

1 *The use of social information in vulture flight decisions*

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18 Abstract

19 Animals rely on a balance of personal and social information to decide when and where to
20 move next in order to access a desired resource, such as food. The benefits from cueing on
21 conspecifics to reduce uncertainty about resources availability can be rapidly overcome by the
22 risks of within-group competition, often exacerbated toward low-ranked individuals. Being
23 obligate soarers, relying on thermal updrafts to search for carcasses around which competition
24 can be fierce, vultures represent ideal models to investigate the balance between personal and
25 social information during foraging movements. Linking dominance hierarchy, social affinities
26 and meteorological conditions to movement decisions of eight captive vultures, *Gyps spp.*,
27 released for free flights in natural-like soaring conditions, we found that they relied on social
28 information (i.e. other vultures using/having used the thermals) to find the next thermal updraft,
29 especially in unfavourable flight conditions. Low-ranked individuals were more likely to
30 disregard social cues when deciding where to go next, possibly to minimise the competitive
31 risk of social aggregation. These results exemplify the architecture of decision-making during
32 flight in social birds. It suggests that the environmental context, the context of risk and the
33 social system as a whole calibrate the balance between personal and social information use.

34

35 **Key-words:** griffon vulture, hierarchy, movement decision, landscape exploration, social information,
36 unpredictable resource

37 1. Introduction

38 Animals must constantly decide where and when to move next in order to find resources
39 such as food, water, shelter, or a mate, necessary for life. To make these decisions, they can
40 rely on two sources of information: personal information and social information. Personal
41 information includes knowledge of the spatiotemporal patterns of resource distribution that
42 individuals may perceive or have memorised from previous encounters [1]. For example, food-
43 storing birds are able to return to locations where they stored or saw food in the past, based on
44 prior expectation of the resource availability [2]. Social information, on the other hand, is
45 obtained by observing the behaviour of others [3–5]. Feeding, fleeing, or mating individuals
46 provide discrete information about the availability and locations of food, predators, or potential
47 mates.

48 For resources that are heterogeneously distributed in the environment, ephemeral and
49 unpredictable, using only personal information for movement decisions may be prone to
50 inaccuracies [6]. In such conditions, social animals may benefit from companions' knowledge
51 and may follow the dominant or oldest individual(s) considered as knowledgeable (e.g. homing
52 pigeons, *Columba livia*, or elephants, *Loxodonta africana*, [7,8]), follow the largest group
53 through shared decision-making [9], or stay with preferred affiliates [10–12]. Because using
54 social information can considerably reduce uncertainty in finding resources, individuals should
55 favour this source of information to achieve cost-efficient movement [13–15]. However,
56 relying heavily on social information can also lead individuals to aggregate on resources,
57 potentially inducing competition by exploitation or interference if the resource is
58 monopolizable and depletable [16]. Since both social and personal information are often
59 available to social animals [1], they need to balance their relative importance, depending on
60 the availability and predictability of the resource. When deciding on the next movement step,
61 social animals must trade-off the decreased uncertainty of locating a resource through social
62 information, with the potential increase in competition risk. Such a balance may be dictated by
63 the immediate needs of the individual and its risk sensitivity [17] but also by the group social
64 organisation. For example, low-ranked individuals are known to suffer more from within-group
65 competition compared to high-ranked individuals [18] and should therefore be more reluctant
66 to engage into social information use, which could eventually trigger proximity to despotic
67 individuals [19].

68 Vultures rely on two unpredictable resources: carcasses to feed and thermal updrafts to
69 move. During foraging flights, these large soaring birds gain altitude by circling into thermal

70 updrafts (i.e. masses of hot air rising from heated surfaces) and glide across the landscape to
71 the next updraft while scanning the ground for carcasses [20]. Although some topographic
72 features are clearly favourable to updrafts presence [21], at the individual level, challenging
73 local meteorological conditions (e.g. high wind speed, low temperature, high cloudiness) can
74 make thermal locations and availability hard to predict [22]. If they fail to detect an updraft,
75 vultures may be forced to switch to flapping flight, or worse to land and take-off again,
76 significantly increasing their energy expenditure [23,24]. While both thermals and carcasses
77 are relatively unpredictable, thermals are not depletable contrary to carcasses. When a vulture
78 discovers a carcass, its sharp drop in altitude while circling before landing is used as a signal
79 by conspecifics, dragging tens of individuals to the food source in a few minutes [25,26]. As
80 the number of vultures around the carcass increases (up to 100-120 individuals, [27,28]),
81 individual feeding rates decrease due to reduced access to the resource, resource depletion by
82 competitors and increased agonistic interactions [27]. Therefore, in these social birds,
83 individuals should balance the advantage of conspecific presence to locate thermal updrafts
84 [29] with the ultimate cost of competition around the carcasses that can be fierce [30–33]. As
85 such, vultures are ideal models to investigate the role of conspecifics in shaping their foraging
86 movement decisions.

87 Using a group of captive but freely-flying ‘griffon’ vultures, *Gyps fulvus* and *G.*
88 *rueppellii*, tagged with high-resolution GPS loggers, we studied how conspecifics’ presence
89 shapes individuals’ movement decisions during soaring flights. Despite being trained birds
90 released for public shows, these individuals sometimes detected and fed on carcasses at
91 surrounding farms (BN and YS, pers. obs.). We therefore consider these flights comparable to
92 natural foraging flights. Focusing on the movement steps from thermal to thermal, we first
93 assessed when do individuals preferentially discover new thermals (i.e. use of personal
94 information) compared to using thermals already discovered by conspecifics (i.e. use of social
95 information). We expected that vultures would favour the use of social information when
96 unfavourable meteorological conditions increased thermal unpredictability and when flight
97 conditions (e.g. low altitude) increased risks of landing [1]. Furthermore, given the hierarchy
98 in vulture groups, we expected low-ranked individuals to be more prone to use personal
99 information than high-ranked individuals to try to find the food source first, in order to avoid
100 large aggregation [34,35]. Second, we investigated the drivers underlying thermal selection
101 when individuals had to choose between simultaneously available thermals. We expected
102 individuals to select thermals providing the maximal positive vertical speed (i.e. climb rate) as
103 it may provide a reliable proxy of the thermal current strength helping them maximise their

104 height gain [29]. To decrease uncertainty about resource finding and risks mentioned above,
105 we expect that individuals should favour thermals hosting the maximum number of individuals
106 to maintain cohesion and secure the possibility to cue on as many conspecifics as possible [29].
107 Finally, social preferences may also influence decision, with individuals preferentially moving
108 together with preferred affiliates [12,36,37], as it could reduce competition due to familiarity
109 between individuals [38].

