging rates in poorer conditions. When feeding chicks, parents spend more time 95 in flight but have to meet the energy needs of their brood in addition to their own needs. Thus, 96 breeding house martins should spend most of their time actively foraging and should mostly be 97 traveling or searching for food otherwise. In this context, studying the flight behaviours of 98 house martins near a colony during the breeding period may improve understanding of the 99 characteristics of flight during this crucial period, and identify possible means by which these 100 birds reduce their energy expenditures.

101 In the present study, we measured the 3D flight trajectories of house martins using 102 rotational stereo videography (RSV; de Margerie et al. 2015) to describe biomechanical 103 characteristics of their flight. One of our goals was to give a first description of the "flight 104 envelope" of house martins in a field study to understand how they use the aerial habitat near 105 their colony during the breeding period. We also tested some of the hypotheses related to energy 106 savings in aerial insectivores: we studied the gliding and flapping behaviours of house martins 107 to determine if they use external energy sources such as thermal currents, wind gusts and wind 108 gradients, and if weather conditions could influence these behaviours. Then, we tested if house 109 martins change their airspeed depending on wind direction to optimise their energy expenditure 110 in the ground reference frame. Finally, we investigated if juvenile house martins differ from 111 adults in some aspects of their flight behaviours.

### 112 Materials and methods

All symbols and abbreviations used in our analyses can be found in Table 1.

114 Recording site and time

House martins were recorded near a colony located in Rennes, France (Fig. 1, see also Fig. S1 for a ground view of the experimental setup). The breeding house martins are present in the colony from May to September, and the colony is composed of several tens of nests built on buildings (3 to 6-floor), surrounded by an urban landscape, with mainly roads, a wide lawn and urban gardens. The RSV device was located on a small hill to the northwest of the colony (48°07'45.55"N 1°40'42.88"W), with a panoramic view over the wide lawn and urban gardens above which the house martins were often flying.

122

Nine recording sessions took place from May to July 2021, corresponding to the time
when house martins are raising their first brood in this region of France (del Hoyo et al. 2020).

Recordings took place in the morning between 9:30 h and 12:00 h, when house martins wereregularly observed flying near the colony.

During each field session, a GILL Instruments MaxiMet GMX501 weather station (Lymington, UK) with ultrasonic anemometer was set up on a tripod to measure the approximate wind speed and direction experienced by house martins flying near the colony. We placed the anemometer at 2 m height above the ground, in the wide lawn located west to the colony to minimise proximity with any tree or building (see Fig. 1). The weather station also recorded temperature, solar radiation, humidity and atmospheric pressure. All variables were recorded at 1 Hz and were averaged over the duration of each trajectory.

134

#### 135 Rotational stereo-videography (RSV)

136 RSV is an optical tracking technique based on a set of mirrors projecting a stereo image 137 of the animal on the sensor of a single camera (de Margerie et al. 2015). The distance to the 138 animal is measured by analysing the lateral shift between animal image pairs. The rigid 139 assembly of camera and mirrors can rotate horizontally and vertically on a tripod and fluid 140 video head. While the operator rotates the device to keep the moving animal's image within the 141 sensor frame, the aiming angles are recorded by angular encoders. The geometrical combination 142 of distance and aiming angles (corrected for the position of the animal image on the sensor) 143 yields a 3D record of the animal's movement.

We used an updated RSV device (Fig. S2) with a 1 m base length between the lateral mirrors, a Manfrotto 504HD fluid head (Cassola, Italy) equipped with 17-bit digital angular encoders (Kübler Sendix F3673, Villingen-Schwenningen, Germany), recording aiming angles at 200 Hz through an Arduino Mega microcontroller (www.arduino.cc) and an Adafruit Data logging shield (New York, USA). The device was equipped with a Panasonic DC-GH5S camera (Osaka, Japan) recording 4096 × 2160-pixel frames at 60 Hz (150 Mbps H.264 compression)

150 from a  $19 \times 10$  mm sensor area. We used a Nikon AF 105 mm f/2 lens (Tokyo, Japan), 151 providing a  $5.2^{\circ}$  horizontal field of view for each side of the stereo image. To get well exposed 152 and sharp images, we used a 1/1300-1/640 s shutter speed and f/11 aperture, with ISO 1000– 153 2500, depending on available light conditions. To help tracking the fast-flying birds, the camera 154 was equipped with a Nikon DF-M1 dot sight viewfinder (Osaka, Japan).

155

#### 156 Calibration and location error

To calibrate the distance measure, which is based on the lateral offset between left and right images of the bird, we recorded six conspicuous targets (signs, street lamps, trees) located at fixed distances from the RSV device, from 16 to 410 m. The real distance to these targets was measured with a Nikon Forestry Pro hand laser rangefinder (Tokyo, Japan).

161 The random positioning error was approximately 0.04 m at 25 m, 0.10 m at 50 m and 162 0.34 m at 100 m (Fig. S3).

163

164

#### 165 Recording methods and data classification

During each field session, we recorded any house martins seen flying at 25–100 m from the RSV device (i.e. convenience sampling), and each individual was followed until it was lost by the operator. To minimize pseudoreplication, we made sure to record a different individual at the end of each recording. Despite this precaution, pseudoreplication may be present to some extent in our data since many individuals were flying back and forth between their nest and the nearby gardens. However, we assume pseudoreplication to be modest, since we recorded 100 trajectories near the colony composed of several tens of pairs.

To analyse a sufficient and comparable portion of trajectories, recordings lasting less
than 30 s were removed. Retained videos were subsampled from 60 to 10 Hz to ensure that the

175 number of frames was manageable for digitizing, yet still appropriate to describe flight176 behaviours at a fine temporal scale.

Moreover, recordings where the bird was out of frame during more than 10 consecutive frames (i.e. more than 1 s missing at 10 Hz) were also removed (with a 5.2° field of view, the operator occasionally struggled to continuously follow the bird's path with the camera). The resulting sample had 97 trajectories with a homogeneous distribution over the nine field sessions (between 9 and 12 trajectories per session).

182 During each recording, photographs were taken with a greater magnification using a second camera mounted on the RSV device (Panasonic DMC-GH4 with a Nikon AF 200 mm 183 184 f/4 lens) to have a clearer view of the plumage of each bird and to be able to identify juveniles, 185 which are recognized mainly by the white tips of their tertials (Blasco-Zumeta and Heinze 186 2014). Five juveniles could be identified during two field sessions in the first half of July, 187 consistent with the breeding phenology of house martins (del Hoyo et al. 2020). Three of these 188 juveniles had trajectories lasting less than 30 s (15, 23 and 29 s). These juveniles were added to 189 the dataset to ensure a minimal sample size for juveniles, increasing the total to 100 trajectories 190 (95 adults and 5 juveniles), with a median duration of 37.5 s and a total duration of 4512 s.

To study the link between house martins' behaviours and biomechanical variables, the flapping behaviour was labelled on each frame by direct observation of the recorded videos, as either "gliding", "flapping" or "not visible" when the bird was too far or flew in front of a very textured background (foliage). Only birds performing at least one full downstroke and upstroke cycle were categorized as flapping, because they occasionally performed short manoeuvring wing movements during gliding.

