

1 **Urban birds' flight responses were unaffected by the COVID-19 shutdowns**

2

3 Peter Mikula^{1*}, Martin Bulla^{2,3}, Daniel T. Blumstein⁴, Yanina Benedetti², Kristina Floigl²,
4 Jukka Jokimäki⁵, Marja-Liisa Kaisanlahti-Jokimäki⁵, Gábor Markó⁶, Federico Morelli^{2,7},
5 Anders Pape Møller^{8,9}, Anastasiia Siretckaia², Sára Szakonyi¹⁰, Michael A. Weston¹¹, Farah
6 Abou Zeid², Piotr Tryjanowski^{12†}, Tomáš Albrecht^{1,13†}

7

8 ¹Institute of Vertebrate Biology, Czech Academy of Sciences, Květná 8, 60365, Brno, Czech
9 Republic

10 ²Faculty of Environmental Sciences, Czech University of Life Sciences Prague, Kamýcká 129,
11 165 00 Prague, Czech Republic

12 ³Department of Behavioural Ecology and Evolutionary Genetics, Max Planck Institute for
13 Ornithology, Eberhard Gwinner Str. 7, 82319 Seewiesen, Germany

14 ⁴Department of Ecology and Evolutionary Biology, University of California, 621 Young Drive,
15 South, Los Angeles CA 90095–1606, USA

16 ⁵Arctic Centre, University of Lapland, PO Box 122, 96101, Rovaniemi, Finland

17 ⁶Department of Plant Pathology, Institute of Plant Protection, Hungarian University of
18 Agriculture and Life Sciences, Ménesi út 44, 1118, Budapest, Hungary

19 ⁷Institute of Biological Sciences, University of Zielona Góra, Prof. Z. Szafrana St. 1, 65-516,
20 Zielona Góra, Poland

21 ⁸Ecologie Systématique Evolution, Université Paris-Sud, CNRS, AgroParisTech, Université
22 Paris-Saclay, F-91405, Orsay Cedex, France

23 ⁹Ministry of Education Key Laboratory for Biodiversity Sciences and Ecological Engineering,
24 College of Life Sciences, Beijing Normal University, 100875, Beijing, China

25 ¹⁰Department of Ecology, Institute of Biology, University of Veterinary Medicine Budapest,
26 Rottenbiller u. 50., H-1077, Budapest, Hungary

27 ¹¹School of Life and Environmental Sciences, Deakin University, Burwood Campus, 221
28 Burwood Highway, VIC 3125, Burwood, Australia

29 ¹²Institute of Zoology, Poznań University of Life Sciences, Wojska Polskiego 71C, 60-625,
30 Poznań, Poland

31 ¹³Department of Zoology, Faculty of Science, Charles University, Viničná 7, 12844, Prague,
32 Czech Republic

33

34 *Author for the correspondence: petomikula158@gmail.com

35 †These authors jointly supervised this work

36

37 ORCID:

38 Peter Mikula: <https://orcid.org/0000-0002-2731-9105>

39 Martin Bulla: <https://orcid.org/0000-0002-1328-1927>

40 Yanina Benedetti: <https://orcid.org/0000-0003-1600-2310>

41 Daniel T. Blumstein: <https://orcid.org/0000-0001-5793-9244>

42 Jukka Jokimäki: <https://orcid.org/0000-0002-7903-4128>

43 Gábor Markó: <https://orcid.org/0000-0003-1351-4070>

44 Federico Morelli: <https://orcid.org/0000-0003-1099-1357>

45 Piotr Tryjanowski: <https://orcid.org/0000-0002-8358-0797>

46 Mike Weston: <https://orcid.org/0000-0002-8717-0410>

47 Tomáš Albrecht: <https://orcid.org/0000-0002-9213-0034>

48 **Abstract**

49 The coronavirus disease 2019 (COVID-19) pandemic has dramatically altered human activities,
50 potentially relieving human pressures on urban-dwelling animals. Here, we evaluated whether
51 birds from five cities in five countries (Czech Republic – Prague, Finland – Rovaniemi,
52 Hungary – Budapest, Poland – Poznan, and Australia – Melbourne) changed their tolerance
53 towards human presence (measured as flight initiation distance) during the COVID-19
54 shutdowns. We collected 6369 flight initiation distance estimates for 147 bird species and found
55 that birds tolerated approaching humans to a similar level before and during the COVID-19
56 shutdowns. Moreover, during the shutdowns, bird escape behaviour did not consistently change
57 with the level of governmental restrictions (measured as the stringency index). Hence, our
58 results indicate that birds do not flexibly and quickly adjust their escape behaviour to the
59 reduced human presence; in other words, the breeding populations of urban birds examined
60 might already be tolerant of human activity and perceive humans as relatively harmless.

61

62 **Keywords:** antipredator behaviour; lockdown; escape distance; habituation; human-induced
63 rapid environmental change; urbanisation

64

65 **1. Introduction**

66 The actions taken to control the coronavirus disease 2019 (hereafter COVID-19) pandemic have
67 locked inhabitants in their dwellings and thus changed the pattern of human outdoor activities
68 (Fang *et al.* 2020; Gatto *et al.* 2020; Huang *et al.* 2020; Kraemer *et al.* 2020; Randler *et al.*
69 2020; Venter *et al.* 2020a, b; March *et al.* 2021). This situation created a quasi-experiment that
70 offers a unique opportunity to study how rapid changes in human behaviour affect wildlife
71 (Derryberry *et al.* 2020; Manenti *et al.* 2020; Rutz *et al.* 2020; Zellmer *et al.* 2020; Soga *et al.*
72 2021; Soto *et al.* 2021). The shifts in human activity linked to the COVID-19 pandemic were
73 tangible particularly in urban areas (Rutz *et al.* 2020; Park *et al.* 2021). Consequently, the
74 COVID-19 shutdowns have elicited a complex mixture of positive and negative effects on
75 urban nature (Bates *et al.* 2021), including changes in animal behaviour (Derryberry *et al.* 2020;
76 Montgomery *et al.* 2021; Vardi *et al.* 2021).

77 The coexistence of wild animals with humans requires animals to tolerate human presence and
78 consequent disturbance (Samia *et al.* 2015; Ducatez *et al.* 2020). Nevertheless, urban-dwelling
79 animals are more tolerant of human approach than their rural conspecifics (Díaz *et al.* 2013;
80 Samia *et al.* 2015, 2017; Mikula *et al.* 2021). Yet, how the intensity of human presence (e.g.
81 human population or pedestrian density) precisely shapes the escape behaviour of urban animals
82 is less clear. While some studies have revealed that tolerance to humans increased with
83 increasing human presence at studied sites (Webb & Blumstein 2005; Mikula 2014), other
84 studies reported much weaker, none or even opposite effects (Morelli *et al.* 2018;
85 Gnanapragasam *et al.* 2021; Mikula *et al.* 2021). Whether the COVID-19 shutdowns induced
86 changes in human activities which then altered the urban-landscape of fear is unknown.

87 The COVID-19 shutdowns locked people at homes and reduced their mobility within cities
88 (Geng *et al.* 2021; Wu *et al.* 2021; but see Venter *et al.* 2020b) and thus created a natural
89 experimental condition which could have affected the tolerance of urban animals towards
90 humans (usually quantified as flight initiation or escape distance) in three ways. First, the lack
91 of or reduced human presence during shutdowns might result in decreased animal tolerance to
92 humans and consequently increased escape distances because (a) some studies have reported
93 that animal tolerance decreases on the gradient from heavily human-visited (“disturbed”) sites
94 to sites with low human activity (Webb & Blumstein 2005; Mikula 2014) and (b) escape
95 distances of urban birds may also increase if less tolerant individuals and species colonise or at
96 least increase their detectability in human-emptied landscapes (Manenti *et al.* 2020; Gordo *et*
97 *al.* 2021). Second, due to reduced human activity, animal vigilance (i.e. the time spent

98 monitoring for potential threats) might decrease, which in turn could increase animal tolerance
99 to human presence because animal vigilance is often positively correlated with escape distance
100 (Samia *et al.* 2013; Samia & Blumstein 2015), leading to reduced escape distances during the
101 shutdowns (Montgomery *et al.* 2021). Third, animal responses to human disturbance may not
102 change during the shutdowns if animals are already habituated to human activity and generally
103 consider humans as harmless. Indeed, keeping escape distances constant may reduce
104 unnecessary responses even when encounters are less frequent or reduction of already short
105 escape distances may be constrained). In addition, if animals changed their escape distance
106 during COVID-19 shutdowns, we expect that the change would follow the levels of
107 governmental restrictions (stringency index) since the stringency index was negatively
108 associated with human use of urban parks (Geng *et al.* 2020).