110 2. Materials and methods

111 2.1 Study site, vultures housing conditions and experimental settings

112 The study was carried out in 2021 and 2022 at the Rocher des Aigles falconry centre,
113 Rocamadour, France, and divided between winter and summer periods each year. During
114 winters, vultures were housed within an aviary (6.7 x 6 x 6 m) equipped with four perches:
115 three of them measuring 3.10 m, placed at 1.7, 2.6 and 3.5 m from the ground, and one of the
116 full width of the aviary at 4 m height. This setting was used to estimate vulture social bonds
117 (see Social bond estimation). In addition, besides being fed daily on small pieces of meat to
118 prevent conflicts, five feeding events (one each week during a five-week period) were
119 organised in the aviary on a butchery carcass occurring after a one-day fasting (to motivate
120 feeding). These feeding events were used to assess dominance hierarchy within the group (see
121 Hierarchy estimation). In summer, these trained vultures were kept perching on individual logs,
122 released several times per day to execute free flight shows for the public within a landscape
123 composed of plateaus interspaced by canyons, similar to “Causses” landscape typically used
124 by french wild vultures [39]. The falconry centre is located near a 120 m-deep canyon and
125 offers natural soaring conditions for raptors, making this study site a great place to investigate
126 natural group flight behaviour (see Group flights), [24].

127 We used GPS data and visual observations to characterise the social and flight
128 behaviour of eight captive vultures (7 Eurasian griffon vultures, *Gyps fulvus*, and 1 closely-
129 related Rüppell’s vulture, *Gyps rueppellii*), including five females and three males (Table S1).
130 Each year, we conducted experiments on a group of six individuals (two griffon vultures were
131 replaced in 2022, Table S1). Experiments followed the animal ethic guidelines of France and
132 the Centre National de la Recherche Scientifique. Handling of birds to fit GPS loggers followed
133 the protocol of telemetry study of vultures authorised in the Programme Personnel 961,
134 coordinated by OD, under the supervision of the French ringing centre, CRBPO, Paris.

135 Furthermore, experiments, observations, handling and flight events were systematically
136 performed under the guidance of the head of animal caretakers, BN.

137 2.1.1 Social bond estimation

138 During five weeks in both years (December/January 2020-2021 and
139 November/December 2021), we recorded pictures of vultures in the aviary from 8:00 to 19:00
140 (local time) at 5 minutes interval, using three camera traps (Wosport Big Eye D3 and Reconyx
141 HyperFire HC600).

142 We identified birds using repeated colours on plastic rings and marks on the ruff and
143 backhead feathers, using harmless colour sticks (Raidex GmbH, Figure 1A). We then processed
144 recorded pictures to extract the individuals' ID and position (bill or head position), using a
145 purpose-built image annotation program in Julia software, JuliaHub Inc., [39]. For subsequent
146 analyses, we relied on R software (v 4.2.2, R software, 2022, [40]).

147 We considered the social bond between a dyad of individuals i and j based on spatial
148 proximity following the Simple Ratio association Index (SRI, equation 1, [41,42])

149 Equation 1: $SRI_{i,j} = SRI_{j,i} = \frac{n_{tog}}{n_{tot}}$,

150 where n_{tog} is the number of pictures in which individuals i and j were on the same perch at a
151 Euclidean distance of less than 1.55 m and n_{tot} is the total number of pictures in which
152 individuals i and j were both detected on the same perch. SRI values varied between 0 and 1,
153 where 0 represented dyads that were never seen associated and 1 represented dyads that were
154 always observed sitting at less than 1.55 m from each other. The distance of 1.55 m was chosen
155 as matching to the mode of the inter-individual distances distribution (Figure S1). This was
156 also consistent with the aviary setting, as it corresponded to half the length of most available
157 perches. Our analyses were robust to other choices for this distance threshold (see
158 Supplementary Material, ESM01).

159 2.1.2 Hierarchy estimation

160 Each winter, we estimated hierarchy within the vulture group by monitoring feeding
161 interactions during the five carcass-based feeding events in the aviary (10 in total, Figure 1B)
162 using a remotely-controlled video camera (GoPro Hero 4, GoPro Inc.) fixed at 2 m height on
163 the aviary wall. These feeding events lasted on average 34 min (SD \pm 4 min).

164

165 We computed individuals' rank relying on the randomised Elo-rating approach [43,44],
166 which accounts for potential temporal instability of the rank using permutations in the agonistic
167 interaction series ('elo_scores' function, *aniDom* package, [44,45]; using 1000 randomisations
168 and fixing the rank adjustment speed along the series, K-factor, to 200). The interaction series
169 consisted in identifying the "wins" and "losses" for a given individual in agonistic interactions
170 [46] with other individuals in each video (annotated with BORIS video analysis software, [47]).
171 We used the ethograms from Bose & Sarrazin (2007, [30]) and Valverde (1959, [48]) to
172 characterise griffon vulture feeding behaviour and between-individual interactions. An
173 individual won the interaction when it interrupted another individual's feeding bout (by pecking
174 it, displacing it or engaging in a fight), and finally accessed the carcass before its opponent. In
175 other cases, the interaction was considered as a "loss" for the initiator. We assessed the
176 reliability of the dominance hierarchy through individual Elo-rating repeatability
177 ('estimate_uncertainty_by_repeatability' function, *aniDom* package, [44]).