- 197 It was not possible to record data blind because our study involved focal animals in the198 field.
- 199

201 Stereo videos and angular records were processed with MATLAB r2018b (The 202 Mathworks, Natick, MA, USA). To digitize the bird's locations in each video frame, the pixel 203 at the centroid of the bird's body in the left half of the frame was selected as the left point of 204 interest (POI), either manually or with the help of semi-automatic tracking (DLTdv version 8a; 205 Hedrick 2008). Then, automated normalized cross-correlation between a  $31 \times 31$ -pixel area 206 around the left POI and the right half of the image was used to find the corresponding right 207 POI. Automated matching of right POI was sometimes misled by variable backgrounds (sky, 208 foliage, buildings), and thus was visually checked and manually corrected when needed. The 209 bird's distance from the RSV device was then computed based on the calibration reference.

210 RSV tracking yields spherical coordinates of the bird for each video frame (i.e. azimuth 211 angle, elevation angle and distance from the device;  $\Theta$ ,  $\Phi$  and P respectively). Raw coordinate 212 series contain noise, due to (i) theoretical positioning uncertainty (increasing with P<sup>2</sup>, see de 213 Margerie et al. 2015) and (ii) POI random positioning error in stereo images, which was 214 exacerbated by variable image backgrounds. Consequently, we smoothed the raw spherical 215 coordinate series using quintic splines (which allow non-zero acceleration at the sequence 216 ends), with an error tolerance based on the sum of (i) the per-point theoretical positioning 217 uncertainty and (ii) the amplitude of high-frequency signal present in the coordinate series (as 218 measured with 3 Hz high-pass filtering). These splines also interpolated short ( $\leq 10$  frames) 219 track bouts where the bird was out of frame. Smoothed spherical coordinates were then 220 converted to cartesian coordinates (X, Y, Z) without additional smoothing. Similarly, smoothed cartesian speeds and accelerations (i.e.  $\dot{X}$ ,  $\dot{Y}$ ,  $\dot{Z}$  and  $\ddot{X}$ ,  $\ddot{Y}$ ,  $\ddot{Z}$ ) were computed from the first and 221 222 second derivatives of the spherical coordinate smoothing spline functions (Hedrick et al. 2018). 223 An initial examination of smoothing results showed that high frequency noise was efficiently 224 removed from position series, but remained present in speed and acceleration data, an issue that

225 could partly be improved by increasing the smoothing tolerance by 20 %. To ensure that the 226 smoothing tolerance value did not affect our results, we performed a sensitivity analysis, where 227 the base smoothing tolerance was increased by 0% and 40%, with no significant effect on the 228 results presented below (see Table S1, Fig. S7). 229 230 **Biomechanical variables** 231 A set of biomechanical variables was calculated to describe the flight behaviours of 232 house martins: Flight speed in the air reference frame  $(m.s^{-1})$ : 233  $S_a = |\mathbf{v}_a| = |\mathbf{v} - \mathbf{A}|$ 234 235 (1) 236 where  $\mathbf{v}_a$  is the velocity vector in the air reference frame, computed by subtracting wind 237 speed vector A, calculated from weather station data averaged over the duration of each 238 trajectory, from v, the bird velocity vector  $(\dot{X}, \dot{Y}, \dot{Z})$ . The norms of the horizontal and vertical 239 components of  $v_a$ ,  $s_{ha}$  (horizontal speed in the air reference frame) and  $s_Z$  (vertical speed) were 240 also calculated. Note that we measured wind speed and direction in the horizontal plane only, 241 hence A has no vertical component and  $s_Z$  values are equal in the ground and air reference 242 frames. 243 Mass-specific rate of change in potential energy (W.kg<sup>-1</sup>):  $P_p = g s_Z$ 244 245 (2) where g is the magnitude of gravitational acceleration. 246 247 248 Mass-specific rate of change in kinetic energy (W.kg<sup>-1</sup>): 249  $P_k = \mathbf{v}_a \cdot \mathbf{a}$ 

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250	(3)
251	where <b>a</b> is the acceleration vector $(\ddot{X}, \ddot{Y}, \ddot{Z})$ .
252	Mass-specific kinematic power (W.kg <sup>-1</sup> ):
253	$P = P_p + P_k$
254	(4)
255	Note that power values are mass-specific, as the body masses of individual birds are unknown.
256	Finally, to measure flight turns in trajectories, we calculated the following
257	variables:
258	Instantaneous radius of curvature (m):
259	$R = \frac{ \mathbf{v} ^3}{\sqrt{ \mathbf{v}_a ^2  \mathbf{a} ^2 - (\mathbf{v}_a \cdot \mathbf{a})^2}}$
260	(5)
261	where $\mathbf{a'}$ is the transpose of $\mathbf{a}$ .
262	Note that $R$ is a measure of flight direction change in any plane, not limited to horizontal turns.
263	Mass-specific centripetal force $(m.s^{-2})$ :
264	$F = \frac{ \mathbf{v}_a ^2}{R}$
265	(6)
266	
267	
268	Statistical analysis
269	Most graphical representations and associated statistical analyses were performed in
270	MATLAB r2018b. To visualize the flight envelope of the recorded house martins, several pairs
271	of variables were represented: $s_{ha}$ (horizontal airspeed) vs $s_Z$ (vertical speed), $P_{p1s}$ (rate of
272	change in potential energy) vs $P_{k1s}$ (rate of change in kinetic energy) and $s_a$ (airspeed) vs R
273	(instantaneous radius of curvature). Rates of change in kinetic and potential energy were

274 averaged over 1 s (10 frames) segments because these derivative variables are more susceptible 275 to noise, even after smoothing. Moreover, they were only averaged over 1 s segments where 276 the flight behaviour (gliding or flapping) did not change to be able to classify each 1 s segment 277 as entirely gliding or flapping. For each pair of variables, the distribution of all data points was 278 visualized by creating a kernel density estimation of the bivariate distribution, by plotting the 279 contours containing 50% and 90% of this estimated distribution, and then by only displaying 280 individual data points if they were outliers, i.e. outside of the 90% contours. For each pair of 281 variables, this process was repeated for flapping data points and gliding data points to separate 282 the two distributions. The univariate distributions of each variable, divided by gliding and 283 flapping, were then statistically compared. The R software v4.1.2 (R Core Team 2021) with the 284 forecast package v8.16 (Hyndman and Khandakar 2008; Hyndman et al. 2022) were used to 285 inspect the autocorrelograms and partial autocorrelograms of the initial time series, which 286 revealed that all variables were temporally autocorrelated, but that keeping one point out of five was enough to remove temporal autocorrelation for all tested variables  $(P, s_a, s_Z, R)$  in most 287 288 trajectories. Autocorrelation was removed independently in each time series (gliding points 289 series and flapping points series) by keeping a minimum interval of 5 frames between each 290 point (except for  $P_{p1s}$  and  $P_{k1s}$  for which averaging over 1 s already removed autocorrelation). 291 The means of these resulting distributions were then compared using t-tests.