109 Here, we explored variation in avian tolerance towards humans before (breeding seasons before
110 March 2020) and during the COVID-19 shutdowns (breeding seasons 2020–2021), and whether
111 the tolerance to humans during the COVID-19 shutdowns changed according to the level of
112 governmental restrictions. Avian tolerance towards human disturbance was measured as the
113 flight initiation distance, the distance from an approaching human observer at which bird starts
114 to escape (Albrecht & Klvaňa 2004; Stankowich & Blumstein 2005; Blumstein 2006; Weston
115 *et al.* 2012; Díaz *et al.* 2013) and measured in four European cities (the Czech Republic –
116 Prague; Finland – Rovaniemi; Hungary – Budapest; Poland – Poznan) and one in Australia
117 (Melbourne).

118

119 **2. Materials and Methods**

120 *2.1. Study areas*

121 Flight initiation distances were collected in the Czech Republic (~90% of data collected by
122 P.M.; the rest were collected by F.M., Y.B., K.F., A.S. and F.A.Z.), Finland (J.J. and field
123 assistant), Hungary (G.M. and S.S.), Poland (P.T.), and Australia (M.W. and field assistants).
124 All field data were collected during the breeding season (Europe: 1 April – 1 August; Australia:
125 15 August – 15 March) and only in urban areas, i.e. areas with continuous urban elements,
126 including multi-story buildings, family houses, or roads, with built-up area >50%, building
127 density >10/ha, residential human density >10/ha (Marzluff *et al.* 2001). Most data were
128 collected in urban green areas, particularly parks and cemeteries. Czech data were collected in

129 Prague (50.083°N, 14.417°E; 1.3 million inhabitants, 177–399 m a. s. l.), Finnish in Rovaniemi
130 (66.5°N, 25.733°E; 64,000 inhabitants, 75–203 m a. s. l.), Hungarian in Budapest (47.498°N,
131 19.041°E; 1.8 million inhabitants, 96–527 m a. s. l.); Polish in Poznań (52.406°N, 16.925°E;
132 0.53 million inhabitants, 60–154 m a. s. l.); and Australian in Melbourne (37.821°S, 144.961°E;
133 5.2 million inhabitants, 5–169 m a. s. l.). For each city, we collected data for two breeding
134 seasons before the pandemic, covering the 2018 and 2019 seasons immediately preceding the
135 emergence of the COVID-19 (Finland, Hungary, Poland; and until March 2020 in Australia)
136 and for up to two breeding seasons during the COVID-19 shutdowns (seasons starting in 2020
137 and 2021, for the Czech Republic, only starting in 2021). The 2019 data were not available for
138 the Czech Republic, and thus we used 2014 and 2018. The flight initiation distances were
139 collected during favourable weather conditions (i.e. no rain and no strong wind) mainly during
140 the weekdays and during the early mornings (till 10:00).

141

142 2.2. *Avian tolerance towards humans*

143 Avian tolerance towards human approaches (“disturbance”) was estimated by a simple but
144 widely used method, the flight initiation distance, i.e. the distance at which birds escape when
145 approached by a human observer (Stankowich & Blumstein 2005; Blumstein 2006; Weston *et*
146 *al.* 2012; Díaz *et al.* 2013). The flight initiation distance reflects a trade-off between the fitness-
147 related benefits of not escaping and the costs of fleeing (Ydenberg & Dill 1986; Albrecht &
148 Klvaňa 2004; Stankowich & Blumstein 2005; Samia *et al.* 2016). The flight initiation distance
149 estimates are highly consistent for individuals, populations, and species tested within similar
150 contexts (Carrete & Tella 2010; Díaz *et al.* 2013; Guay *et al.* 2016; Mikula *et al.* 2018).

151 All data were collected by trained researchers skilled in bird identification, and using a standard
152 procedure outlined previously (Blumstein 2006; Samia *et al.* 2015; Mikula *et al.* 2018, 2021).
153 Briefly, when a focal bird was spotted, a single observer moved at a normal walking speed (~1
154 ms⁻¹) directly towards the bird (with head and gaze oriented towards this bird). When the focal
155 bird first started to escape (i.e. hopped, walked, ran, or flew away), the distance of the observer
156 to the bird was noted. The escape distance was measured either by counting the number of ~1
157 m long steps, counting the number of steps of known approximate length and converting them
158 to metres or using a rangefinder (with ±1 m resolution). The escape distance of birds positioned
159 above the ground (e.g. perching on vegetation) was estimated as the Euclidean distance that
160 equals the square-root of the sum of the squared horizontal distance and the squared height

161 above the ground. We approached only individuals without initial signs of distress. Birds often
162 occur in flocks; in these cases, we randomly selected a single individual from a flock and
163 measured its response. All fieldworkers wore outdoor clothes without any bright colours.
164 Within each city we collected data at many sites (at the level of park, cemetery, etc.); to avoid
165 repeated sampling of the same individuals, we did not re-sample the same location during the
166 same breeding season. Within the sampling occasion at a given site, the same species
167 individuals were sampled only if it was obvious that they represent different individuals (e.g.
168 because of their concurrent presence or if morphological features, e.g. sex-specific colouration
169 or age, enabled us to distinguish between different individuals). In total, we collected 6369
170 flight initiation distance estimates for 147 bird species representing 2693 before-shutdowns
171 estimates for 68 species and 3676 during-shutdowns estimates for 135 species.

172

173 *2.3. Predictors and covariates*

174 Each observation was scored as collected before (0) or during (1) the COVID-19 shutdowns
175 (hereafter “Period”).

176 To further explore whether changes in human activity during the shutdowns affected avian
177 escape behaviour, we extracted data for each country and day on the strength of governmental
178 measures characterised by governmental stringency index (hereafter “stringency index”) from
179 Our World in Data database (<https://ourworldindata.org/covid-stringency-index>, based on data
180 originally published in Hale et al. (2021)). This index is rescaled to values from 0 to 100 (0 =
181 no restrictions; 100 = strictest restrictions) and represents a composite measure based on nine
182 response indicators, including school closures, workplace closures, cancellation of public
183 events, restrictions on public gatherings, closures of public transport, stay-at-home
184 requirements, public information campaigns, restrictions on internal movements, and
185 international travel controls.

186 Life-history, social, contextual and environmental factors may influence escape responses of
187 birds and potentially confound associations between avian escape responses and changes in
188 human outdoor activity before and during the COVID-19 shutdowns. Hence, we extracted
189 information on seven parameters. (1) The ‘starting distance’ (Blumstein 2006; Weston *et al.*
190 2012; Mikula *et al.* 2018) was estimated as the distance to the bird (in metres) when an observer
191 started the escape distance trial. (2) The ‘flock size’ (Samia *et al.* 2015; Mikula *et al.* 2018;

192 Morelli *et al.* 2019) was calculated as the number of all conspecific individuals moving,
193 feeding, or perching together that were visually separated from other conspecific or mixed-
194 species individuals. Note that we avoided approaching mixed-species bird groups. (3) The
195 species-specific ‘body size’ (Stankowich & Blumstein 2005; Díaz *et al.* 2013) was
196 approximated as body mass (in grams) and obtained as the mean of female and male values
197 from EltonTraits 1.0 database (Wilman *et al.* 2014). (4) The ambient ‘temperature’ (Reynolds
198 *et al.* 2020) was estimated as the air temperature (°C) at the site during the data collection. (5)
199 The ‘time’ of data collection (in hours) (Piratelli *et al.* 2015) was rounded to the nearest hour.
200 (6) The date of data collection (Legagneux & Ducatez 2013; Piratelli *et al.* 2015; Mikula *et al.*
201 2018) was noted as a ‘day’ since the start of the breeding season (Europe: Day 1 = 1 April;
202 Australia: Day 1 = 15 August). (7) The ‘site’ represents a unique identifier of each sampled
203 park, cemetery, the city district, etc.

