

1 **Rapid behavioural response of urban birds to covid-19** 2 **lockdown**

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15 **Abstract**

16 Biodiversity is threatened by the current exponential growth of urban areas. However, it is still
17 poorly understood how animals can cope with and adapt to these rapid and dramatic
18 transformations of natural environments. The COVID-19 pandemic provides us with a unique
19 opportunity to unveil the adaptive mechanisms involved in this process. Lockdown measures
20 imposed in most countries are causing an unprecedented reduction of human activities giving us
21 an experimental setting to assess the effects of our lifestyle on biodiversity. We studied the birds'
22 response to the Spanish population lockdown by using more than 126,000 bird records collected
23 by a citizen science project. We compared the occurrence and detectability of birds during the
24 spring 2020 lockdown with baseline data from previous years in the same urban areas and dates.
25 We found that birds did not increase their probability of occurrence in urban areas during the
26 lockdown, refuting the hypothesis that nature has recovered its space in human emptied urban
27 areas. However, we found an increase in bird detectability, especially during early morning,
28 suggesting a rapid change in the birds' daily routines in response to quieter and less crowded cities.
29 In conclusion, urban birds show high behavioural plasticity to rapidly adapt to novel environmental
30 conditions, as those imposed by the COVID-19.

31

32 **Keywords:** Coronavirus disease, behavioural plasticity, urban ecology, detectability, citizen
33 science

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35 **Running Title:** COVID-19 and birds

36

37 **1. Introduction**

38 Since the first human settlements some millennia ago, the anthropogenic transformation of the
39 natural environment to build towns and cities has been a hallmark of humanity. During the last
40 century, urbanization has experienced exponential growth across the world and it is expected to
41 continue as more people will move from rural to urban areas [1-4]. As a result, urbanization has
42 become one of the most important drivers of global change and a major threat to biodiversity
43 [2,4,6,7]. Novel, human created environments, such as urban areas, represent a formidable
44 challenge for organisms because the magnitude and pace of the environmental alterations
45 imposed by humans usually exceed their limits of tolerance leading to populations shrinkage and
46 extinction [6,8]. Urban challenges include combating chemical [3], acoustic [9,10] and light pollution
47 [11, 12], human disturbance [6,13], new pathogens [14,15] and predators [16,17], and human
48 infrastructures [16,18]. However, some species are able to overcome these challenges and thrive
49 in urban environments [4,8,13]. Therefore, a key question in urban ecology is how species cope
50 with urbanization. Countless studies have demonstrated that adapting to urban environments imply
51 some kind of phenotypical differentiation from non-urban relatives [8-10,13]. Indeed, organisms are
52 forced to adjust their physiology, behaviour and life histories to the novel conditions imposed by
53 the city [6,8]. However, the mechanisms underlying the differences between urban and non-urban
54 dwellers remain largely unknown [7,8]. Observed adjustments are mostly consistent with
55 phenotypically plastic responses [13], but individual sorting and microevolutionary changes by
56 divergent selection could be playing a role [4,6,8]. Perhaps, our inability to disentangle these
57 mechanisms comes from a deficit of experimental studies in urban ecology [10], in spite of the fact
58 that human transformed environments provide often ready-made experiments. The current spread
59 of the novel coronavirus disease (COVID-19) and its consequences represents an excellent
60 example, as we are involuntarily involved in a major unintended social experiment.

61 After the declaration of the COVID-19 pandemic in March 2020 by the World Health
62 Organization, most countries have implemented social and health measures unprecedented in
63 recent history. These measures, aimed at containing the virus spread [19-23], have focused on
64 social distancing and population confinement, as well as the cease of non-essential productive and
65 social activities. Overall, the measures have contributed to a global diminishing of human activities
66 [24]. This abrupt and dramatic disruption of most human social and economic activities have
67 already had quantifiable effects on urban environments by marked reductions in air pollution [25-
68 27] and noise [28,29]. One of the most noticeable and generalized measure has been applying
69 certain degree of population lockdown, which renders our city streets empty and virtually silent.
70 This situation provides a once-in-a-lifetime opportunity to study urban wildlife responses to less
71 active, noisy and polluted cities and gain unprecedented mechanistic insights into how human
72 activities affect wildlife [24,30,31]. Product of the human lockdown, unusual observations of animals

73 in urban areas worldwide have flooded the media and the internet planting in the social imaginary
74 the idea that “nature is getting back its space” (*sensu* [32]). Although plausible, this idea is, in most
75 cases, based on anecdotal records, sometimes false [32], without any quantitative scientific
76 investigation supporting such claim [24].