To test for the effect of wind on flight speed, data points were divided into three directional bins based on the angle between the bird's instantaneous horizontal direction and the wind vector direction: downwind (0–60 deg), crosswind (60–120 deg) and upwind (120– 180 deg). The directional bins were separated between gliding and flapping, totalling to six bins. For each trajectory, a mean airspeed value was calculated for each bin, and statistical comparisons were carried out on the 95 trajectories having at least one point classified into every bin. The distributions of the six bins were visualized using violin plots created with the

violinplot function in MATLAB (Bechtold 2016), and the means of each directional bin were compared within each behavioural category using ANOVA. Significant ANOVA were followed by Tukey-Kramer post hoc tests. Furthermore, a linear model was created for each directional bin to analyse the relationship between wind speed ( $s_w$ ) and bird's airspeed ( $s_a$ ).

The link between weather variables and vertical speed ( $s_Z$ ) was studied by dividing data points into flapping or gliding and then by averaging vertical speed over all the data points of both behavioural categories for each trajectory. Three weather variables were also averaged over the entire trajectory: temperature, solar radiation and humidity. Six linear models were then created to analyse the relationship between mean vertical speed and these three weather variables for each behavioural category.

309 Finally, airspeed  $(s_a)$ , vertical speed  $(s_z)$  and mass-specific power averaged over 10 310 consecutive frames where flight behaviour did not change  $(P_{1s})$  were analysed to test if their 311 distributions differed between juveniles and adults. Only the 20 adult individuals recorded 312 during the two field sessions when juveniles were observed were retained to ensure that all 313 individuals were recorded in similar conditions (same weather and same period in the breeding 314 season). For each variable and each behavioural category (gliding or flapping), we pooled data 315 points available for the 5 juveniles (after removing temporal autocorrelation) to obtain a 316 distribution. For adults, we randomly sampled 5 individuals out of 20 adults to obtain a 317 comparable distribution and performed a two-sample Kolmogorov-Smirnov (KS) test (kstest2 318 function in Matlab). This KS test was replicated 100 times with different random adult samples. 319 Each KS test returned a *D* statistic (Sokal and Rohlf 1981), which is considered significant if:

320 
$$D > \sqrt{-\ln\left(\frac{\alpha}{2}\right)\frac{1}{2}\frac{(m+n)}{(m\,n)}}$$

321

(7)

where m and n are the number of data points for adults and juveniles respectively, andalpha the significance level.

In the present case, as randomly sampled adult birds had variable flight track duration, m varied and *D* values were not directly comparable between the KS tests. Instead, we derived a sample-size-independent *C* value:

327 
$$C = D / \sqrt{\frac{(m+n)}{(m n)}}$$

328

We then computed the significance level from the mean *C* value over the 100 KS tests:

$$\alpha = 2e^{-2C^2}$$

$$331 \tag{9}$$

332

### 333 **Results**

#### 334 General description of flight behaviour

Fig. 2 shows a first investigation of the flight speed distribution of house martins flying 335 336 near their colony by comparing the distribution of vertical speed and horizontal airspeed for all 337 data points (N = 45,170, Fig. 2A) or by comparing gliding and flapping flight (N = 25,414 and 338 15,810 respectively, Fig. 2B). Note that gliding and flapping totals do not add up to the total 339 number of data points, because flight mode was not visible for 8.7% of video frames. The 90% 340 area for all data points (Fig. 2A) shows that most of the time, house martins have a vertical speed between -4 and 4 m.s<sup>-1</sup>, and an horizontal airspeed between 3 and 11 m.s<sup>-1</sup>. The data 341 points also show the most extreme values exhibited by the recorded house martins, with vertical 342 speeds higher than 6 m.s<sup>-1</sup> and lower than -8 m.s<sup>-1</sup>, and horizontal airspeeds near 15 m.s<sup>-1</sup>. 343 344

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(8)

345 Dividing the data points into gliding and flapping flight (Fig. 2B) reveals that both 346 vertical speed and horizontal airspeed are significantly higher when house martins are flapping (mean vertical speed: -0.36 m.s<sup>-1</sup> vs 0.85 m.s<sup>-1</sup>, mean horizontal speed: 6.77 m.s<sup>-1</sup> vs 7.21 m.s<sup>-1</sup> 347 <sup>1</sup>, for gliding vs flapping, respectively; see Table 2 for details). It is expected to find that 348 349 flapping birds have more positive vertical speeds since flapping is often used to gain altitude, 350 but it is worthwhile to note that a significant proportion of data points associated with gliding 351 show a positive vertical speed, as even the 50% area contains points with positive vertical 352 speeds. Positive vertical speeds while gliding can be associated with the use of external energy 353 sources (thermal soaring, slope soaring, wind gradients) but also with a decelerating ascent. It 354 is necessary to study the rates of change in kinetic and potential energy to discriminate between 355 these two scenarios.

Fig. 3 shows the distribution of rates of change in potential and kinetic energy averaged over 1 s for all data points (Fig. 3A) or by comparing gliding and flapping flight (Fig. 3B), with isolines corresponding to several kinematic power values (i.e. the sum of rates of change in potential and kinetic energy, see Eq. 4). The 90% area for all data points (Fig. 3A) shows that house martins have power values between -25 and 30 W.kg<sup>-1</sup> during most of their flight behaviours near the colony.

When comparing gliding and flapping flight (Fig. 3B), the rate of change in potential energy is significantly higher when house martins are flapping (mean -6.00 vs 10.35 W.kg<sup>-1</sup>, gliding vs flapping; see Table 2). The difference is less noticeable for the rate of change in kinetic energy, but it is significantly higher for gliding flight (mean 1.57 vs 0.53 W.kg<sup>-1</sup>, gliding vs flapping). Kinematic power values exhibited by gliding house martins are usually negative (as expected due to adverse air friction and drag), but a significant portion of the gliding distribution shows positive power values, and the P = 0 W.kg<sup>-1</sup> isoline even crosses the 50%

area of the distribution. This demonstrates that the use of external energy sources is commonfor gliding house martins in this context.

371 The magnified view of the kernel contours for gliding (Fig. 3C) and flapping (Fig. 3D) 372 allows to identify several kinds of flight behaviours. As stated above, gliding flight (Fig. 3C) in the zone above the P = 0 W.kg<sup>-1</sup> isoline is not uncommon and reflects mechanical energy 373 374 gain, i.e. the use of external energy sources, which can be divided in several categories: data points where  $P_{p1s}$  (and consequently vertical speed  $s_Z$ ) is positive while P is also positive 375 376 corresponds to soaring house martins (thermal soaring, slope soaring, zone 1 in Fig. 3C) which 377 can be associated with a decreasing  $(P_{k1s} < 0)$  or increasing flight speed  $(P_{k1s} > 0)$ . Gliding flight 378 with positive P can also happen for house martins losing altitude ( $P_{p1s} < 0$ ) but accelerating  $(P_{k1s} > 0, zone 2)$ , which could reflect that birds can also use downward or forward wind gusts 379 380 to accelerate and gain some energy.

381 At the opposite, gliding flight is often associated with a negative *P* and a descent ( $P_{p1s}$ 382 < 0), as expected for typical, passive gliding (zone 3). Note that negative *P* while gliding can 383 also be observed with positive  $P_{p1s}$ , (zone 4) which reflects passive ascents, implying 384 deceleration ( $P_{k1s} < 0$ ) and some expected energy loss (P < 0).