204

205 2.4. Statistical analyses

206 We used R version 4.3.0 (R Development Core Team 2021) for all statistical analyses. We used
207 two sets of mixed effect models.

208 First, we explored the differences in escape behaviour of birds before and during the COVID-
209 19 shutdowns by fitting the flight initiation distance (ln-transformed) as a response variable and
210 a Period (0 – before, 1 – during shutdowns) as a predictor of interest, while controlling for
211 starting distance (ln-transformed), flock size (ln-transformed), temperature (also a proxy for a
212 day within the breeding season: $r_{\text{Pearson}} = 0.48$; Fig. S1) and time of day. To account for circular
213 properties of time, time was transformed into radians ($2 \times \text{time} \times \pi/24$) and fitted as sine and
214 cosine of radians (Bulla *et al.* 2016). All continuous variables were standardised by subtracting
215 the mean and dividing by the standard deviation. Potential multicollinearity among explanatory
216 variables was checked by the correlation matrix, which suggested that correlations between
217 variables were generally weak (Fig. S1). To account for the non-independence of data points
218 (Schielzeth & Forstmeier 2009; Barr *et al.* 2013), we attempted to fit random intercepts of year,
219 genus, species, species at a given day and year, country, site, and species within a site, while
220 fitting Period as random slope within all random intercept, except for year and species at given
221 day and year. We then simplified the random structure of the model to avoid singular fit (Barr
222 *et al.* 2013), i.e. a situation when one or more random effects are estimated as zero or close to
223 zero. The outcomes of the models with alternative random structures were similar (Fig. S2;

224 Table S1). We used this approach with a full dataset with all observations ($n = 6369$), as well
225 as with conservative datasets, one with at least five observations per species and Period (i.e. at
226 least five observations before and five during the COVID-19 shutdowns; $n = 5260$), the other
227 with at least 10 observations per species and each Period ($n = 5106$).

228 Second, we explored whether the escape behaviour of birds during the COVID-19 shutdowns
229 was modulated by the varying levels of governmental restrictions, the stringency index. We
230 thus used only data gathered during the shutdowns (i.e. Period = 1; $n = 3676$) and adopted the
231 same models and the same random structure simplification procedure as for the first set of
232 models on Period, but using stringency index instead of Period and year as fixed (not as random)
233 effect. Similarly, as in the models for Period, varying the random structure did not influence
234 the outcomes of the models, nor did limiting species to only those with at least five or at least
235 10 observations (Fig. S2; Table S2).

236 The models were fitted with the *lme4* package (Bates *et al.* 2015). We then used the *sim* function
237 from the *arm* package and a non-informative prior distribution (Gelman & Hill 2007; Gelman
238 & Su 2018) to create a sample of 5,000 simulated values for each model parameter (i.e. posterior
239 distribution). We report effect sizes and model predictions by the medians, and the uncertainty
240 of the estimates and predictions by the Bayesian 95% credible intervals represented by the 2.5
241 and 97.5 percentiles (95% CI) from the posterior distribution of 5,000 simulated or predicted
242 values. We graphically inspected the goodness of fit, and the distribution of the residuals (see
243 Bulla *et al.* 2022).

244 As the need for phylogenetic control depends on the phylogenetic signal in the residuals of the
245 model (Uyeda *et al.* 2018), we tested whether the residual variance contained a phylogenetic
246 signal. Thus, we extracted the residuals from the model on full data, presented in the main text
247 (Fig. 1 – all data estimates; Period: Table S1 – model 1d, Stringency index: Table S2 – model
248 1c), and fitted the residuals as a new response variable in an intercept-only Bayesian linear
249 regression fitted with STAN (Stan Development Team 2022) using *brm* function from *brms*
250 package v. 2.17 (Bürkner 2017, 2018), with species and their phylogenetic relatedness as
251 random effects. The phylogenetic relatedness was included as a phylogenetic covariance matrix
252 calculated with *inverseA* function in the *MCMCglmm* v. 2.33 package (Hadfield 2010) from
253 the maximum credibility tree build using *maxCladeCred* function in the *phangorn* v. 2.8.1
254 package (Schliep 2011) and 100 randomly sampled species-level phylogenetic trees (Hackett
255 backbone) from BirdTree online tool (<http://birdtree.org>) (Jetz *et al.* 2012). Priors were

256 specified using the *get_prior* function from *brms*, which uses Student's t distribution for the
257 intercept and standard deviation (Bürkner 2017). The target average proposal acceptance
258 probability was increased to 0.99 to make the sampling more conservative to posterior
259 distributions with high curvature (Bürkner 2017). Five MCMC chains ran for 5,000 iterations
260 each while discarding the first 2,500 iterations as burn-in, and sampling every 5th iteration,
261 which resulted in 5,000 samples of model parameters. The independence of samples in the
262 Markov chain was assessed using graphic diagnostics and the convergence using the Gelman-
263 Rubin diagnostics which was 1 for all parameters, indicating model convergence (Brooks &
264 Gelman 1998). Phylogeny explained zero (for Period model, 95%CI: 0–0.2% of variance in
265 random effects) or little variation (3.6% [0–16%] for a Stringency index model) and the models
266 without phylogeny fitted the data on residuals better than the models with phylogeny (i.e. the
267 estimated Bayes factor in favour of non-phylogenetic model was 155 for Period and 100 for
268 Stringency index, their posterior probabilities 0.995 and 0.986 respectively), which justifies our
269 use of non-phylogenetic comparative methods.

270 All results are reproducible with the open-access data and code available from Bulla *et al.*
271 (2022), which also provides visual representations of model assumptions.

272

273 **3. Results**

274 We found no consistent differences in avian tolerance towards humans (i.e. in their escape
275 distance) before and during the COVID-19 shutdowns (Figs 1a, 2, S2–S4; Table S1). Moreover,
276 during the shutdowns, we found no clear changes in escape distance in relation to the strength
277 of governmental COVID-19 restriction, i.e. stringency index, (Figs 1b and 3; Table S2). These
278 results were robust to changes in random effects structure and sample size (Fig. S2; Tables S1
279 and S2). The within and between species responses highly varied and changed within and
280 between species, as well as within and between sites and years (Figs 2 and 3, Figs S3 and S4).

281

282 **4. Discussion**

283 Our observations of avian tolerance towards humans (measured as escape distance) revealed
284 that populations of urban birds did not exhibit major shifts in their tolerance towards humans
285 (a) between periods before and during the COVID-19 shutdowns and (b) as a function of the
286 stringency of governmental COVID-19 restrictions. Responses to both the shutdowns and the

287 stringency of governmental restrictions varied widely across spatial and temporal scales.
288 Altogether, our results indicate that escape responses of sampled breeding populations of urban
289 birds are unaffected by temporal changes in human presence and abundance at studied sites.
290 This is consistent with some degree of urban tolerance not being further modified by even
291 dramatic changes in human activity patterns and could reflect the (1) differential settlement of
292 bold and shy individuals in and outside cities, (2) natural selection on urban tolerance, or (3)
293 habituation-like processes reducing plasticity.

294 First, the escape distance of birds is an individual trait, i.e. flight initiation distance is highly
295 consistent within individuals tested under similar contexts (Carrete & Tella 2010; Díaz *et al.*
296 2013; Guay *et al.* 2016; Mikula *et al.* 2018). If the urban environment filters birds based on
297 inherent levels of tolerance towards human disturbance with bold and shy individuals inhabiting
298 areas with varying levels of human disturbance (Carrete & Tella 2010, 2013; van Dongen *et al.*
299 2015), the lack of consistent change in escape behaviour of urban birds during the COVID-19
300 shutdowns may indicate an absence or generally low influx of shy, less tolerant individuals and
301 species from rural areas into studied cities. Although some studies have reported animal species
302 that occupied new areas changed their local abundance, or shifted the timing of their main
303 activity in response to change in human activities during the COVID-19 pandemic, these
304 changes are often area- or species-specific and did not occur everywhere (Manenti *et al.* 2020;
305 Bates *et al.* 2021; Gordo *et al.* 2021; Vardi *et al.* 2021). Importantly, it remains to be shown
306 whether such “changes” in animal behaviour are real or an artefact of increased activity of
307 researchers and citizen scientists in cities during the COVID-19 pandemic (Basile *et al.* 2021;
308 Crimmins *et al.* 2021; Hochachka *et al.* 2021; Randler *et al.* 2021).