178 2.1.3 Group flights

179 We recorded vulture flights decisions during 42 flight sessions (21 sessions each year)
180 in the vicinity of the Rocher des Aigles. In general, birds were released for a flight session three
181 times per day (in rare occasions from 2 to 4 times), at around 11:00, 14:30 and 16:00 (local
182 time) for a mean duration of 26.03 min (SD \pm 14.15 min) of flight. These captive vultures are
183 trained to fly freely, searching for thermals, gaining altitude and coming back to their trainers
184 (Supplementary Video 1). Vultures were equipped with a high-resolution GPS logger (4 Hz,
185 TechnoSmart, models Gipsy 1, Gipsy 5 or Axytreck) positioned at their lower back using a
186 Teflon leg-loop harness (Figure 1C, [49]). They were released in two groups, built according
187 to social preferences with the three most socially-bonded birds together, at 2-min intervals.
188 Release order alternated between consecutive days. For each flight session, we recorded and
189 considered as stable the cloudiness (i.e. the proportion of clouds covering the sky, on a scale
190 from 0 - no clouds - to 8 - sky fully covered by clouds), horizontal wind speed and temperature
191 (all extracted from meteofrance.com based on local meteorological models).

192 To further investigate how vulture thermal choices were shaped by personal and social
193 information, we pre-processed flight tracks in three consecutive steps. We subsampled
194 individuals' tracks from 4 to 1 GPS fix per second, and segmented their flight behaviour into
195 gliding, linear soaring and circular soaring. We then created spatio-temporally dynamic maps
196 of thermal availability based on the spatial clustering of individual's circular soaring phases.
197 Leaning on these maps, we retraced the history of thermal use/choice by individuals.

198 2.1.3.1 Thermal use identification

199 To segment vulture flight between circular soaring, linear soaring and gliding flight we
200 first calculated turning angle and vertical speed between consecutive locations using the *move*
201 R package [50]. We applied a moving window of 30 s to calculate the absolute cumulative sum
202 of the turning angles (hereafter cumulative turning angle) and a moving window of 5 s to
203 calculate the average vertical speed. We then applied a k-means approach ($k = 2$, ‘kmeans’
204 function, *stats* R package) on the smoothed vertical speed (positive speed when flying upwards)
205 to distinguish between soaring (ascending flight) and gliding (descending flight, [51,52]). We
206 further classified soaring locations into circular soaring (indicating use of thermal updrafts) and
207 linear soaring (also called slope soaring, expected to occur outside of thermals), with circular
208 soaring being associated with a cumulative turning angle ≥ 300 degrees. A result of
209 segmentation is illustrated in Figure 2B. Finally, we inferred the use of a thermal when the
210 individual engaged in circular soaring for more than 30 s, with no interruption of more than 5
211 s of gliding.

212 2.1.3.2 Dynamic mapping of available thermals

213 Within each flight session, we created a dynamic map of thermals (Figure 3). First, we
214 spatially clustered vulture circular soaring locations (reflecting the use of the same thermal
215 updraft) independently of time by using a 3D density-based spatial clustering approach
216 (‘dbscan’ function, *dbscan* R package, [53]). This algorithm relies on a spherical
217 neighbourhood to perform density-based neighbour joining, i.e. clustering (Figure 2C). We
218 assumed this neighbourhood to be of a 40-m radius, and a minimum number of five locations
219 within this range for the algorithm to consider the neighbourhood further. This 40 m threshold
220 corresponded to the largest 4-nearest-neighbour distance observed when considering locations
221 attributed to thermal use only (‘kNNdistplot’ function, *dbscan* R package) and matched with
222 empirical expectations of radius during circular soaring phases [54].

223 We then made those maps dynamic in time by considering the lifetime of each thermal.
224 We considered a thermal as “available” from the moment when the first individual entered it
225 until the last individual left it (Figure 3).

226 2.2 Statistical analyses

227 We defined collective flight events as any time of a flight session when at least two
228 individuals were flying. For each of these events, we first analysed the use of social information
229 (the tendency to join thermals already discovered by conspecifics) as a function of external

230 (meteorological) and internal drivers (individual traits). We then used step selection functions
231 to define, at each movement step, which drivers determined the selection of the chosen next
232 location (thermal updraft) relative to other potential locations.

233 2.2.1 Drivers of social information use

234 We investigated the effect of local meteorological context, individual traits and flight
235 mechanics on the use of social information, defined here as the tendency to join thermals
236 already discovered by conspecifics. We considered that an individual discovered a thermal
237 when it was the first, among all individuals, to adopt circular soaring flight into it. For the
238 analysis, we discarded the discovery of the first thermal in each flight session (as this thermal
239 was necessarily discovered).

240 To investigate the drivers underlying the use of social information we modelled the
241 probability to join a thermal already discovered by others using generalised linear mixed
242 models (GLMMs) with binomial error structure and a logit link function [55]. Our full model
243 contained the following ten fixed effects: **meteorological variables** with the (i) wind speed
244 (categorical predictor), (ii) cloudiness and (iii) temperature (both continuous predictors); **social**
245 **variables** with (iv) the age (continuous predictor) and (v) rank in the dominance hierarchy of
246 the individual (ordinal categorical predictor), and variables related to the **mechanic of flight**
247 with (vi) the glide-ratio (horizontal distance travelled during a 1-m altitude loss, only measured
248 on glides with straightness > 0.95 in each flight), (vii) the altitude of and (viii) the 3D distance
249 to the exit location from the previous thermal used (all continuous predictors). We also added
250 (ix) the group in which individuals have been released (first or second group released for the
251 flight) and (x) the time elapsed since the first individual take-off (continuous predictor) as
252 control variables. Individual ID was considered as a random factor.

253 To compare the relative importance of the fixed effects we scaled all non-categorical
254 variables to use their estimate as dimensionless effect size [56]. We examined the significance
255 of each variable by comparing the goodness of fit of models with and without the variable of
256 interest using a likelihood ratio test ('drop1' function, *stats* R package). Assumptions required
257 for these statistical approaches (homoscedasticity, Gaussian distribution of residuals) were
258 checked with plot diagnosis (histogram of residuals, residual Q-Q plot, distribution of residuals
259 vs fitted values, *DHARMA* R package, [57]). We also tested for the presence of outliers, and
260 calculated the variance inflation factor (VIF) to test for collinearity (VIF values ≥ 3 suggesting
261 a strong collinearity [58]). We did not detect collinearity in our predictors (VIF_{max} = 1.74)
262 (Figure S3). Furthermore, we extracted the marginal coefficient of determination (R_m^2) and the

263 conditional coefficient of determination (R_c^2) which describe, respectively, the proportion of
264 variance explained by fixed effects and by the fixed and random effects combined [59]. Finally,
265 as the flight time period, and the tested individuals differed, we cross-compared models fitting
266 on the two years separately (see Supplementary Material ESM01).