385 Regarding flapping flight (Fig. 3D), it is obviously most of the time associated with 386 positive P, whether it be for ascending (bird accelerates or decelerates, zone 1) or descending 387 flight (bird accelerates, zone 2). However, it is worthwhile to note that a part of the 90% area 388 of the flapping distribution surprisingly shows negative power values. Data points with negative 389 P in ascent (zone 4 in Fig. 3D) could be associated with cases when the bird is struggling to 390 gain altitude and is losing more kinetic energy than the gain in potential energy. Finally, data 391 points with negative P in descent ( $P_{p1s} < 0$ , zone 3) could be associated with flapping birds 392 encountering an unfavourable downward wind gust that results in mechanical energy loss, 393 despite the flapping muscular work. It is also possible that house martins sometimes flapped

their wings to brake (i.e. dissipate energy) and/or to generate lateral forces and perform sharper
turns in front of an obstacle (e.g. building wall) or to catch prey.

Finally, as the wind measurement method had several limitations (constant wind speed and direction were assumed during each recording and wind was only measured at a single point in space), we cannot exclude that the speeds and accelerations we measured are slightly different compared to the real airspeeds experienced by the birds if the wind varied in space and time during our recordings. This could influence the positions and spread of individual points in Fig. 3 to some extent.

These results show that house martins perform a wide diversity of flight behaviours near the colony, from fast traveling to slow manoeuvring, and that they regularly use external energy sources. The difference between gliding and flapping flight is not clear-cut with regards to vertical speed and power, and house martins are able to exhibit a wide diversity of behaviours in both flight modes.

407

408 Flight turns

409 Figure 4 helps to understand the turning behaviours of house martins by showing the 410 distribution of airspeed and instantaneous radius of curvature for all data points (Fig. 4A) or by 411 comparing gliding and flapping flight (Fig. 4B), with isolines corresponding to several 412 centripetal force values. The 90% area for all data points (Fig. 4A) shows that house martins 413 have a radius of curvature comprised between 1 and 100 m most of the time, associated with a centripetal force comprised between 0.1 and 2 g. Smaller radius of curvature was usually 414 415 associated with lower airspeed, which always maintained centripetal forces below 5 g. 416 Exceptionally small radiuses of curvature (near 10<sup>-1</sup> m in Fig. 4A) show that house martins are 417 occasionally able to perform decimetre-scale turns (mostly u-turns in front of nests), but at very low airspeeds ( $< 1 \text{ m.s}^{-1}$ ) and hence low centripetal forces (< 2 g). At the opposite, very large 418

radiuses of curvature (above 100 m) are also uncommon, which suggests that, in thisbehavioural context, house martins are turning most of the time and rarely fly in straight line.

The most common radiuses of curvature were comprised in the interval 2–20 m (50% area in Fig. 4A), clearly indicating a tortuous flight behaviour. Comparing gliding and flapping turns (Fig. 4B) does not show strong differences in distributions, but flapping is associated with significantly higher airspeeds (mean 6.98 vs 7.42 m.s<sup>-1</sup>, gliding vs flapping; see table 2) and larger radiuses of curvature (mean of Log<sub>10</sub>(R): 0.89 vs 0.99, gliding vs flapping). Centripetal force was significantly higher in gliding, but the differences were again small (mean 0.73 vs 0.68 g, gliding vs flapping).

428

429 Notable behaviours

Fig. 3 allowed to identify several types of flight behaviours exhibited by house martins, which may be more clearly understood by looking at individual trajectories. Figs S4, S5 and S6 show the 3D views of trajectories, along with several biomechanical variables. Several types of notable flight behaviours can be identified on these trajectories.

Firstly, thermal soaring is visible on some trajectories (e.g. Fig. S4), when a positive power is observed for a gliding bird gaining altitude. Long sequences with birds rising and circling in thermal updrafts, as can be seen for large soaring birds, were rarely observed for house martins. Rather, they seem to frequently extract environmental energy in small bursts while they fly near the colony. In addition to thermal soaring, slope soaring was also occasionally observed for birds flying near high buildings where upward wind gusts could occur.

Secondly, a temporal oscillation of vertical speed appeared on several trajectories (e.g.
Fig. S5). While the bird is mostly gliding, it is alternatively ascending and descending, again
probably using external energy sources since power is often positive. During these sequences,

444 vertical acceleration shows negative values that are regularly close to  $-1 \text{ g} (-9.81 \text{ m.s}^{-2})$  which 445 is observed for an object in free fall. This suggests that the gliding bird is alternating sequences 446 of ascensions and free falls.

Finally, some atypical flight behaviours described in Fig. 3 can be seen on individual trajectories, such as birds with a positive power during gliding descents (e.g. Fig. S5), which is probably due to downward wind gusts, and birds with a negative power during flapping descents (e.g. Fig. S6), which suggests that flapping is sometimes used to generate adverse forces used for braking or to perform a sharp turn (e.g. for prey capture), or even for a purpose other than transport (e.g. in-flight preening).

453

454 Effect of wind on flight speed

455 Fig. 5 shows a comparison of the distributions of bird mean airspeed according to the 456 wind direction relative to the bird's direction, for gliding flight (Fig. 5A) and flapping flight 457 (Fig. 5B). Significant differences were only observed for flapping flight, where mean airspeed 458 is significantly higher (ANOVA, p < 0.001) for birds flying upwind (7.67 ± 1.01 m.s<sup>-1</sup>, mean  $\pm$  SD) compared to birds flying downwind (7.10  $\pm$  0.99 m.s<sup>-1</sup>) and crosswind (7.32  $\pm$  0.92 459 460 m.s<sup>-1</sup>). However, linear models studying the link between airspeed and wind speed show that 461 wind has a significant effect on both gliding (Fig. 6A–C) and flapping flight (Fig. 6D–F). Birds 462 flying downwind show a significant decrease in airspeed with increasing wind speeds for gliding (Fig. 6A) and a non-significant decrease for flapping (Fig. 6D), while birds flying 463 464 upwind show a significant increase of their airspeed with windspeed for both flight behaviours 465 (Fig. 6C, 6E). These results suggest that, overall, house martins adjust their flight speed, 466 reducing their airspeed when wind is pushing them, and increasing it when they have to fly 467 against the wind.

468

#### 470 Effect of weather on flight behaviours

471 Studying the effect of several weather variables on vertical speed (Fig. 7) shows that 472 only the vertical speed in gliding flight increases with temperature and solar radiation (Fig. 7A-473 B) and decreases with humidity (Fig. 7C), while there is no significant effect on the vertical 474 speed in flapping flight. Hot and sunny conditions are favourable to the formation of thermal 475 updrafts, and they are associated with less negative or even positive vertical speeds for gliding 476 house martins (note that here each point represents the mean vertical speed for a given trajectory, i.e. is a sum of sequences of thermal/slope soaring and descending gliding flight 477 478 bouts). This observation is consistent with the use of thermal updrafts by house martins, and 479 this confirms that this behaviour is frequent and important for these birds near their colony since 480 it is still visible at the scale of whole trajectories.