309 Second, another possibility is that the lack of change in avian tolerance towards humans is a
310 result of an adaptation to urban lifestyle, i.e. evolutionary change in escape response due to
311 natural selection on urban tolerance (Symonds *et al.* 2016). Species entering urban
312 environments have larger brains (Sayol *et al.* 2020), greater innovation and problem-solving
313 abilities (Audet *et al.* 2016; Ducatez *et al.* 2017; Griffin *et al.* 2017), are less neophilic and
314 habituate to novel conditions faster than rural birds (Tryjanowski *et al.* 2016; Vincze *et al.*
315 2016; Griffin *et al.* 2017). Indeed, birds that colonised urbanised areas earlier are more tolerant
316 toward humans (Symonds *et al.* 2016). Yet, it is possible that avian tolerance reaches limits
317 (e.g. in an extreme case, not escaping would translate to being stepped upon) after several
318 generations living in cities. Indeed, escape responses persist in even the most habituated birds.
319 Our results may thus indicate that natural selection on escape behaviour of these species living

320 in the cities is currently at an optimum; hence, the changes in human presence associated with
321 COVID-19 shutdowns were insufficient and too temporary to select for different escape
322 distances.

323 Third, our results may indicate that the examined breeding populations of birds were already
324 habituated to or otherwise tolerant of human presence and abundance and humans might have
325 been generally perceived as harmless, irrespective of the number of people outside (Ellenberg
326 *et al.* 2009; Rankin *et al.* 2009; Zaccaroni *et al.* 2010). Previous studies have found that changes
327 in human density and activity may strongly influence escape behaviour of animals, especially
328 when working with dichotomic comparison of natural (rural) and human-inhabited (urban)
329 areas (Díaz *et al.* 2013). For example, neotropical birds occurring close to small and recently
330 established human settlements, such as farms, have dramatically shorter escape distances (i.e.
331 higher tolerance towards humans) than their counterparts living further from these settlements
332 (Tryjanowski *et al.* 2020). However, studies exploring associations between escape behaviour
333 of animals and various human activity regimes within urban areas revealed mixed results,
334 indicating that within urban areas association between escape behaviour of animals and human
335 density is often absent or weak (Morelli *et al.* 2018; Mikula *et al.* 2021). Such a finding, along
336 with ours, contradict the expectation that “...we expect location-specific habituation where even
337 a slight deviation in a predator’s (or human’s) routine behaviour can re-elicite fearful responses”
338 (Samia *et al.* 2015). Our results indicate a minor role of behavioural plasticity in observed
339 patterns and may indicate that habituation-like processes have already reduced plasticity.

340 Finally, worldwide, quarantine and stay-at-home orders have reduced the use of public spaces,
341 particularly in cities. Yet, the changes in the use of public spaces during COVID-19 shutdowns
342 seem to be country specific and dependent on the type of the public space (Derryberry *et al.*
343 2020; Randler *et al.* 2020; Rutz *et al.* 2020; Venter *et al.* 2020b; Geng *et al.* 2021). For example,
344 the restrictions on social gathering and movement, as well as the closures of workplaces and
345 indoor recreational facilities were associated with increased visits to parks in Norway (Venter
346 *et al.* 2020b). However, on a global scale stay-at-home restriction and the stringency of
347 governmental restrictions was negatively correlated with park visitation (Geng *et al.* 2020).
348 Thus, the lack of a general change in escape distances of birds during COVID-19 shutdowns in
349 our sample might be due to heterogeneous effects of country-specific governmental measures
350 on human densities across cities and their outdoor space. We note that country explained up to
351 15% in escape distance during the COVID-19 shutdowns (Table S2), country and site explained
352 little variance in the escape distance before and during the COVID-19 shutdowns (Table S1),

353 escape distance changed stochastically across years and species (Fig. S4) and country-specific
354 effects were largely absent in our data (Figs 2 and 3) suggesting inconsistent effects even within
355 the countries (cities). Importantly, the changes in use of urban green-spaces during the COVID-
356 19 shutdowns should be similar within a specific city. Thus, even if the use of public green
357 spaces increased in some of our studied cities and decreased in others, we should, but do not,
358 see consistent city-specific escape distance responses of the studied species (Figs 2, 3, S3 and
359 S4). In other words, despite dramatic changes in human outdoor presence the studied
360 populations of birds did not adjust their tolerance to humans again supports the idea that the
361 urban bird populations are already habituated to human presence and resilient to fluctuations in
362 human densities and encounters. The results of our study might also help to explain why several
363 previous studies haven't detected increased probability of occurrence of wildlife in emptied
364 cities (e.g. Gordo *et al.*, 2021; Vardi *et al.*, 2021) – even lower than casual human activity in
365 cities may require sharp behavioural adjustments in wild-living animals.

366 In conclusion, human population, and hence disturbance, is predicted to further increase during
367 the 21st century (Sanderson *et al.* 2002). Therefore, animals will be increasingly forced to
368 occupy human-altered environments, and altered environments will host more humans. Our
369 results indicate that urban birds do not flexibly and quickly modify their escape behaviour to a
370 temporarily reduced human presence, rather the birds had similar levels of tolerance towards
371 humans in pre- and during-pandemic periods, and regardless of the stringency of governmental
372 measures. In other words, our results reveal a degree of urban tolerance that is not further
373 modified by even dramatic changes in human activity patterns. Whether such tolerance reflects
374 a differential settlement of individuals in cities, natural selection on urban tolerance, or
375 habituation-like processes that reduce plasticity requires detailed investigation. Studies of
376 individually-marked birds that are repeatedly tested over time (*sensu* Carrete & Tella 2010,
377 2013), will be particularly important as will unambiguous quantification of individual birds'
378 history of exposure to humans. There is an important role of experimental manipulations of
379 human densities in the studied areas and on capitalising on large changes in population size
380 over short periods of time which might be associated with natural disasters as well as rapid
381 urban development. These observations of animal behaviour under various regimes of human
382 activity and disturbance can help us quantify costs and benefits of behavioural responses to
383 humans. Such insights help us better understand how species tolerate humans and, importantly,
384 how urbanisation modifies behaviour.

385

386 **Acknowledgments**

387 M.W. thanks Eliza Anderson, Mickey Balzereit, Sammi Cunningham, Alisha Dabonde, Chloe
388 Daws, Megan Dennis, Zoe Kellett, Clement Masse, Aaron Moore, Bailey Raats, Max Radvan,
389 Frost Rebecca, Anthony Rendall, Natalie Searle, Mel Sheedy, Will Standish, Mia Stott, and
390 Millie Toomey for help with the collection of escape responses of birds in Melbourne. Clement
391 Masse helped us with data collection in Rovaniemi. J.J. was partly supported by the project
392 “Responsible Tourism Planning” (2019-2023; project number 326348) founded by the
393 Academy of Finland (PROFI 5 Competitive funding to strengthen universities’ research
394 profiles).

395

396 **Ethical approval**

397 The fieldwork protocols comply with the current laws of the countries. This kind of research
398 requires no special permits in Europe. In Australia, Animal ethics approvals (Deakin University
399 Animal Ethics Committee Permits B10-2018 and B08-2021) and permits (DEWLP, 10008731
400 and 10010123) were obtained. All fieldwork was conducted in accordance with the approved
401 guidelines. Data were collected in public places and on private lands where no special permit
402 was required. The method used to estimate avian tolerance towards human disturbance was
403 designed to cause only brief and minimal disturbance to birds; in cities, this disturbance
404 typically does not differ from standard background disturbance caused by other site visitors.

405

406 **Declaration of competing interest**

407 The authors declare that they have no known competing financial interests or personal
408 relationships that could have appeared to influence the work reported in this paper.

409

410 **Data statement**

411 Data and computer code used to generate the results, as well as all supporting material including
412 the plots of model assumptions are freely available from Open Science Framework repository
413 <https://doi.org/10.17605/OSF.IO/WUZH7> (Bulla *et al.* 2022).