267 2.2.2 Drivers of thermal updraft selection

268 To study the drivers underlying thermal selection, we embedded our work in the Step
269 Selection framework [60]. We considered the series of thermals used by each individual. In
270 that series, we focused on movement steps involving a flight to a thermal previously (or
271 currently) used by a conspecific when other thermals were available. Using a conditional
272 logistic regression, we compared the “chosen” thermal characteristics to those “available” but
273 not chosen. The conditional logistic regression included seven predictors, respectively
274 characterising the **thermal profitability** with (i) the distance to it and (ii) maximum vertical
275 speed reached in the thermals by any individual since the focal individual has been released in
276 the flight session (continuous predictors), **individual personal experience** considering
277 whether (iii) the thermal was previously used by the focal individual (binary predictor), and
278 **social information** with (iv) the presence of the focal individual’s preferred affiliates in the
279 thermal or not (binary predictor), (v) the number of individuals present in the thermal, (vi) the
280 weighted mean (by the number of previous visits to the thermal) of the social bond with
281 individuals that used the thermals, and (vii) the negative cubed difference of ranks between the
282 focal individual and those in the thermals (all continuous predictors, set to 0 for the two latter
283 if no individuals used it/were present). We used the negative cubed difference to consider an
284 attraction-repulsion effect (high rank toward low rank and the opposite respectively) by
285 translating a linear rank difference to a relative hierarchy scale which enhances large rank
286 differences. For example, following the curve of the negative cube function, if the difference
287 of rank was five (e.g. the focal individual is ranked 6th - a low rank, a conspecific in another
288 thermal is ranked 1st - a high rank) the probability that the focal individual joined the
289 conspecific should be drastically decreased, mimicking a repulsion effect.

290 Also for this model, we scaled all non-categorical variables to better compare their
291 relative importance. We fitted the conditional regression considering all individuals together,
292 yet considering data stratified at the individual-step level. We finally reported the relative
293 selection strength (RSS) of significant variables which provides the magnitude of estimated
294 selection coefficients, holding all other covariates fixed [61,62].

295 3. Results

296 Vulture dominance hierarchy was steep (Figure S2) and reliably inferred (individual
297 Elo-rating repeatability = 0.82 and 0.83 in 2021 and 2022 respectively). The rank orders among
298 individuals present in both years were relatively consistent and unrelated to sex or age (Table
299 S1). During the 21 flight sessions performed each year, we identified a total of 520 and 578
300 thermalling events in 2021 and 2022. On average, 63% (SD \pm 7%, Table S1) of these circular
301 soaring behaviours took place in thermals discovered by a conspecific.

302 3.1 Flight risks and hierarchy shapes the use of social information

303 Our model was significantly better than the null model (considering only control
304 effects; AIC = 1237.4 and 1412.7 respectively) and explained 30% of the variance (Table S2).
305 The probability for an individual to use a thermal previously discovered by a conspecific
306 decreased with temperature (from 0.74 at 17°C to 0.43 at 31°C, Figure 4A, Figure 5A, Table
307 S2), but tended to increase with cloudiness and wind speed (Figure 4A, Table S2). This
308 probability dropped also with the distance from the previous thermal and the altitude at which
309 the bird left it (from 0.63 when being at a distance of 12 m from the last thermal used to 0.16
310 at a distance of 6776 m and from 0.76 when exiting the last thermal at an altitude of 195 m to
311 0.039 at 1574 m of altitude, Figure 4A, Figure 5B, C, Table S2). Individuals lower in the
312 dominance hierarchy were approximately twice as likely to discover new thermals than high-
313 ranked individuals (Figure 4A, Figure 5D). We did not detect significant effects of age and
314 glide-ratio on the probability to use thermal previously discovered by conspecifics (Figure 4A,
315 Table S2). Fitting the same model structure on 2021 and 2022 data separately yielded the same
316 overall results, suggesting that the observed pattern was robust to changes in hierarchy and
317 between-year conditions (Figure S4, Table S2).

318 3.2 Vultures select thermal updrafts hosting the most conspecifics

319 We identified 178 movement steps where an individual entered a thermal while at least
320 one other thermal was available simultaneously. Individuals were significantly more likely to
321 select thermals hosting the largest number of conspecifics at time of decision (Relative
322 Selection Strength [95% confidence interval] = 27.94 [5.99, 131.63]; Figure 4B, Table S3). On
323 the contrary, the probability to choose a thermal tended to decrease when the preferred affiliate
324 was using it. The distance to the previous thermal and the maximal vertical speed reached in
325 the thermal and whether individuals used this thermal in the past did not significantly affect

326 thermal selection (Figure 4B, Table S3). At time of decision, the difference in dominance ranks
327 as well as the presence of its preferred affiliate did not drive the individual's probability of
328 selecting the thermal. This pattern was consistent when considering only movement steps
329 where individuals had to choose between thermals used at time of decision (N = 61, Figure S5,
330 Table S3). Furthermore, considering all decision events, the sensitivity analysis on the inter-
331 individual distance threshold for the social bond strength estimation yielded the same results
332 (i.e. 1.55 m, 1.30 m and 1 m; see Supplementary Material ESM01, Figure S6, Table S4).

333 4. Discussion

334 Using a combination of high-resolution tracking and social structure monitoring, we
335 identified contextual drivers for the differential weighting of personal and social information
336 in movement decisions. We showed that vultures' movement decisions predominantly relied
337 on social information, especially in unfavourable flight conditions that increased thermal
338 unpredictability or put individuals at risk of undesired landing. Overall, individuals
339 preferentially joined thermals with the largest number of conspecifics. However, the use of
340 social information depended on the individual social status: low-ranking individuals were more
341 inclined to use personal information and discovered more thermals on their own than high-
342 ranking individuals.