481

#### 482 Differences between juveniles and adults

483 Fig. 8 shows the distribution of airspeed (Fig. 8A, 8C) and vertical speed (Fig. 8B, 8D) 484 for gliding and flapping flight for the 5 juveniles and the 20 adults recorded during two field sessions (8<sup>th</sup> and 15<sup>th</sup> of July). Airspeed during gliding (Fig. 8A) did not differ significantly 485 486 between adults and juveniles (randomized KS tests, C = 1.13, p = 0.16), nor did vertical speed during gliding or flapping (Fig. 8B, 8D; C = 1.27, p = 0.078 and C = 1.14, p = 0.15, 487 488 respectively). Only airspeed during flapping significantly differed between adults and juveniles 489 (C = 2.23, p < 0.0001), with a flatter, right-shifted distribution for juveniles (Fig. 8C). Median 490 airspeed during flapping was about 0.8 m.s<sup>-1</sup> higher in juveniles (7.78 vs. 6.94 m.s<sup>-1</sup>, juveniles 491 vs. adults). No significant difference in kinematic power was found between adults and 492 juveniles (C = 0.75, p = 0.64 and C = 0.69, p = 0.78 for  $P_{1s}$  during gliding and flapping, 493 respectively).

### **Discussion**

496	Our study gives a quantitative description of the flight behaviours of the house martin
497	near the colony during the breeding season at fine spatial and temporal scales. Our results show
498	that house martins do use some strategies to save energy during this critical period of their life
499	cycle, such as extraction of environmental energy (Fig. 3), or optimisation of their cost of
500	transport in the ground reference frame (Fig. 5, 6).
501	
502	Distribution of biomechanical variables
503	The 90% area for horizontal and vertical speed (Fig. 2) was rather large $(3-11 \text{ m.s}^{-1} \text{ for}$
504	horizontal speed and $-4-4$ m.s <sup>-1</sup> for vertical speed), showing that house martins perform a wide
505	diversity of flight behaviours near the colony, whether it be fast traveling, or slow manoeuvring.
506	The total range of airspeeds (including horizontal and vertical components) was $0.5-15.1 \text{ m.s}^{-1}$ .
507	This speed range is quite similar to those observed in other hirundine species, such as foraging
508	barn swallows (3.7–19.4 m.s <sup>-1</sup> ; Warrick et al. 2016) and cliff swallows performing intraspecific
509	chases (2.8–14.0 m.s <sup>-1</sup> ; Shelton et al. 2014).
510	The distribution of rates of change in kinetic and potential energy (Fig. 3) highlighted
511	the use of external energy sources by house martins (discussed in a later section), but some
512	parts were rather unexpected, such as the positive power values exhibited by some house
513	martins in gliding descent, or the negative power values of some individuals during active
514	flapping. These unexpected behaviours can be associated with specific purposes (e.g. braking
515	in the case of flapping with negative power) but could also be associated with specific
516	environmental conditions (e.g. favourable wind gust in the case of gliding descent with positive
517	power, or adverse wind gust in the case of flapping with negative power). The difference

518 between gliding and flapping flight is not as clear-cut as expected with regards to vertical speed 519 and power, and house martins are able to exhibit a wide diversity of behaviours in both flight 520 modes.

521 House martins only performed the sharpest turns at low speeds, so their centripetal force 522 never exceeded 5 g (Fig. 4) and was most of the time below 2 g, a value consistent with the 523 average maximum centripetal force of 1.38 g found in foraging common swifts (Hedrick et al. 524 2018). By contrast, other aerial insectivores perform sharp turns with higher centripetal forces, 525 such as cliff swallows reaching 8 g during intraspecific chases (Shelton et al. 2014), or foraging 526 barn swallows performing 7 g turns (Warrick et al. 2016). These differences are consistent with 527 the contrasting foraging behaviours of house martins and barn swallows, since the former often 528 forage at higher altitudes in more open spaces (del Hoyo et al. 2020), while the latter often 529 forage near the ground in relatively cluttered environments (Brown and Brown 2020). In this 530 regard, foraging house martins may be more comparable to common swifts and could thus rely 531 on "gleaning" unsuspecting prey rather than catching evasive prey with sharp turns.

532

#### 533 Environmental energy extraction

534 In our study, positive power values are often observed in gliding house martins (Fig. 3), 535 which shows that they regularly use external energy sources such as thermal updrafts, upward 536 wind gusts and wind gradients. Most of the time, they apparently only use these energy sources in short bursts, and individuals circling in thermal updrafts for an extended period were rarely 537 538 observed. Even when a house martin uses a thermal updraft for a longer duration, vertical speed 539 is not constantly positive and often shows temporal oscillations (see Fig. S5) which could be 540 associated either with prey capture, or with aerial preening (the latter behaviour was clearly 541 visible on some video recordings). Thermal soaring may be the main source of energy 542 extraction, as shown by the significant effects of temperature, solar radiation and humidity on

543 vertical speed (Fig. 7), but other strategies were occasionally observed such as slope soaring 544 along the high buildings on which the colony was based, or occasional extraction of 545 environmental energy during accelerating gliding descent, presumably from downward wind 546 gusts (Fig. S5).

547 The use of thermal updrafts was also commonly observed in foraging common swifts 548 (Hedrick et al. 2018), and these updrafts may be an important environmental feature for 549 foraging aerial insectivores, both as a source of mechanical energy and as a substrate for patches 550 of aerial arthropods (de Margerie et al. 2018), because rising air currents can contain a wide 551 diversity of floating prey (Geerts and Miao 2005; Wainwright et al. 2017). For large soaring 552 raptors feeding on the ground, a framework suggested by Shepard et al. (2011) considers that 553 the distribution of mechanical energy sources (thermal updrafts) may be an important constraint 554 in the foraging behaviour of these species. Even if soaring per se is not as vital for aerial 555 insectivores, which can flap their wings at a much lower cost than large raptors (Pennycuick 556 2008), here thermal updrafts can be considered as a source of both types of energy (mechanical 557 energy and food energy), so their spatial and temporal distribution may also have drastic 558 consequences on the foraging behaviour of aerial insectivores. Consequently, atmospheric 559 conditions may strongly impact the availability of resources for aerial insectivores, and rapidly 560 changing conditions could impact their foraging and breeding success.

561

562 Effect of wind on flight speed

House martins follow the general tendency to reduce cost of transport, observed in migrating and commuting birds (Wakeling and Hodgson 1992; Hedenström et al. 2002; Kogure et al. 2016; Sinelschikova et al. 2019) and also in foraging swifts (Hedrick et al. 2018), decreasing their airspeed when flying downwind, and increasing it when flying upwind (Fig. 5, 6). This tendency was visible on gliding flight, and partly on flapping flight, despite a relatively

568 narrow range of wind speed variation during our field sessions (mean wind speed over a 569 trajectory never exceeded 2.5 m.s<sup>-1</sup>). It is also worth noting that our method of averaging wind 570 speed and direction over a complete trajectory cannot detect more subtle effects of wind 571 variation at finer temporal scales, such as wind gusts. Moreover, we only measured wind speed 572 and direction at one fixed position, which does not take into account wind variations caused by 573 height and the presence of obstacles. Even so, our results suggest that house martins optimise 574 their movements in the ground reference frame, probably because of the presence of their nest 575 at a fixed ground position (central-place foraging; Bryant and Turner 1982).