414

415 **References**

- 416 Albrecht, T. & Klvaňa, P. (2004). Nest crypsis, reproductive value of a clutch and escape
417 decisions in incubating female mallards *Anas platyrhynchos*. *Ethology*, 110, 603–613.
- 418 Audet, J.-N., Ducatez, S. & Lefebvre, L. (2016). The town bird and the country bird: problem
419 solving and immunocompetence vary with urbanization. *Behavioral Ecology*, 27, 637–
420 644.
- 421 Barr, D.J., Levy, R., Scheepers, C. & Tily, H.J. (2013). Random effects structure for
422 confirmatory hypothesis testing: Keep it maximal. *Journal of Memory and Language*, 68,
423 255–278.
- 424 Basile, M., Russo, L.F., Russo, V.G., Senese, A. & Bernardo, N. (2021). Birds seen and not
425 seen during the COVID-19 pandemic: The impact of lockdown measures on citizen
426 science bird observations. *Biological Conservation*, 256, 109079.
- 427 Bates, A.E., Primack, R.B., Biggar, B.S., Bird, T.J., Clinton, M.E., Command, R.J., *et al.*
428 (2021). Global COVID-19 lockdown highlights humans as both threats and custodians of
429 the environment. *Biological Conservation*, 263, 109175.
- 430 Bates, D., Mächler, M., Bolker, B. & Walker, S. (2015). Fitting linear mixed-effects models
431 using lme4. *Journal of Statistical Software*, 67, 1–48.
- 432 Blumstein, D.T. (2006). Developing an evolutionary ecology of fear: how life history and
433 natural history traits affect disturbance tolerance in birds. *Animal Behaviour*, 71, 389–
434 399.
- 435 Brooks, S.P. & Gelman, A. (1998). General methods for monitoring convergence of iterative
436 simulations. *Journal of Computational and Graphical Statistics*, 7, 434–455.
- 437 Bulla, M., Blumstein, D.T., Benedetti, Y., Floigl, K., Jokimäki, J., Kaisanlahti-Jokimäki, M.-
438 L., Markó, G., Morelli, F., Siretckaia, A., Szakony, S., Weston, M.A., Zeid, F.A.,
439 Tryjanowski, P., Albrecht, T. & Mikula, P. (2022). Supporting information for 'Urban
440 birds' flight responses were unaffected by the COVID-19 shutdowns'. *Open Science*
441 *Framework* <https://doi.org/10.17605/OSF.IO/WUZH7>.
- 442 Bulla, M., Valcu, M., Dokter, A.M., Dondua, A.A., Kosztolányi, A., Rutten, A., *et al.* (2016).
443 Unexpected diversity in socially synchronized rhythms of shorebirds. *Nature*, 540, 109–
444 113.
- 445 Bürkner, P.-C. (2017). brms: An R package for Bayesian multilevel models using Stan.
446 *Journal of Statistical Software*, 80, 1–28.

- 447 Bürkner, P.-C. (2018). Advanced Bayesian multilevel modeling with the R package brms. *The*
448 *R Journal*, 10, 395–411.
- 449 Carrete, M. & Tella, J.L. (2010). Individual consistency in flight initiation distances in
450 burrowing owls: a new hypothesis on disturbance- induced habitat selection. *Biology*
451 *Letters*, 23, 167–170.
- 452 Carrete, M. & Tella, J.L. (2013). High individual consistency in fear of humans throughout
453 the adult lifespan of rural and urban burrowing owls. *Scientific Reports*, 3, 1–7.
- 454 Crimmins, T.M., Posthumus, E., Schaffer, S. & Prudic, K.L. (2021). COVID-19 impacts on
455 participation in large scale biodiversity-themed community science projects in the United
456 States. *Biological Conservation*, 256, 109017.
- 457 Derryberry, E.P., Phillips, J.N., Derryberry, G.E., Blum, M.J. & Luther, D. (2020). Singing in
458 a silent spring: Birds respond to a half-century soundscape reversion during the COVID-
459 19 shutdown. *Science (1979)*, 370, 575–579.
- 460 Díaz, M., Møller, A.P., Flensted-Jensen, E., Grim, T., Ibáñez-Álamo, J.D., Jokimäki, J., *et al.*
461 (2013). The geography of fear: a latitudinal gradient in anti-predator escape distances of
462 birds across Europe. *PLoS ONE*, 8, e64634.
- 463 Ducatez, S., Audet, J. N., Rodriguez, J. R., Kayello, L. & Lefebvre, L. (2017). Innovativeness
464 and the effects of urbanization on risk-taking behaviors in wild Barbados birds. *Animal*
465 *Cognition*, 20, 33–42.
- 466 Ducatez, S., Sol, D., Sayol, F. & Lefebvre, L. (2020). Behavioural plasticity is associated with
467 reduced extinction risk in birds. *Nature Ecology & Evolution*, 1–6.
- 468 Ellenberg, U., Mattern, T. & Seddon, P.J. (2009). Habituation potential of yellow-eyed
469 penguins depends on sex, character and previous experience with humans. *Animal*
470 *Behaviour*, 77, 289–296.
- 471 Fang, H., Wang, L. & Yang, Y. (2020). Human mobility restrictions and the spread of the
472 Novel Coronavirus (2019-nCoV) in China. *Journal of Public Economics*, 191, 104272.
- 473 Gatto, M., Bertuzzo, E., Mari, L., Miccoli, S., Carraro, L., Casagrandi, R., *et al.* (2020).
474 Spread and dynamics of the COVID-19 epidemic in Italy: Effects of emergency
475 containment measures. *Proc Natl Acad Sci U S A*, 117, 10484–10491.
- 476 Gelman, A. & Hill, J. (2007). *Data analysis using regression and multilevel/hierarchical*
477 *models*. Cambridge University Press, Cambridge.
- 478 Gelman, A. & Su, Y.-S. (2018). *arm: Data analysis using regression and*
479 *multilevel/hierarchical models*. Available at: <http://CRAN.R-project.org/package=arm>.
480 Last accessed 18 May 2022.

- 481 Gelman, A., Su, Y.-S., Yajima, M., Hill, J., Pittau, M., Kerman, J., *et al.* (2016). Data
482 Analysis Using Regression and Multilevel/Hierarchical Models. *CRAN Repository*.
- 483 Geng, D.C., Innes, J., Wu, W. & Wang, G. (2021). Impacts of COVID-19 pandemic on urban
484 park visitation: a global analysis. *Journal of Forestry Research*, 32, 553-567.
- 485 Gnanapragasam, J.J., Ekanayake, K.B., Ranawana, K., Symonds, M.R.E. & Weston, M.
486 (2021). Civil war is associated with longer escape distances among Sri Lankan birds. *The*
487 *American Naturalist*, 198, 653-659.
- 488 Gordo, O., Brotons, L., Herrando, S. & Gargallo, G. (2021). Rapid behavioural response of
489 urban birds to COVID-19 lockdown. *Proceedings of the Royal Society B: Biological*
490 *Sciences*, 288, 20202513.
- 491 Griffin, A. S., Netto, K. & Peneaux, C. (2017). Neophilia, innovation and learning in an
492 urbanized world: a critical evaluation of mixed findings. *Current Opinion in Behavioral*
493 *Sciences*, 16, 15–22.
- 494 Guay, P.J., van Dongen, W.F.D., Robinson, R.W., Blumstein, D.T. & Weston, M.A. (2016).
495 AvianBuffer: An interactive tool for characterising and managing wildlife fear responses.
496 *Ambio*, 45, 841–851.
- 497 Hadfield, J.D. (2010). MCMC methods for multi-response generalised linear mixed models:
498 the MCMCglmm R package. *Journal of Statistical Software*, 33, 1–22.
- 499 Hadfield, J.D. & Nakagawa, S. (2010). General quantitative genetic methods for comparative
500 biology: phylogenies, taxonomies and multi-trait models for continuous and categorical
501 characters. *Journal of Evolutionary Biology*, 23, 494–508.
- 502 Hale, T., Angrist, N., Goldszmidt, R., Kira, B., Petherick, A., Phillips, T., *et al.* (2021). A
503 global panel database of pandemic policies (Oxford COVID-19 Government Response
504 Tracker). *Nature Human Behaviour* 2021 5:4, 5, 529–538.
- 505 Hochachka, W.M., Alonso, H., Gutiérrez-Expósito, C., Miller, E. & Johnston, A. (2021).
506 Regional variation in the impacts of the COVID-19 pandemic on the quantity and quality
507 of data collected by the project eBird. *Biological Conservation*, 254, 108974.
- 508 Huang, X., Li, Z., Jiang, Y., Li, X. & Porter, D. (2020). Twitter reveals human mobility
509 dynamics during the COVID-19 pandemic. *PLOS ONE*, 15, e0241957.
- 510 Jetz, W., Thomas, G.H., Joy, J.B., Hartmann, K. & Mooers, A.O. (2012). The global diversity
511 of birds in space and time. *Nature*, 491, 444–448.
- 512 Kraemer, M.U.G., Yang, C.H., Gutierrez, B., Wu, C.H., Klein, B., Pigott, D.M., *et al.* (2020).
513 The effect of human mobility and control measures on the COVID-19 epidemic in China.
514 *Science (1979)*, 368, 493–497.