77 In this work, we aimed to assess the behavioural responses of birds to the unexpected and
78 drastic changes occurring in urban environments resulting from the COVID-19 lockdown in a
79 densely populated area of north eastern Spain (Catalonia). Following China [23] and Italy [20],
80 Spain was the third country worldwide to impose a heavy population lockdown to stop COVID-19
81 spread. The declaration of the national emergency in March 14th 2020 by the Spanish Government
82 imposed the strictest lockdown measures in Europe. Since then, social restrictions were alleviated
83 progressively until the end of June (electronic supplementary material, figure S1, table S1). As in
84 other parts of the world, this big halt of human activities has had significant environmental effects
85 with reduced air contamination and noise in Spanish cities [26,27]. The severity of the lockdown
86 measures imposed in Spain, make this country especially suitable to study COVID-19 lockdown
87 effects in urban fauna, as they enjoyed exceptionally quieter and peaceful towns and cities during
88 many weeks.

89 We compared bird records collected during the first four weeks of the lockdown in towns
90 and cities of Catalonia with the available records for the same region and dates since 2015. These
91 historical records were used as baseline data. Our broad scale approach (hundreds of study sites
92 covering an area of 32,000 km²) at community level (we studied 16 different species) allowed us
93 a robust testing of two key questions:

94 1) Did urban birds become more common in response to human empty cities? It can be
95 predicted that decreased human presence and disturbance allowed animals to occupy spaces that
96 used to be above their fear tolerance thresholds [6,32]. Therefore, we expected a higher occurrence
97 in 2020 compared to the historical records for the same urban areas. This effect being likely
98 stronger for shier species (i.e., urban adapters), who are less tolerant to human disturbances [6,13].

99 2) Were urban birds more detectable as a consequence of quieter cities? It can be predicted
100 that decreased anthropogenic noise increased the effective distance of among bird
101 communications [6,9,10,13] and be more easily perceived by observers [33,34]. Moreover, as the
102 masking effect of human acoustic contamination mostly disappeared, we expected an increase in
103 singing activity, including potential shifts in its timing, to profit from the new urban soundscape
104 [9,10,35,36]. Therefore, we expected a higher detectability of urban birds during the lockdown than
105 in previous years, with possible changes in the daily patterns of detection.

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108 **2. Material and methods**

109 *a) Bird data*

110 On March 14th 2020, the Spanish Government declared the national emergency due to COVID-19
111 outbreak and imposed severe social restrictions. These restrictions included mandatory and
112 permanent confinement of the population, borders closure, limitations in public transport, on-line
113 education, working from home whenever possible, and closure of non-essential business and
114 public services, such as supermarkets, pharmacies or hospitals. One day later, we launched the
115 project “#JoEmQuedoACasa” (I stay at home) within the citizen science on-line platform *ornitho*
116 (www.ornitho.cat). This platform aims to collect wildlife records in Catalonia (NE Spain) from
117 birdwatchers and naturalists to improve the biodiversity knowledge of the region. *Ornitho* has been
118 running since 2009 and has gathered more than 6.5 million records to date. The project launched
119 during the lockdown aimed to collect information about wildlife responses to the new environmental
120 conditions resulting from people confinement. In addition to this valuable information, the project
121 was important to keep engaged birdwatchers in this citizen science program by encouraging them
122 to continue complete checklists (checklist with all identified species) submission, even during a
123 period of constrained outdoor activities [37].