576

#### 577 Differences between adults and juveniles

578 A significant difference between juveniles and adults was found in the distribution of 579 airspeed during flapping, with juveniles flying at higher, more variable speeds (Fig. 8C). This 580 suggests that the development of flight behaviour in house martins might not be fully mature at 581 fledging (as in many other bird species; Ruaux et al. 2020). Similar differences might exist for 582 other variables (such as a slightly flatter distribution of vertical speed during gliding, Fig. 8B), 583 but the low number of trajectories from clearly identified juvenile birds prevented more precise 584 investigation. The recorded juveniles were likely performing some of their first flights, so they 585 might not be as precise as adults in controlling their flight speed and altitude and would thus 586 need more efforts to adjust their speed and their trajectory. In house martins, post-fledging 587 locomotor ontogeny may consist in a reduction of speed variability (i.e. improvement of flight 588 speed control) in order to converge towards the most energy-efficient speeds in a given context. 589 As a consequence, juvenile house martins might be less effective aerial foragers than 590 adults, because of a lower energy intake from feeding and/or because of a higher energy output 591 in flight. Indeed, catching arthropods in flight is a complex behaviour, and for example, it has 592 been shown in juvenile black phoebes (Sayornis nigricans) that the proportion of successful

593 foraging attempts increased gradually in juveniles to reach the same level as adults at the age 594 of seven weeks. This increase is potentially due to trial-and-error learning, but the maturation 595 of cognitive or visual systems cannot be ruled out (Marchetti and Price 1989; Gall et al. 2013). 596 Juvenile house martins return to the nest to roost and are still fed by their parents for a few days 597 after fledging (del Hoyo et al. 2020), which suggests that they are not immediately as efficient 598 as adults in catching prev. During this period, juvenile house martins likely benefit from social 599 learning when foraging near the colony (Varland et al. 1991; Bustamante 1994; Heyes 1994; 600 Kitowski 2009). Further studies comparing the energy intake and energy expenditure of 601 juvenile and adult house martins could clarify these potential differences. It is also possible to 602 hypothesize that some of the differences observed here between juveniles and adults are due to 603 playful behaviours specific to juveniles. Such behaviours can also represent a way to 604 experiment different flight behaviours and to gain experience.

To conclude, our study gives a first general description of the flight behaviours of house martins near the colony during the breeding season and suggests several mechanisms by which they might save energy. House martins have little margin for lower energy intake and higher energy expenditure during this critical period, so their flight behaviours reflect a set of adaptations to optimise energy gain. Juveniles may not be immediately as efficient as adults in maximising their energy input while minimising their output, so parental care and social learning potentially play a critical role during the first few days out of the nest.

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- 614

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621

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#### 627 Data availability

628 The flight trajectory and environmental data used as a basis for this analysis are publicly

629 available from the figshare digital repository:

630 <u>https://doi.org/10.6084/m9.figshare.21118408.v1</u>

631

### 632 **Compliance with Ethical Standards**

633 Conflicts of interest:

634 The authors have no relevant financial or non-financial interests to disclose.

635 Ethics approval:

636 This is an observational study, without any animal manipulation or disturbance.

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- 731

#### 733 Table 1 List of symbols and abbreviations

а	bird acceleration vector in the ground reference frame						
az	bird vertical acceleration						
Α	wind speed vector						
F	mass-specific centripetal force						
g	magnitude of gravitational acceleration						
P	mass-specific kinematic power (sum of rates of change in mass-specific						
1	kinetic and potential energy)						
$P_k$	mass-specific rate of change in kinetic energy						
P <sub>p</sub>	mass-specific rate of change in potential energy						
R	instantaneous radius of curvature						
RSV	rotational stereo-videography						
Sa	bird speed in the air reference frame						
<b>S</b> <sub>ha</sub>	bird horizontal speed in the air reference frame						
Sw	wind speed						
SZ	bird vertical speed						
V	bird velocity vector in the ground reference frame						
Va	bird velocity vector in the air reference frame						
X, Y, Z	bird cartesian coordinates in the ground reference frame						
Θ	azimuthal angle measurement from RSV						
Р	radial distance measurement from RSV						
Φ	elevation angle measurement from RSV						
	dot-over character, indicating first derivative						
	with respect to time						
	double dot-over character, indicating second derivative						
	with respect to time						
Subscript 1s	variable averaged over 1 second (10 consecutive frames where flight						
Casconpt 10	behaviour did not change)						

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735

#### 736 Table 2 Summary of the quantified variables for gliding vs. flapping flight of house martins. N

- values reflect sample sizes after autocorrelation was removed from time series (see methods)
- 738

Variable	Abb.	Unit	Gliding flight	Ν	Flapping	Ν	t-test
			$(mean \pm SD)$		flight (mean $\pm$		
					SD)		
					,		
Airspeed	Sa	m.s <sup>-1</sup>	$6.98 \pm 1.60$	5756	7.42 ± 1.58	3807	t(9561) = -
1							13.25
							p <0.001
Horizontal airspeed	<b>S</b> ha	m.s <sup>-1</sup>	$6.77 \pm 1.62$	5756	$7.21 \pm 1.64$	3807	t(9561) = -
							12.94
							p <0.001
Vertical airspeed	Sz	m.s <sup>-1</sup>	$-0.36 \pm 1.63$	5756	$0.85 \pm 1.45$	3807	t(9561) = -
							37.22
							p <0.001
Mass-specific rate	$P_{k1s}$	W.kg <sup>-1</sup>	$1.57\pm10.73$	1821	$0.53\pm10.55$	884	t(2703) =
of change in kinetic							2.38
energy, averaged							p = 0.018
over 1 s							
Mass-specific rate	$P_{p1s}$	W.kg <sup>-1</sup>	$-6.00 \pm 11.92$	1821	$10.35\pm10.84$	884	t(2703) = -
of change in							34.43
potential energy,							p <0.001
averaged over 1 s							
Radius of	$Log_{10}(R)$	-	$0.89\pm0.35$	5756	$0.99\pm0.37$	3807	t(9561) = -
curvature, Log-							13.19
transformed							p <0.001
Mass-specific	F	g	$0.73\pm0.41$	5756	$0.68\pm0.42$	3807	t(9561) =
centripetal force							5.70
							p <0.001

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740

**Fig. 1 Aerial view of the recording site.** The red dot indicates the location of the RSV device, the green dot indicates the location of the weather station, and blue dots indicate the location of calibration points. The blue arrow indicates the direction of the sixth calibration point, located 410 m away from the RSV device and not represented here for legibility. The yellow line shows an example of a trajectory, with the white dot marking the beginning and the black dot marking the end. The colony is located on all the buildings on the right side of the aerial view (e.g. where the example trajectory ends). Source for aerial view: Google Earth