- 515 Legagneux, P. & Ducatez, S. (2013). European birds adjust their flight initiation distance to
516 road speed limits. *Biology Letters*, 9, 20130417.
- 517 Manenti, R., Mori, E., di Canio, V., Mercurio, S., Picone, M., Caffi, M., *et al.* (2020). The
518 good, the bad and the ugly of COVID-19 lockdown effects on wildlife conservation:
519 Insights from the first European locked down country. *Biological Conservation*, 249,
520 108728.
- 521 March, D., Metcalfe, K., Tintoré, J. & Godley, B.J. (2021). Tracking the global reduction of
522 marine traffic during the COVID-19 pandemic. *Nature Communications* 2021 12:1, 12,
523 1–12.
- 524 Marzluff, J.M., Bowman, R. & Donnelly, R. (2001). A historical perspective on urban bird
525 research: trend, terms, and approaches. In: *Avian Ecology and Conservation in an*
526 *Urbanizing World* (eds. Marzluff, J.M., Bowman, R. & Donnelly, R.). Kluwer, New
527 York, pp. 20–47.
- 528 Mikula, P. (2014). Pedestrian density influences flight distances of urban birds. *Ardea*, 102,
529 53–60.
- 530 Mikula, P., Díaz, M., Albrecht, T., Jokimäki, J., Kaisanlahti-Jokimäki, M.-L., Kroitero, G., *et*
531 *al.* (2018). Adjusting risk-taking to the annual cycle of long-distance migratory birds.
532 *Scientific Reports*, 8, 13989.
- 533 Mikula, P., Jokimäki, J., Kaisanlahti-Jokimäki, M.L., Markó, G., Morelli, F., Møller, A.P., *et*
534 *al.* (2021). Face mask-wear did not affect large-scale patterns in escape and alertness of
535 urban and rural birds during the COVID-19 pandemic. *Science of The Total Environment*,
536 793, 148672.
- 537 Montgomery, R.A., Raupp, J. & Parkhurst, M. (2021). Animal Behavioral Responses to the
538 COVID-19 Quietus. *Trends in Ecology & Evolution*, 36, 184–186.
- 539 Morelli, F., Benedetti, Y., Díaz, M., Grim, T., Ibáñez-Álamo, J.D., Jokimäki, J., *et al.* (2019).
540 Contagious fear: Escape behavior increases with flock size in European gregarious birds.
541 *Ecology and Evolution*, 9, 6096–6104.
- 542 Morelli, F., Mikula, P., Benedetti, Y., Bussière, R., Jerzak, L. & Tryjanowski, P. (2018).
543 Escape behaviour of birds in urban parks and cemeteries across Europe: Evidence of
544 behavioural adaptation to human activity. *Science of the Total Environment*, 631–632,
545 803–810.
- 546 Park, M., Lim, J.T., Wang, L., Cook, A.R. & Dickens, B.L. (2021). Urban-Rural Disparities
547 for COVID-19: Evidence from 10 Countries and Areas in the Western Pacific. *Health*
548 *Data Science*, 2021, 1–9.

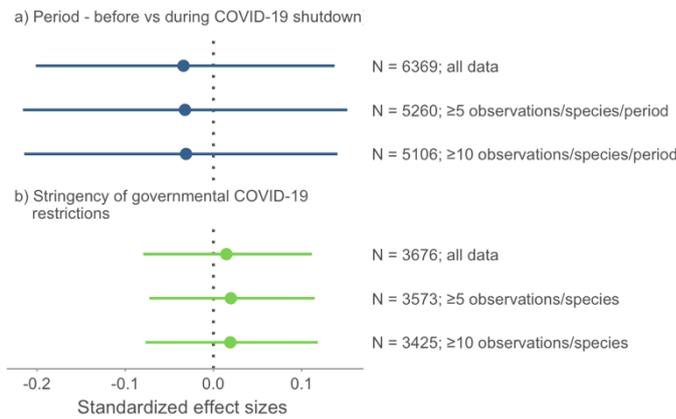
- 549 Piratelli, A.J., Favoretto, G.R. & de Almeida Maximiano, M.F. (2015). Factors affecting
550 escape distance in birds. *Zoologia (Curitiba)*, 32, 438–444.
- 551 R Development Core Team. (2021). *A Language and Environment for Statistical Computing*.
552 R Foundation for Statistical Computing, Vienna, Austria.
- 553 Randler, C., Tryjanowski, P., Jokimäki, J., Kaisanlahti-Jokimäki, M. L., & Staller, N. (2020).
554 SARS-CoV2 (COVID-19) Pandemic lockdown influences nature-based recreational
555 activity: The case of birders. *International journal of environmental research and public
556 health*, 17, 7310.
- 557 Rankin, C.H., Abrams, T., Barry, R.J., Bhatnagar, S., Clayton, D.F., Colombo, J., *et al.*
558 (2009). Habituation revisited: An updated and revised description of the behavioral
559 characteristics of habituation. *Neurobiology of Learning and Memory*, 92, 135–138.
- 560 Rutz, C., Loretto, M.C., Bates, A.E., Davidson, S.C., Duarte, C.M., Jetz, W., *et al.* (2020).
561 COVID-19 lockdown allows researchers to quantify the effects of human activity on
562 wildlife. *Nature Ecology and Evolution*, 4, 1156–1159.
- 563 Samia, D.S.M. & Blumstein, D.T. (2015). Birds Flush Early and Avoid the Rush: An
564 Interspecific Study. *PLOS ONE*, 10, e0119906.
- 565 Samia, D.S.M., Blumstein, D.T., Díaz, M., Grim, T., Ibáñez-Álamo, J.D., Jokimäki, J., *et al.*
566 (2017). Rural-urban differences in escape behavior of european birds across a latitudinal
567 gradient. *Frontiers in Ecology and Evolution*, 5, 66.
- 568 Samia, D.S.M., Blumstein, D.T., Stankowich, T. & Cooper, W.E. (2016). Fifty years of
569 chasing lizards: new insights advance optimal escape theory. *Biological Reviews*, 91,
570 349–366.
- 571 Samia, D.S.M., Nakagawa, S., Nomura, F., Rangel, T.F. & Blumstein, D.T. (2015). Increased
572 tolerance to humans among disturbed wildlife. *Nature Communications*, 6, 8877.
- 573 Samia, D.S.M., Nomura, F. & Blumstein, D.T. (2013). Do animals generally flush early and
574 avoid the rush? A meta-analysis. *Biology Letters*, 9, 20130016.
- 575 Sanderson, E.W., Jaiteh, M., Levy, M.A., Redford, K.H., Wannebo, A. v. & Woolmer, G.
576 (2002). The Human Footprint and the Last of the Wild: The human footprint is a global
577 map of human influence on the land surface, which suggests that human beings are
578 stewards of nature, whether we like it or not. *BioScience*, 52, 891–904.
- 579 Sayol, F., Sol, D. & Pigot, A.L. (2020). Brain size and life history interact to predict urban
580 tolerance in birds. *Frontiers in Ecology and Evolution*, 8, 58.
- 581 Schielzeth, H. & Forstmeier, W. (2009). Conclusions beyond support: overconfident estimates
582 in mixed models. *Behavioral Ecology*, 20, 416–420.