343 We found that low-ranked individuals, likely the ones suffering the most from
344 interference competition, had higher probabilities of discovering new thermals, thus likely
345 exploring their environment more intensively than the high-ranked individuals. Such flight
346 strategy would enable subdominant individuals to reach carcasses first, or at least to arrive at
347 the beginning of the feeding event when the rate of interference is lower [27] hence avoiding
348 to loose opportunities due to conformity with conspecific behaviour [64]. From this may
349 emerge a producer-scrounger dynamic [65,66] wherein the use of personal information from
350 low-ranked individuals to arrive at food sources with lower competition levels would be
351 exploited by dominant individuals to reduce their own searching effort [16,66,67]. This is
352 coherent with previous observations of low-ranked vultures being “pioneers”: the very first
353 individuals to land and feed on the carcasses before being displaced by high-ranked individuals
354 arriving afterwards [27]. This influence of dominance on foraging tactics where low-ranked
355 individuals explore and find food while dominant profit has also been observed in other social
356 bird species such as common cranes *grus grus*, oystercatcher *Haematopus ostralegus*, house
357 sparrows, *Passer domesticus*, and barnacle goose, *Branta leucopsis*, [18,67–69]. Eviction of

358 subordinates from food patches has even recently been identified as a trigger for collective
359 movements [70]. In contrast, in activities where individuals do not experience competition,
360 such as tool-use learning in chimpanzees, naïve individuals will generally copy dominant (and
361 knowledgeable) individuals [71]. Our study hence stands as a clear cut illustration of the “copy
362 when asocial learning is costly” rule [72]: the vulture position in the dominance hierarchy,
363 through the costs it imposes on access to food, seems to calibrate the balance between the use
364 of personal and social information in foraging movements. In some cases, however, trading
365 personal information in favour of social information is inevitable.

366 When the environment is largely unpredictable or whenever using error-prone personal
367 knowledge can be energetically costly, individuals should tend to eavesdrop, and rely more on
368 information provided by conspecifics to reduce uncertainty about resources availability
369 [15,73]. Here, we evidenced both cases. First, vultures prioritised the use of social information
370 when the temperature was low and tended to when cloudiness and wind speed increased. These
371 weather conditions may translate into fewer and weaker thermals, drifting into the wind,
372 making them less predictable [74–78]. Second, they also favoured social information when the
373 altitude at which they left their previous thermal was low. When exiting a thermal at low
374 altitude, individuals have limited time to glide to the next thermal before having to shift to
375 flapping flight to stay aloft, or else landing in an undesired place, which both would add high
376 energetic cost associated with flapping and take-off [23,24,79,80]. Reaching high altitudes
377 quickly to avoid this risk may also explain why vultures used more thermals previously
378 discovered by conspecifics if those were close to the last thermal they used. While vultures are
379 able to cope with difficult flight conditions (e.g. turbulence and strong wind) by adjusting their
380 banking angles [55], anticipating such risky events may remain the most efficient way to
381 maximise the trade-off between time, energy and risk which largely dictates their flight strategy
382 [81]. Adult individuals, through experience, are generally better at coping with difficult flight
383 conditions [82], yet we did not evidence an effect of age relative to the use of social
384 information, as observed in other group living species (e.g. [83]). More than age *per se*, the
385 familiarity of individuals with a given situation might shape their tendency to rely or not on
386 social knowledge (e.g. in spider monkeys, *Ateles geoffroyi*, during collective foraging [84]).
387 The captive individuals tested in this experiment are all adults and fly in the same landscape
388 every day since their birth, thus they are probably very familiar with the areas favourable to
389 thermal emergence. This could explain why we did not detect any effect of age on the use of
390 social information, but also indicates that the relative importance of this source of information
391 is probably underestimated due to the birds' familiarity with the surroundings. An interesting

392 complementary experiment (though technically difficult) to disentangle the effect of familiarity
393 with the landscape would be to move the whole group and repeat the experiment in an
394 unfamiliar area to better estimate the strength of personal versus social information use.

395 When faced with a choice between simultaneously available thermals, the previous
396 experience of individuals (i.e. whether the thermal was used previously or not by the focal) or
397 current expertise of the group (i.e., relative age/hierarchy difference) impacted very little
398 vulture movement decisions compared to other social cues, contrasting with previous findings
399 from insects to mammals, including birds [85–89]. In the current system, ascending currents
400 can be very ephemeral phenomena, sometimes only lasting a few minutes [90,91]. Certainly, a
401 “live report” is therefore better provided by the accumulation of convergent information
402 sources (i.e. numerous conspecifics, [92]) rather than relying on a unique individual source (i.e.
403 the individual itself or one reference individual). In that line, and surprisingly, the presence of
404 one preferred affiliate in a thermal tended to reduce the probability to join it. There is evidence
405 that social bonds assessed “on the ground” are often unrelated to co-flight preferences [93]. It
406 therefore questions whether collective flights might be used by vultures to strengthen initially
407 weak social bonds. Maintaining in-flight bounds can indeed be important, as evidenced in the
408 migratory behaviour of other soaring bird species to enable accurate collective mapping of the
409 distribution of uplifts [94,95]. Furthermore, for soaring birds, the presence of conspecifics
410 should provide not only information on the location and strength of updrafts [20,95] but could
411 also indicate flight speed and circling radius needed to optimise climb rate, by remaining close
412 to the centre of the thermal where uplift is highest [55]. Yet, the maximum speed reached by
413 individuals using the thermal little affected vulture decision choices. Possibly, climb rate or
414 individual speed are not as easy to assess at a distance, compared to the number of conspecifics.
415 In other words, vultures tended to favour quantity signals (with the number of conspecifics)
416 over quality signals (maximal vertical speed) [96]. The “power of the group” may indeed in
417 turn drive cohesion, which could itself make social information even more profitable [96,97].