Fig. 2 Distribution of vertical speed (s<sub>z</sub>) versus horizontal speed in the air reference frame (s<sub>ha</sub>). (A) All data points are represented by grey circles, with two contours indicating the areas containing 50% and 90% of the kernel density distribution. (B) 50% and 90% contours for gliding and flapping, only data points outside of the 90% areas appear. Gliding is represented by blue circles and contours and flapping by red triangles and contours. The univariate distributions of data points are represented along the axes of each panel

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755 Fig. 3 Distribution of rate of change in potential energy over 1 s ( $P_{p1s}$ ) versus rate of change in 756 kinetic energy over 1 s (Pk1s). (A) All data points are represented by grey circles, with two contours 757 indicating the areas containing 50% and 90% of the kernel density distribution. (B) 50% and 90% 758 contours for gliding and flapping, only data points outside of the 90% areas appear. Gliding is 759 represented by blue circles and contours and flapping by red triangles and contours. Equivalent values 760 of vertical speed averaged over 1 s ( $s_{Z_{1s}}$ ) are given in the y axis. The univariate distributions of data 761 points are represented along the axes of the upper panels. The dashed lines are isolines for power 762 values from -40 to 40 W.kg<sup>-1</sup>. The lower panels are magnified views of only the kernel contour of gliding 763 (C) or flapping flight (D). On these lower panels, zone 1 represents positive power combined with 764 positive vertical speed, zone 2 represents positive power combined with negative vertical speed, zone 765 3 represents negative power combined with negative vertical speed, and zone 4 represents negative 766 power combined with positive vertical speed. Flight behaviours associated with these zones are 767 discussed in the text

Fig. 4 Distribution of airspeed ( $s_a$ ) versus instantaneous radius of curvature (R). R is represented in logarithmic scale. (A) All data points are represented by grey circles, with two contours indicating the areas containing 50% and 90% of the kernel density distribution. (B) 50% and 90% contours for gliding

and flapping, only data points outside of the 90% areas appear. Gliding is represented by blue circles
and contours and flapping by red triangles and contours. The univariate distributions of data points are
represented along the axes of each panel. The dotted lines are isolines for centripetal force values from
0.1 to 5 g

775 Fig. 5 Distribution of mean airspeed ( $s_a$ ) versus wind direction category. Each dot represents the 776 mean vertical speed for all downwind/crosswind/upwind flight bouts in a given trajectory. (A) Gliding. (B) 777 Flapping. White dots represent the medians, vertical bars represent the ranges from the 25<sup>th</sup> to the 75<sup>th</sup> 778 percentiles, and coloured zones represent the kernel density distributions of each category. Lowercase 779 letters (a and b) indicate significant differences after a significant single-factor ANOVA followed by 780 Tukey-Kramer tests (i.e. groups with label a are significantly different groups with label b). No significant 781 difference was found for gliding. Data for 95 trajectories for which at least one data point was available 782 in each category

Fig. 6 Mean airspeed ( $s_a$ ) versus wind speed ( $s_w$ ) divided by wind direction category (downwind in blue, crosswind in green and upwind in red). (A–C) Gliding flight. (D–F) Flapping flight. The formula of each linear model, its p-value and R<sup>2</sup> are indicated in each panel. The black dotted lines represent the 95% confidence interval of the slope. Data for 95 trajectories for which at least one data point was available in each category

788 Fig. 7 Mean vertical speed (*sz*) versus temperature (A), solar radiation (B), and humidity (C).

789 Each dots represents the mean vertical speed of all gliding/flapping bouts in a given trajectory.

790 Gliding is represented by blue circles and flapping by red triangles. The formula of each linear model,

791 its p-value and  $R^2$  are indicated in each panel. The black dotted lines represent the 95% confidence

- interval of the slope
- Fig. 8 Distribution of airspeed ( $s_a$ ) and vertical speed ( $s_z$ ) values according to age class for the 25 birds recorded during sessions 7 and 8 (8<sup>th</sup> and 15<sup>th</sup> of July). (A–B) Gliding. (C–D) Flapping
- 795
- 796
- 797
- 798
- 799



Fig 1 

























822 823 Fig 8 824

### Electronic supplementary material for the article "Flight behaviours and energy savings in adult and juvenile house martins (*Delichon urbicum*) foraging near their breeding colony"

#### in the journal Behavioral Ecology and Sociobiology

by Geoffrey Ruaux, Kyra Monmasson, Tyson L. Hedrick, Sophie Lumineau and Emmanuel de Margerie<sup>1</sup>.

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Figure S1: RSV device in the field, in front of the wide lawn and urban gardens where house martins were recorded.



#### Figure S2: RSV device.

Rigid assembly of a camera and a set of mirrors rotating on a tripod with a fluid video head equipped with angular encoders. A second camera is visible on the right side, and was used to take pictures of the filmed birds with a greater magnification.



#### Figure S3: Random error in 3D location reconstruction, as a function of distance from the RSV device.

Red dots: mean error for calibration points. Red dotted lines: error for individual calibrations. Black dotted line: theoretical random error (from 3D space quantization only – see Methods section and de Margerie et al. (2015) for additional sources of error). The background histogram shows the distance distribution for all sampled bird locations.



Figure S4: 3D view of a house martin's trajectory exhibiting thermal soaring, along with several biomechanical variables versus time. On all panels, gliding is represented by blue segments, flapping by red segments, and undetermined behaviours by black thin segments (wing movement not visible on video record). (A) 3D view of the trajectory with temporal indications every 5 s. (B) Height (Z) versus time. (C) Vertical speed ( $s_z$ ) versus time. (D) Vertical acceleration ( $a_z$ ) versus time. (E) Airspeed ( $s_a$ ) versus time. (F) Mass-specific power (P) versus time. Asterisks indicate specific moments described in the text.

Figure S4 shows the 3D view of a trajectory where thermal soaring is apparent. Indeed, several sequences show positive vertical speeds (Fig. S4C) with positive powers (Fig. S4F) while the bird is gliding, for example between 7 and 10 s, or between 28 and 30 s. Each time, the gliding bird is gaining a few meters in altitude (Fig. S4B).



Figure S5: 3D view of a house martin's trajectory exhibiting temporal oscillations in vertical speed and gliding descents with positive power values, along with several biomechanical variables versus time. On all panels, gliding is represented by blue segments, flapping by red segments, and undetermined behaviours by black thin segments (wing movement not visible on video record). (A) 3D view of the trajectory with temporal indications every 5 s. (B) Height (Z) versus time. (C) Vertical speed ( $s_z$ ) versus time. (D) Vertical acceleration ( $a_z$ ) versus time. (E) Airspeed ( $s_a$ ) versus time. (F) Mass-specific power (P) versus time. Asterisks indicate specific moments described in the text.

Figure S5 shows the 3D view of a trajectory where several phenomena are visible. Firstly, vertical speed (Fig S5C) shows temporal oscillations, mostly between -2 and  $2 \text{ m.s}^{-1}$ . While the bird is mostly gliding, it is alternatively ascending and descending, again probably using external energy sources as P > 0 during gliding is often observed (Fig. S5F). Vertical acceleration (Fig. S5D) shows negative values that are regularly close to  $-1 \text{ g} (-9.81 \text{ m.s}^{-2})$  which is observed for an object in free fall. Thus, this suggests that the bird is alternating sequences of ascensions and free falls while gliding. Besides, another phenomenon is visible on this trajectory: some birds have a positive power during gliding descents (e.g. at 11 s or at 35 s), which means that it is accelerating more than what its altitude loss would imply. This could be due to downward wind gusts.