- 583 Schliep, K.P. (2011). phangorn: phylogenetic analysis in R. *Bioinformatics*, 27, 592–593.
- 584 Soga, M., Evans, M.J., Cox, D.T.C. & Gaston, K.J. (2021). Impacts of the COVID-19
585 pandemic on human–nature interactions: Pathways, evidence and implications. *People*
586 *and Nature*, 3, 518–527.
- 587 Sol, D., Lapedra, O. & González-Lagos, C. (2013). Behavioural adjustments for a life in the
588 city. *Animal Behaviour*, 85, 1101–1112.
- 589 Sol, D., Maspons, J., Gonzalez-Voyer, A., Morales-Castilla, I., Garamszegi, L.Z. & Møller,
590 A.P. (2018). Risk-taking behavior, urbanization and the pace of life in birds. *Behavioral*
591 *Ecology and Sociobiology*, 72, 59.
- 592 Sol, D., Timmermans, S. & Lefebvre, L. (2002). Behavioural flexibility and invasion success
593 in birds. *Animal Behaviour*, 63, 495–502.
- 594 Soto, E.H., Botero, C.M., Milanés, C.B., Rodríguez-Santiago, A., Palacios-Moreno, M., Díaz-
595 Ferguson, E., Velázquez, Y.R., Abbehusen, A., Guerra-Castro, E., Simoes, N., Muciño-
596 Reyes, M. (2021). How does the beach ecosystem change without tourists during
597 COVID-19 lockdown? *Biological Conservation*, 255, 108972.
- 598 Stan Development Team (2022). RStan: the R interface to Stan. R package version 2.21.5.
599 Available at <https://mc-stan.org/>.
- 600 Stankowich, T. & Blumstein, D.T. (2005). Fear in animals: a meta-analysis and review of risk
601 assessment. *Proceedings of the Royal Society of London B: Biological Sciences*, 272,
602 2627–2634.
- 603 Symonds, M. R., Weston, M. A., Van Dongen, W. F., Lill, A., Robinson, R. W. & Guay, P. J.
604 (2016). Time since urbanization but not encephalisation is associated with increased
605 tolerance of human proximity in birds. *Frontiers in Ecology and Evolution*, 4, 117.
- 606 Tryjanowski, P., Kosicki, J.Z., Hromada, M. & Mikula, P. (2020). The emergence of
607 tolerance of human disturbance in Neotropical birds. *Journal of Tropical Ecology*, 36, 1–
608 5.
- 609 Tryjanowski, P., Møller, A. P., Morelli, F., Biaduń, W., Brauze, T., Ciach, M., Czechowski,
610 P., Czyz, S., Dulisz, B., Golawski, A., Hetmański, T., Indykiewicz, P., Mitrus, C.,
611 Myczko, L., Nowakowski, J.J., Polakowski, M., Takacs, V., Wysocki, D. & Zduniak, P.
612 (2016). Urbanization affects neophilia and risk-taking at bird-feeders. *Scientific Reports*,
613 6, 1–7.
- 614 Uyeda, J.C., Zenil-Ferguson, R. & Pennell, M.W. (2018). Rethinking phylogenetic
615 comparative methods. *Systematic Biology*, 67, 1091–1109.

- 616 van Dongen, W. F., Robinson, R. W., Weston, M. A., Mulder, R. A. & Guay, P. J. (2015).
617 Variation at the DRD4 locus is associated with wariness and local site selection in urban
618 black swans. *BMC Evolutionary Biology*, 15, 1–11.
- 619 Vardi, R., Berger-Tal, O. & Roll, U. (2021). iNaturalist insights illuminate COVID-19 effects
620 on large mammals in urban centers. *Biological Conservation*, 254, 108953.
- 621 Venter, Z.S., Aunan, K., Chowdhury, S. & Lelieveld, J. (2020a). COVID-19 lockdowns cause
622 global air pollution declines. *Proc Natl Acad Sci U S A*, 117, 18984–18990.
- 623 Venter, Z.S., Barton, D.N., Gundersen, V., Figari, H. & Nowell, M. (2020b). Urban nature in
624 a time of crisis: Recreational use of green space increases during the COVID-19 outbreak
625 in Oslo, Norway. *Environmental Research Letters*, 15, 104075.
- 626 Webb, N. v & Blumstein, D.T. (2005). Variation in Human Disturbance Differentially Affects
627 Predation Risk Assessment in Western Gulls. *Condor*, 107, 178–181.
- 628 Weston, M.A., McLeod, E.M., Blumstein, D.T. & Guay, P.-J. (2012). A review of flight-
629 initiation distances and their application to managing disturbance to Australian birds.
630 *Emu*, 112, 269–286.
- 631 Wickham, H. (2016). *ggplot2: Elegant Graphics for Data Analysis*. Springer-Verlag, New
632 York.
- 633 Wilman, H., Belmaker, J., Simpson, J., de la Rosa, C., Rivadeneira, M.M. & Jetz, W. (2014).
634 EltonTraits 1.0: Species-level foraging attributes of the world’s birds and mammals.
635 *Ecology*, 95, 2027–2027.
- 636 Wu, C., Guo, Y., Guo, H., Yuan, J., Ru, L., Chen, H., Du, B. & Zhang, L. (2021). An
637 investigation of traffic density changes inside Wuhan during the COVID-19 epidemic
638 with GF-2 time-series images. *International Journal of Applied Earth Observation and*
639 *Geoinformation*, 103, 102503.
- 640 Ydenberg, R.C. & Dill, L.M. (1986). The economics of fleeing from predators. *Advances in*
641 *the Study of Behavior*, 16, 229–249.
- 642 Zaccaroni, M., Ciuffreda, M., Paganin, M. & Beani, L. (2010). Does an early aversive
643 experience to humans modify antipredator behaviour in adult Rock partridges?
644 <http://dx.doi.org/10.1080/08927014.2007.9522561>, 19, 193–200.
- 645 Zellmer, A.J., Wood, E.M., Surasinghe, T., Putman, B.J., Pauly, G.B., Magle, S.B., *et al.*
646 (2020). What can we learn from wildlife sightings during the COVID-19 global
647 shutdown? *Ecosphere*, 11, e03215.
- 648

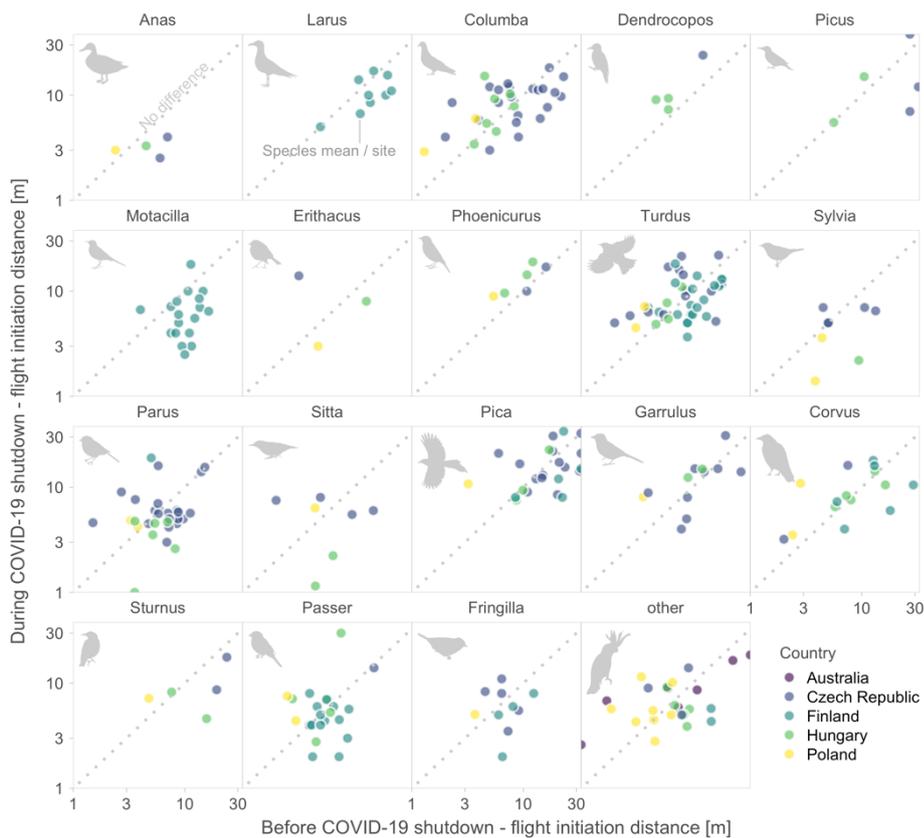
649 **Figure 1.** Changes in avian tolerance towards humans (measured as the flight initiation
650 distance) (a) before vs during the COVID-19 shutdowns and (b) due to stringency of
651 governmental restrictions during COVID-19 shutdowns. The dots with horizontal lines
652 represent estimated standardised effect size and their 95% confidence intervals based on the
653 joint posterior distribution of 5000 simulated values generated by the *sim* function from the *arm*
654 package (Gelman *et al.* 2016) using the mixed model outputs (Tables S1 and S2). The colour
655 highlights the investigated association with Period before vs during the COVID-19 shutdowns
656 – (blue) and stringency index (green). Depicted are effect sizes based on full and reduced
657 datasets with ≥ 5 or ≥ 10 observations per species and period. Note that all estimates centre
658 around zero.