124 Lockdown surveys were conducted between March 15th and April 13th of 2020. During
125 these four weeks, people was subjected to the most restrictive conditions of mobility and
126 consequently this period showed the most drastic reduction of human activities (electronic
127 supplementary material, figure S1, table S1). Therefore, lockdown checklists were carried out only
128 from homes (e.g., balconies, rooftops or yards) in urban environments. To determine the effect of
129 lockdown on bird behaviour, we also gathered all complete checklists uploaded to *ornitho* recorded
130 during the same dates between 2015 and 2019. We classified these surveyed sites as urban or
131 non-urban according to the 2017 land use/land cover map of Catalonia [38]. Therefore, we obtained
132 three groups of checklists: urban lockdown, historical urban, and historical non-urban, which
133 contained a total of 126,315 bird records. Historical urban data represented baseline data, while
134 historical non-urban data was included as control data without human disturbances.

135 All checklists had associated basic information about the survey: site (geographical
136 coordinates), date, hour, time invested (which was used as a proxy for sampling effort) and
137 observer identity. We excluded checklists lasting >3 h, as they might be discontinuous surveys. We
138 also excluded those checklists started one hour earlier or later than dawn or sunset, respectively,
139 as they represented nocturnal surveys. To correct for the adjustment of daylight saving time at the
140 end of March, we rescaled recorded hours in civil time to the relevant daily sun events: dawn, noon
141 and sunset, which were established as -1, 0 and 1, respectively. Dawn, noon and sunset were
142 calculated for every geographical coordinate and date by the ‘suncalc’ library (version 0.5.0) for R
143 software [39]. Rescaling was calculated as the quotient between the difference of noon and

144 checklist hour and the difference of dawn or sunset and checklist hour, depending on whether
145 checklist started earlier or later than noon, respectively. This transformation allowed to fix the small
146 bias caused by the longitudinal differences in dawn and sunset across Catalonia as well as by the
147 progressive day length increase during the study period. Not many observers recorded the number
148 of individuals for each species. For this reason, we opted to work with presence/absence data.

149 During the lockdown, observers recorded 1,289 complete bird checklists at 149 sites. The
150 number of replicated surveys per site and observer ranged from 1 to 91 (mean=8.7, SD=12.4).
151 Historical records in urban areas were the scarcest: 1,062 checklists in 410 sites with up to 48
152 replicates per site (mean=2.6, SD=5.2). As expected, data from non-urban areas were the most
153 abundant, as observers usually preferred birdwatching in wild habitats. We gathered 5,849
154 checklists from 3,113 sites. Although one observer made 84 replicates for the same site, on
155 average, observers in this group showed the lowest site fidelity (mean replicates=1.9, SD=3.6).

156 We selected data for the 16 most common sedentary urban species in Catalonia [40,41]
157 (see figure 1). We focused only on sedentary birds to avoid seasonal changes in occurrence and
158 abundance associated with migration. Data from the common and the spotless starlings (*Sturnus*
159 *vulgaris* and *S. unicolor*, respectively) were merged as *Sturnus* spp. as both were not usually
160 identified at species level in most observations due to their high resemblance [42]. Both species
161 are common, well spread, sympatric and share similar habits and behaviour [41]. Thus, we did not
162 expect important differences in their occurrence or detectability.

163 *b) Statistical analyses*

164 To disentangle the effects of individuals' presence (first question) and detection (second question)
165 in our bird data, we used hierarchical occupancy models [43,44]. We considered as replicated
166 surveys those checklists reported by the same observer within the same 1x1 km UTM cell. By
167 combining observer and location, we avoided variability in detection rates due to observer
168 expertise. We could assume confidently that observer experience was randomly distributed across
169 our study area. The equations defining our model were:

170

$$171 \quad \text{logit}(\Psi_j) = \beta_0 + \beta_1 \text{group}_{Lj} + \beta_2 \text{group}_{Uj} + \beta_3 \text{group}_{Nj}$$

$$172 \quad \text{logit}(\rho_i) = \alpha_0 + \alpha_1 \text{group}_{Li} + \alpha_2 \text{group}_{Ui} + \alpha_3 \text{group}_{Ni} + \alpha_4 \text{time}_i + \alpha_5 \text{time}_i \cdot \text{group}_{Li}$$

$$173 \quad + \alpha_6 \text{time}_i \cdot \text{group}_{Ui} + \alpha_7 \text{time}_i \cdot \text{group}_{Ni} + \alpha_8 f(\text{hour}_i) + \alpha_9 f(\text{hour}_i)$$

$$174 \quad \cdot \text{group}_{Li} + \alpha_{10} f(\text{hour}_i) \cdot \text{group}_{Ui} + \alpha_{11} f(\text{hour}_i) \cdot \text{group}_{Ni}$$