418 Altogether, our results provide insights into the architecture of decision-making
419 during movement in a social bird. It highlighted the trade-offs between personal and social
420 information these birds have to consider in order to optimise both their flying efficiency and
421 their foraging success. As a first approximation, we considered social cues as coming from
422 “conspecifics”. Strictly speaking however, our study included two species, Griffon vulture and
423 Rüppell’s vulture, albeit phylogenetically close and with similar biology. The one Rüppell’s
424 vulture in fact, used social information provided by surrounding vultures and did not stand out
425 as an outlier in its behaviour. It is known that even phylogenetically distant individuals could

426 be an important source of social information, not only about the presence of carcasses [98], but
427 also about the availability of thermals when sharing the same airspace (e.g. from black kites,
428 *Milvus migrans*, or common swifts, *Apus apus*, [99,100]). Interactions with heterospecifics can
429 indeed drastically affect animals' daily life [101], up to shaping the cognitive machinery
430 underpinning their foraging decisions [102]. How heterospecific cues are used when foraging
431 remains clearly overlooked. Future studies in this direction could provide valuable insights into
432 understanding the fundamental rules dictating how animals decide where to go.

433 CRediT authors' contributions

- 434 **Yohan Sassi**: Conceptualization, Methodology, Software, Investigation, Formal Analysis,
435 Visualization, Writing - original draft, Writing - Review & Editing.
436 **Basile Nousières**: Investigation, Resources.
437 **Martina Scacco**: Software, Writing - Review & Editing.
438 **Yann Tremblay**: Conceptualization, Writing - Review & Editing.
439 **Olivier Duriez**: Conceptualization, Investigation, Supervision, Writing - Review & Editing.
440 **Benjamin Robira**: Conceptualization, Methodology, Software, Visualization, Validation,
441 Supervision, Writing - original draft, Writing - Review & Editing

442 Declaration of competing interest

- 443 The authors declare to have no conflict of interest

444 Data availability

- 445 Scripts for review and supplementary video are available here:
446 <https://github.com/YohanSassi/UpdraftsDecisions>

- 447 A perennial storage of data and scripts will be provided after revision (e.g. Zenodo, Dryad)

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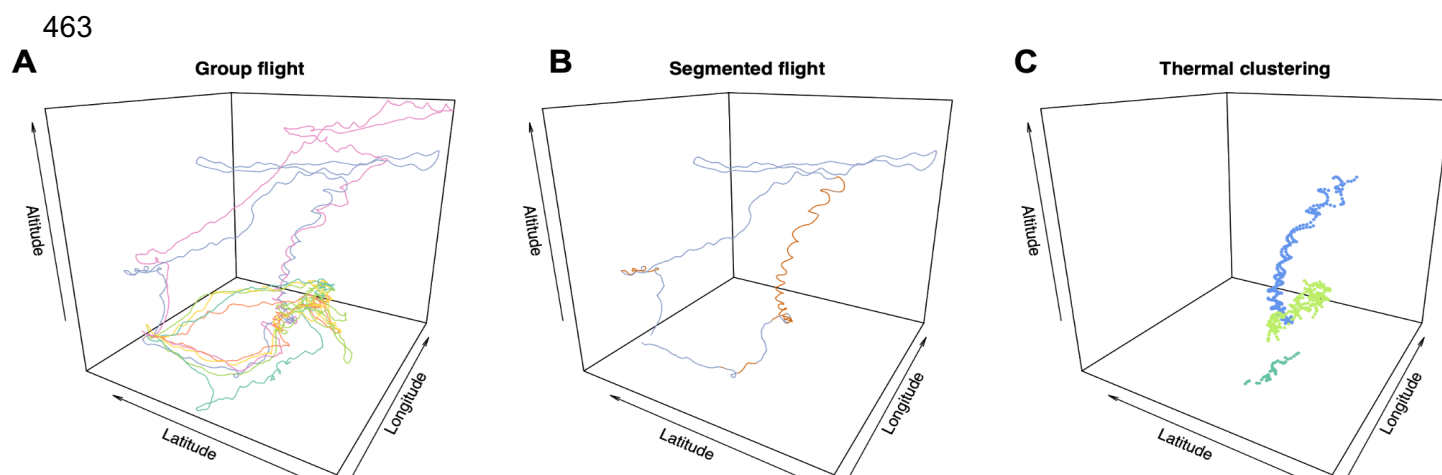
454 Fundings

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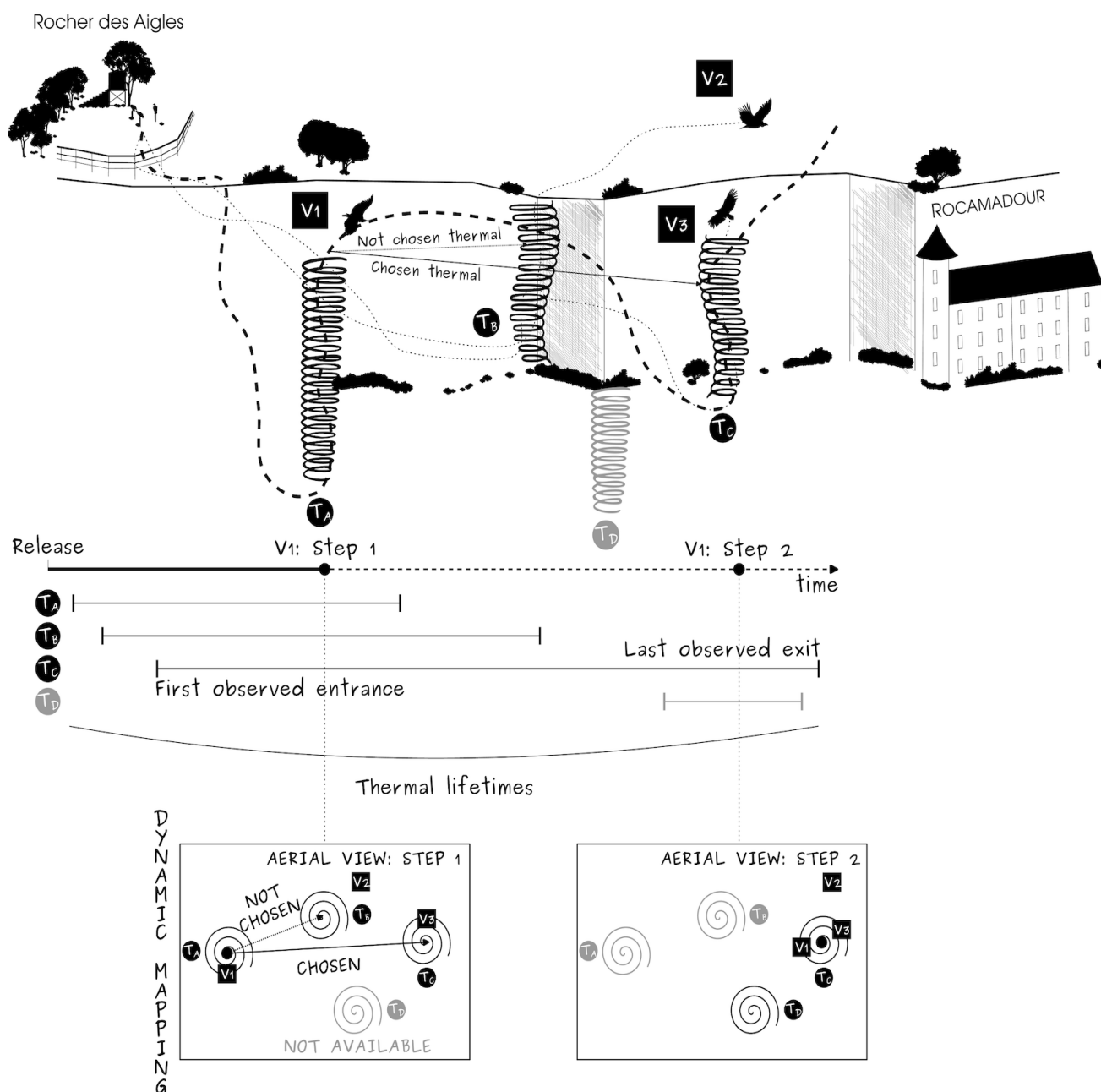
457 Main text figures