659



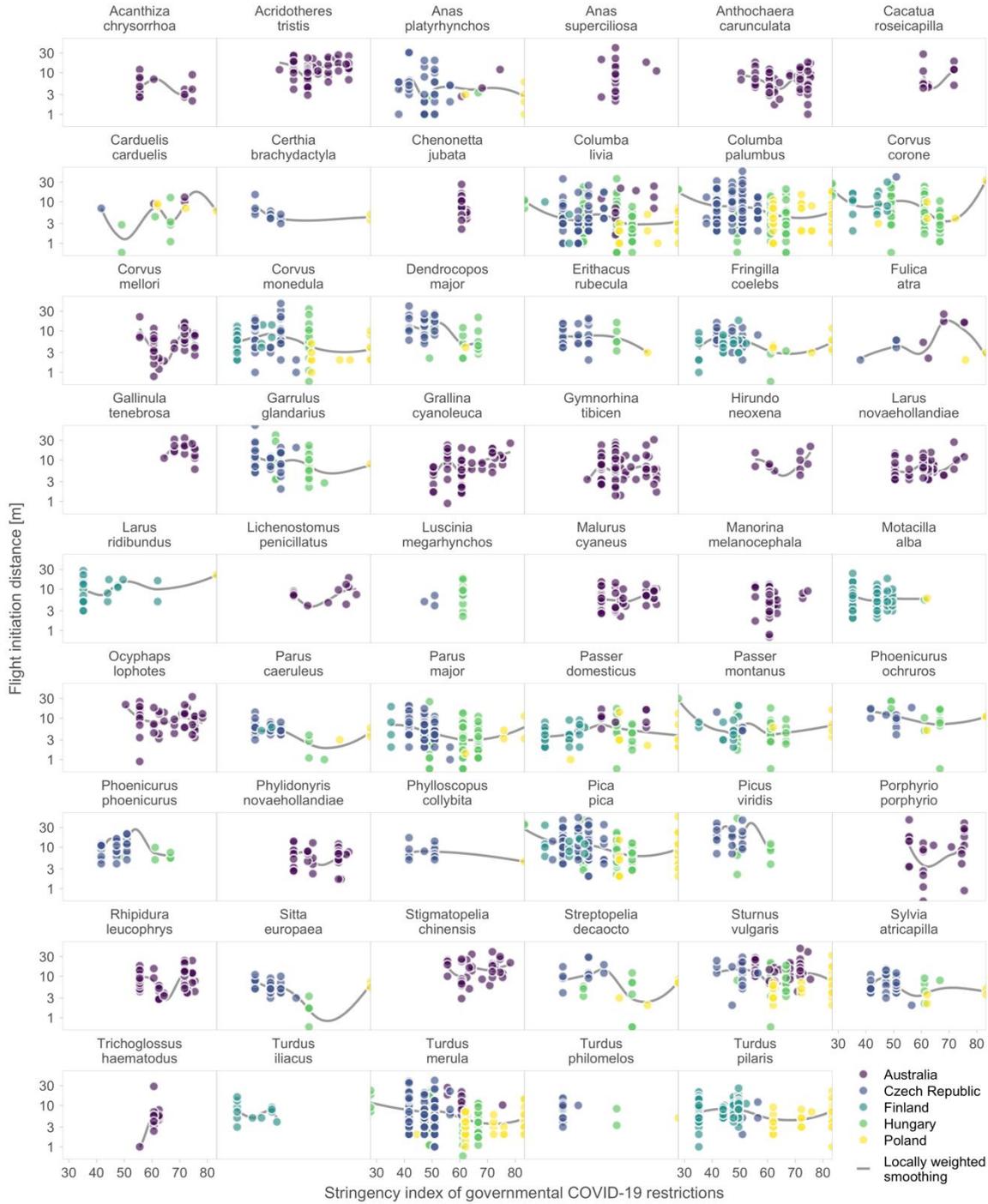
660

661 **Figure 2.** Avian tolerance towards humans (measured as the flight initiation distance) before
662 and during the COVID-19 shutdowns according to genus. Dots represent means or single escape
663 distance observations of species at specific sites (e.g. specific park or cemetery) with data for
664 both periods (before and during the shutdowns) and not corrected for other factors such as
665 starting distance of the observer. Dot colour highlights the country (Australia in violet, the
666 Czech Republic blue, Finland turquoise, Hungary green, and Poland yellow). Dotted lines
667 indicate no difference, dots above the line indicate lower tolerance towards humans (i.e. longer
668 escape distances), dots below the line indicate lower tolerance before than during the COVID-
669 19 shutdowns. Panels are ordered according to evolutionary history of birds with top left panels
670 representing the oldest genera, and bottom right, the youngest. Panel titled ‘other’ contains
671 genera with only one or two data points. The axes are on the log-scale. Note the within- and
672 between-genera and cross-country variation and stochasticity in response to the shutdowns (for
673 species-specific plots see Fig. S3, for within- and between-year variation Fig. S4) and that using
674 escape distance values controlled for starting distance had little influence on the depicted
675 relationships (Fig. S5). Silhouette of *Garrulus glandarius*, *Motacilla alba*, *Picus viridis*,
676 *Phoenicurus ochruros*, *Sylvia borin* were drawn by Martin Bulla and are available at
677 <https://doi.org/10.17605/OSF.IO/WUZH7> (Bulla *et al.* 2022), silhouette of *Erithacus rubecula*
678 drawn by Rebecca Groom, and silhouettes of *Fringilla coelebs* and *Sturnus vulgaris* by Maxime
679 Dahirel and available at PhyloPic (<http://phylopic.org>); all these silhouettes are under Creative
680 Commons Attribution 3.0 Unported licence (<https://creativecommons.org/licenses/by/3.0/>).
681 The remaining silhouettes are available at PhyloPic under the Public Domain Dedication 1.0
682 license (<https://creativecommons.org/publicdomain/zero/1.0/>).



683

684 **Figure 3.** Species-specific avian tolerance towards humans (measured as the flight initiation
685 distance) in relation to severeness of governmental restrictions during COVID-19 shutdowns
686 quantified as a stringency index. Each dot represents a single escape distance observation (not
687 corrected for other factors such as starting distance of the observer) and a day-specific value of
688 governmental stringency index in a given country. Dot colour highlights the country (Australia
689 in violet, the Czech Republic blue, Finland turquoise, Hungary green and Poland yellow). Grey
690 lines represent locally weighted smoothing, a non-parametric local regression fitted with the
691 *ggplot* function of *ggplot2* package (Wickham 2016), highlighting heterogenous (and usually
692 unclear) within- and between- species trends. Note, the y-axis is on the log-scale, some species
693 lack trend lines because data distribution hindered the smoothing and visualised are only data
694 for species with ≥ 10 escape distance observations.



695