175

176 where Ψ_j is the occurrence of a species at site j and ρ_i is its detectability in the checklist i ; groups
177 L , U and N refer to lockdown, urban historical and non-urban, respectively; time refers to the
178 duration of the survey; and hour refers to the starting hour. Hour was included as an unpenalized
179 thin plate regression spline basis function (f) with five degrees of freedom because we expected
180 that detectability could vary in a non-linear way along the day [45,46]. Interactions between group
181 and time and between group and hour allowed to model the effect of these two variables on
182 detectability within each group. To test the significance of hour and interactions, we used log
183 likelihood ratio tests. Basis functions were built by the `smooth.construct` function from package
184 'mgcv' (version 1.8-22 [47]), while occupancy models were run with the `occu` function of package
185 'unmarked' (version 0.12-3 [48]) for R.

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187

188 **3. Results**

189 Probability of occurrence of a species during the lockdown did not differ significantly from the
190 occurrence recorded in urban areas in previous years in 12 out of the 16 studied species after
191 accounting for their imperfect detection (figure 1; electronic supplementary material, table S2). In
192 the four species with significant differences, three increased their occurrence and one decreased
193 it. As expected, most of the species (10) showed significant differences in their occurrence between
194 lockdown and non-urban checklists (electronic supplementary material, table S2). On average,
195 these species were approximately a 15% more common in the lockdown checklists than in the non-
196 urban checklists, confirming that most of the studied species were preferentially urban dwellers.

197 For most species (10), probability of detection was higher in lockdown checklists than in
198 historical urban ones, but this difference was not statistically significant in most cases (electronic
199 supplementary material, figure S2, table S3). However, most species were less detectable in non-
200 urban checklists than in urban ones.

201 As we predicted, detectability varied along the day in a non-linear way for all species (figure
202 2; electronic supplementary material, figure S3). Excepting two species, the pattern of daily
203 variation in detectability was significantly different among groups (electronic supplementary
204 material, table S4). A difference consistently found in most species was higher detectability in the
205 first hours of the morning during the lockdown compared to the urban records from previous years
206 (figure 2, 3). In most species, during the lockdown detectability peaked at dawn and decreased
207 until midday, while in the historical urban checklists the peak of detectability was around mid-
208 morning. In fact, the pattern of detectability along the day in the lockdown group resembled more
209 to the non-urban pattern than to the urban pattern in many species. Predicted detectability at dawn

210 by our models in the lockdown group was on average a 27% higher than in the urban group (sign
211 test: $Z=3.25$, p -value=0.001; figure 3).

212 As expected, in all but one species, chances of detection increased with longer surveys
213 (electronic supplementary material, figure S4, table S5). In most of them (11), such time effect was
214 significantly different among groups (electronic supplementary material, table S4), demonstrating
215 that increasing the sampling time does not always imply the same increase in a species' chances
216 of detection. For half of the species, time effect was significantly lower in the lockdown group than
217 in the historical urban group (mean reduction of 17%; electronic supplementary material, table S5).
218 This systematic reduction contrasts with the comparison of time effect between lockdown and non-
219 urban groups, where for nine species there were significant differences between both groups, but
220 such differences were disparate (mean change -0.8%; electronic supplementary material, table
221 S5).

222

223

224 **Discussion**

225 Birds did not occur in higher rates in towns and cities during the lockdown than before it, non-
226 supporting the hypothesis that birds moved into the human emptied urban areas [32]. As the
227 changes induced by the COVID-19 lockdown were drastic and sudden and did not last enough,
228 they probably did not allow for colonization processes. The few species with a significant increase
229 of their prevalence in urban surveys during lockdown were, interestingly, the ones that are mostly
230 urban. As these species are not present in large numbers away from urban areas, they could hardly
231 rely on non-urban source populations to occupy cities and towns during the lockdown.