458 **Figure 1. Data collection.** (A) Perched vultures. Distance between vultures during perching
459 events were used to estimate social-bond strength. (B) Feeding event around a butchery carcass.
460 Agonistic interactions during those feeding events were used to estimate dominance hierarchy.
461 (C) Flying vulture. Vultures were released for free flight into a 120-m canyon, equipped with
462 high-resolution GPS loggers.

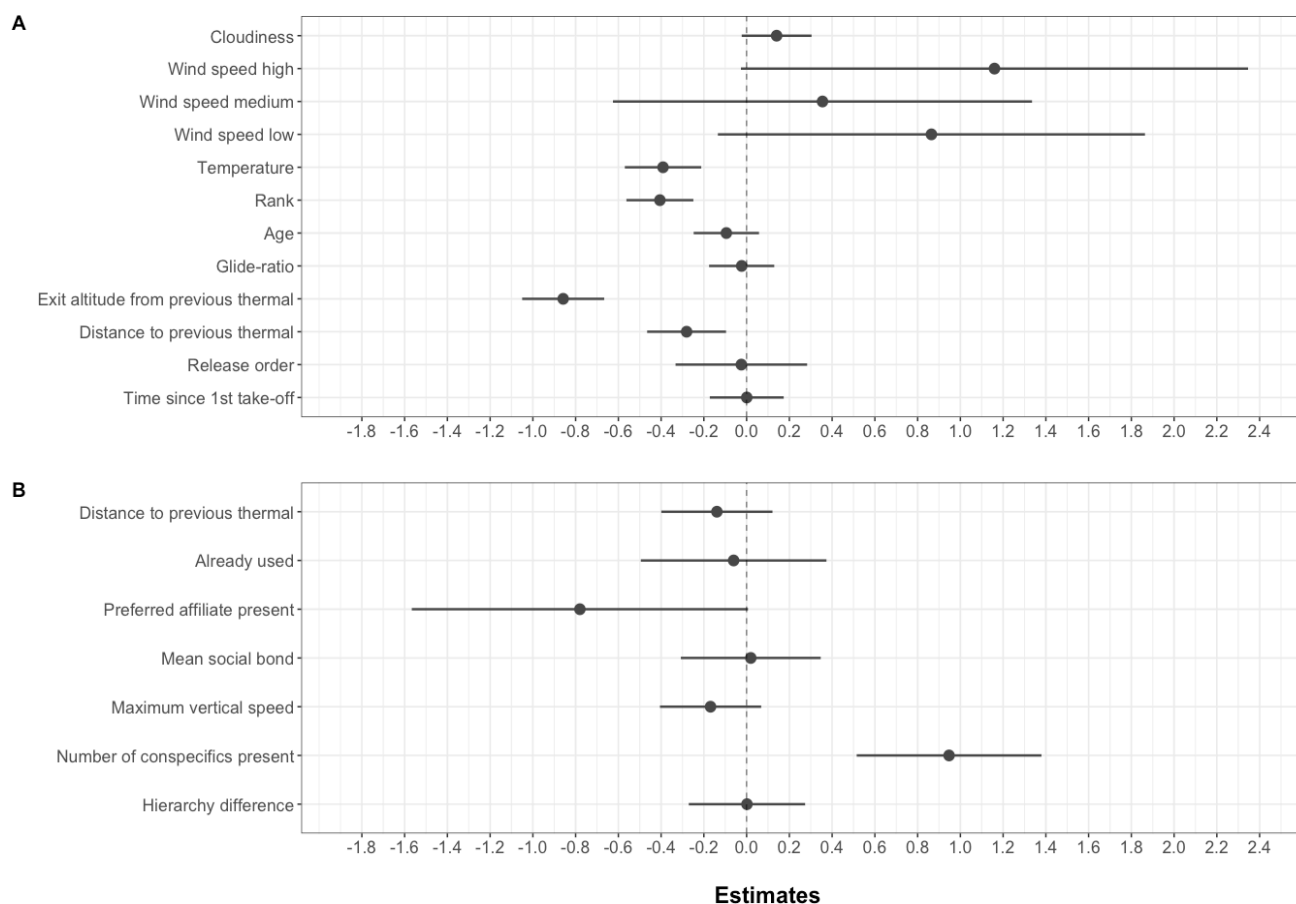


464 **Figure 2. Flight data pre-processing.** Pre-processing steps of group flight GPS data, example
465 of one flight session. The altitude ranges from 200 m to 600 m. (A) shows a group flight (see
466 Supplementary Video 1), with colours corresponding to each individual. (B) illustrates the
467 segmentation of an individual's flight (blue individual in (A)), with the orange segments
468 corresponding to circular soaring phases. (C) illustrates the 3D density-based spatial clustering
469 of individuals circular soaring phases, with colours indicating the three thermals identified in
470 this flight session.