Figure S6: 3D view of a house martin's trajectory exhibiting flapping descents negative power values, along with several biomechanical variables versus time. On all panels, gliding is represented by blue segments, flapping by red segments, and undetermined behaviours by black thin segments (wing movement not visible on video record). (A) 3D view of the trajectory with temporal indications every 5 s. (B) Height (Z) versus time. (C) Vertical speed ( $s_z$ ) versus time. (D) Vertical acceleration ( $a_z$ ) versus time. (E) Airspeed ( $s_a$ ) versus time. (F) Mass-specific power (P) versus time.

Figure S6 shows the 3D view of a trajectory where the bird, contrary to previous examples, does not seem to use thermal updrafts, performs most ascents with active flapping, and gliding is limited to descent (Fig. S6B). In fact, this bird shows a peculiar behaviour: mechanically efficient flapping should be associated with positive mechanical power (as is the case in previous examples, see Fig. S4F and Fig. S5F where red bouts are mostly above zero power), but here the bird often used flapping flight during descents (red bouts below 0 in Fig. S6C), sometimes resulting in negative power (fig. S6F). Although surprising, it is possible that house martins can use active flapping to generate adverse forces used for braking or to perform a sharp turn (e.g. for prey capture), or even for a purpose other than transport (e.g. in-flight preening).

#### Table S1: Sensitivity analysis for smoothing tolerance.

The results of the main statistical tests carried out in our analyses are presented for the three values of smoothing tolerance considered. The value of 1.2 was used in our final analyses. Overall, the significance of results is not influenced by smoothing tolerance.

			1	
Statistical test	Statistical test	1.0	1.2	1.4
s glide vs flan	t tost	<i>t</i> (9561) = -37.09	t(9561) = -37.22	t(9561) = -37.33
s <sub>z</sub> glide vs hap	t-test	p < 0.001 ***	p < 0.001 ***	p < 0.001 ***
c glide vs flap	t tost	<i>t</i> (9561) = -12.79	<i>t</i> (9561) = -12.94	<i>t</i> (9561) = -12.95
s <sub>ha</sub> gilde vs hap	t-test	p < 0.001 ***	p < 0.001 ***	p < 0.001 ***
R glide vs flap	t tost	<i>t</i> (2703) = -34.40	<i>t</i> (2703) = -34.43	<i>t</i> (2703) = -34.48
<i>P</i> <sub>p1s</sub> gride vs hap	t-test	p < 0.001 ***	p < 0.001 ***	p < 0.001 ***
P., glide vs flap	t tost	<i>t</i> (2703) = 2.07	<i>t</i> (2703) = 2.38	<i>t</i> (2703) = 2.61
	t-test	p = 0.039 *	p = 0.018 *	p = 0.009 **
s glide vs flan	t-test	<i>t</i> (9561) = -13.08	<i>t</i> (9561) = -13.25	<i>t</i> (9561) = -13.27
	1 1051	p < 0.001 ***	p < 0.001 ***	p < 0.001 ***
log <sub>10</sub> ( <i>B</i> ) glide vs flan	t-test	t(9561) = -12.89	t(9561) = -13.19	<i>t</i> (9561) = -13.24
	1 1051	p < 0.001 ***	p < 0.001 ***	p < 0.001 ***
E glide vs flan	t-test	<i>t</i> (9561) = 4.87	<i>t</i> (9561) = 5.70	<i>t</i> (9561) = 6.17
	t test	p < 0.001 ***	p < 0.001 ***	p < 0.001 ***
Gliding s <sub>a</sub> , downwind (DW) vs	ANOVA			
crosswind (CW) vs upwind				
(UW)		p = 0.499 NS	p = 0.872 NS	p = 0.529 NS
Flapping <i>s<sub>a</sub></i> , DW vs CW vs UW	ANOVA	p < 0.001 ***	p < 0.001 ***	p < 0.001 ***
		DW vs CW: p = 0.2239	DW vs CW: p = 0.2622	DW vs CW: p = 0.2766
		NS	NS	NS
Flapping s <sub>a</sub> , Pairwise	lukey post-hoc test	DW vs UW:	DW vs UW:	DW vs UW:
comparisons		p < 0.001 ***	p < 0.001 ***	p < 0.001 ***
		CW vs UW:	CW vs UW:	CW vs UW:
		p = 0.0365 *	p = 0.0355 *	p = 0.0317 *
		DW: y = -0.534x	DW: y = -0.540x	DW: y = -0.542x
		p = 0.014 *	p = 0.014 *	p = 0.014 *
s vs s gliding	linear	CW: y = 0.005x	CW: y = 0.002x	CW: y = -0.004x
	model	p = 0.982 NS	p = 0.991 NS	p = 0.985 NS
		UW: y = 0.830x	UW: y = 0.827x	UW: y = 0.817x
		p < 0.001 ***	p < 0.001 ***	p < 0.001 ***
	linear	DW: y = -0.410x	DW: y = -0.410x	DW: y = -0.411x
		p = 0.078 NS	p = 0.072 NS	p = 0.070 NS
s <sub>a</sub> vs s <sub>w</sub> , flapping		CW: y = -0.310x	CW: y = -0.312x	CW: y = -0.316x
	model	p = 0.149 NS	p = 0.139 NS	p = 0.138 NS
		UW: y = 0.672x	UW: y = 0.682x	UW: y = 0.699x
		p = 0.003 **	p = 0.003 **	p = 0.002 **
	linear model	Temperature: y =	Temperature: y =	Temperature: y =
		0.02/x	0.02/x	0.02/x
$s_Z$ vs weather variables,		p = 0.001 **	p = 0.001 **	p = 0.001 **
giiding		Solar rad.: y = 0.001x	Solar rad.: y = 0.001x	Solar rad.: y = 0.001x
		n = 0.002 **	n = 0.002 **	n = 0.002 **
		n = 0.155  NC	n = 0.150  MS	n = 0.150 NC
sz vs weather variables,	linear	$p = 0.135 \text{ m}^3$ Solar rad $\cdot y = -0.000 \text{ v}$	p = 0.130  ms Solar rad $\cdot y = -0.000y$	$p = 0.139 \text{ ms}$ Solar rad $\cdot y = -0.000y$
flapping	model	p = 0.818  NS	p = 0.842  NS	p = 0.855  NS
		Humidity: $v = -0.001x$	Humidity: $v = -0.001x$	Humidity: $v = -0.001x$
		p = 0.711 NS	p = 0.691 NS	p = 0.685 NS



Figure S7: Example of the minor influence of smoothing tolerance on the graphical results: distribution of airspeed ( $s_a$ ) versus instantaneous radius of curvature (*R*). Three values of smoothing tolerance were tested: (A) 1.0, (B) 1.2 and (C) 1.4.