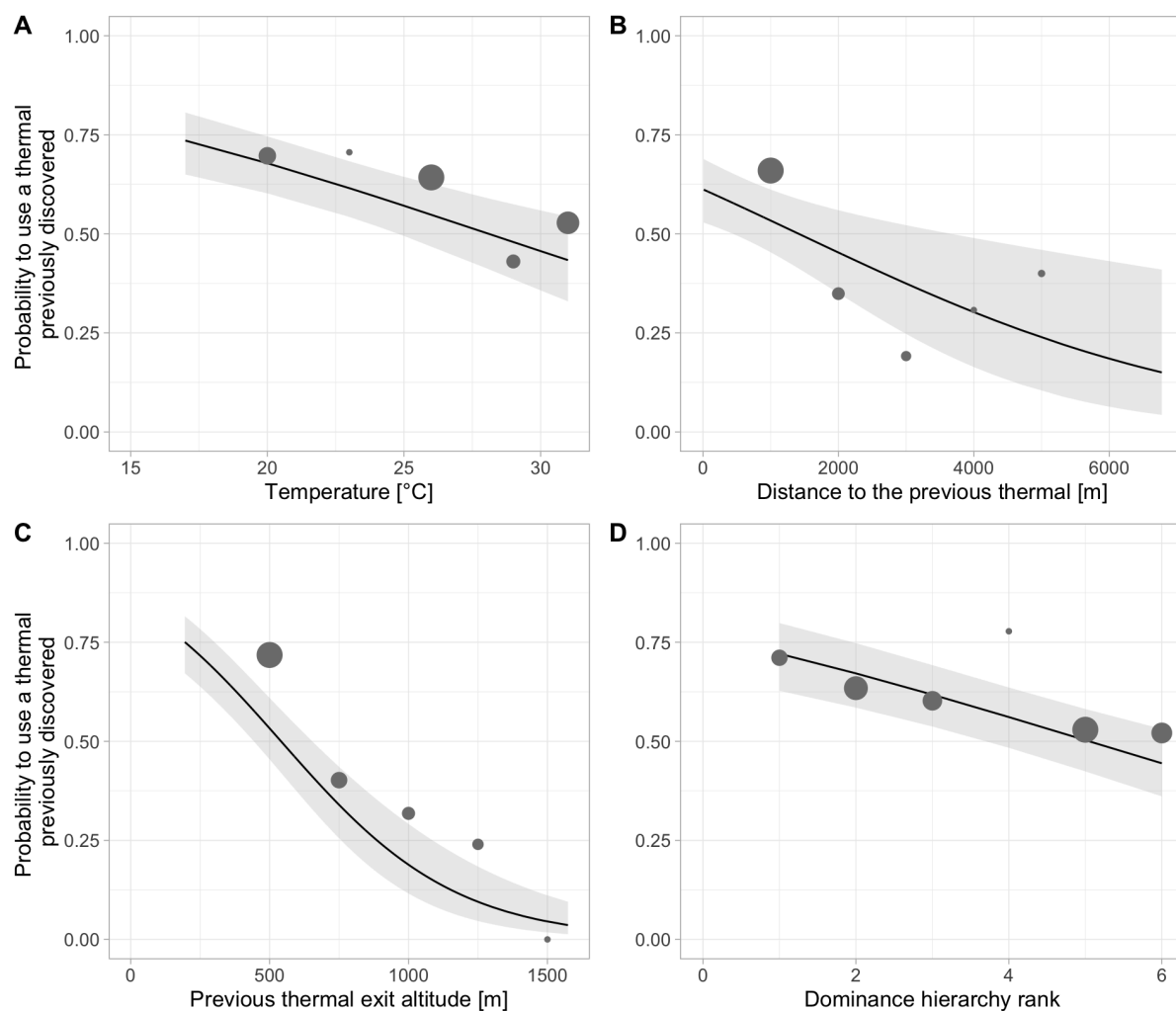


471 **Figure 3. Illustration of the step selection framework used to investigate thermal**
 472 **selection.** We focused on the movement of vultures released from the Rocher des Aigles when
 473 flying from thermal to thermal (i.e. a step). To do this, we mapped each thermal used during a
 474 flight session based on movement segmentation and clustering (see method section) to create
 475 dynamic maps of thermal availability over the flight session (as represented by the aerial
 476 views). The illustrated example focuses on the decision of a vulture (V₁; step 1) when leaving
 477 the thermal (T_A) and having to choose between two available thermals (T_B, close but not
 478 currently used by another vulture, and T_C, further away but currently used by another
 479 individual). T_D was not available until step 2, when it was discovered and used by another
 480 individual, and is therefore shown in grey at step 1. At step 2, V₁ joined V₃ in T_C and both

481 thermals T_A and T_B were no longer available. A thermal was available from the moment when
482 the first individual entered it until the last individual left it. Therefore, the number of available
483 thermals could change during the flight session (see differences between maps in step 1 and 2).



484 **Figure 4. Estimates of models investigating the drivers of social information use (A) and**
485 **thermal selection (B).** Rows correspond to each predictor. Each point represents the
486 standardised estimate value. Segments give the associated 95% confidence intervals.



487 **Figure 5. The probability to use a thermal already discovered by conspecifics decreases**
488 **with temperature, distance to the thermal, flight altitude and hierarchy rank.** Points
489 represent the probability of using a thermal already discovered by a conspecific, estimated on
490 the raw data. Their size is relative to the sample size. To do so, for (A), (B) and (C), predictors
491 were binned. Black lines with grey shades show the GLMM estimated probability with its 95%
492 confidence interval (N = 1098 thermals).

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