232 Birds changed their detectability pattern as a consequence of the lockdown. In general,
233 there was an increase in detection probability, which was especially marked in the early morning.
234 As observed in non-urban habitats, detectability during the lockdown decreased from dawn
235 onwards, while at the same urban locations detectability was historically low at dawn and increased
236 until reaching a peak two or three hours later. It is interesting to note that the Eurasian blackbird, a
237 model species in urban ecology studies [8,9,13,49,50], was the only exception to this pattern.
238 Overall, many species showed a "wilder" pattern of detection during the lockdown in urban areas.

239 Urban birds during lockdown may have shown this detectability peak at dawn, typical of
240 non-urban habitats, because of a rapid behavioural response to adapt to the new environmental
241 conditions imposed by the COVID-19 measures [51]. Birds rely heavily on acoustic communication.
242 During reproduction, males sign to attract females and defend their territories, becoming highly
243 conspicuous and detectable. COVID-19 lockdown was imposed just at the beginning of the
244 breeding season, when singing activity was expected to be especially high [52]. Therefore, there

245 was a strong pressure to time singing activity to the optimal moment of the day. This moment is
246 dawn because the physical properties of the atmosphere enhance acoustic transmission [53,54]
247 and consequently, birds can reach the maximum audience. Thus, urban birds during the lockdown
248 may have advanced their main period of singing activity to dawn, increasing their detection at those
249 hours, similar to what is observed in non-urban areas.

250 During the lockdown, human presence and activities decreased drastically (electronic
251 supplementary material, figure S1, table S1), being this especially notable during rush hour, which
252 virtually disappeared [27]. During the spring in Spain, morning rush hour matches with the first
253 hours of light, when birds are expected to be especially communicative [36,50,55]. The dramatic
254 decrease in noise during the lockdown released early morning acoustic space that could be
255 recovered by the dawn chorus. Empirical and experimental evidence demonstrates that urban birds
256 avoid the masking effect of anthropogenic noise [9,10,35,36,56]. Our findings match these previous
257 studies, but instead of advancing the dawn chorus [36,49,50,57], our historical urban data suggests
258 that birds would delay their peak of activity (and consequently of detectability) to mid-morning. In
259 our study context, this can be explained because civil and solar time are heavily decoupled in Spain
260 since the country is located in the westernmost part of its time zone [58]. For this reason, if birds in
261 Catalonia advance their activities before sunrise, they would be still suffering an important overlap
262 with morning noisy human activities, such as commuting, school attendance, shop opening, etc.
263 [50,55]. Hence, the best option for birds would be to delay the peak of activity to after the morning
264 rush hour [56]. Moreover, most of the previous studies have been carried out in more northern
265 latitudes [9,49,59], where climate conditions can still be severe at night in early spring. Under these
266 circumstances, individual survival can be challenged by a strong nocturnal energy demand [60,61].
267 There, dawn singing can become a relevant and honest signal of phenotypic quality of males, as
268 only those individuals in best physical condition can undergo dawn fasting [49]. In Mediterranean
269 regions, where spring nights are mild, the role of dawn singing as signal of male quality might be
270 less important. Attracting mates would be the prime objective for singing and consequently, males
271 would be more pressed to place this activity when interference of anthropogenic noise is at its
272 lowest. Since sunrise, these lowest levels of noise are just after the morning rush hours (i.e. later
273 than 9 a.m.), when the air physical properties still keep sound attenuation and fluctuation low [53].

274 If birds have changed their behaviour, this adaptive, flexible behavioural response must
275 have been mediated by phenotypic plasticity. Lockdown was sudden and the environmental
276 scenario in urban areas changed radically from one day to the next (electronic supplementary
277 material, figure S1) [27]. This unprecedented social experiment imposed by the COVID-19 allowed
278 us to test and support the hypothesis of the high plasticity displayed by individuals living in urban
279 areas in order to cope with a constantly changing environment [6-8,51]. However, this fast adaptive
280 response might have been facilitated by a previous conditioning of birds to weekly rhythms of

281 human activities. Birds change their behaviour from working to weekend days [35,56,62] to match
282 with human behaviours. Therefore, birds could assimilate the lockdown as a very long and
283 especially peaceful weekend. Nevertheless, it would be interesting to explore the long-lasting
284 consequence at a community level of this environmental change [59]. Weekends are just two days
285 long, while strict people lockdown lasted for at least two months in most regions of Spain. One may
286 speculate that bolder and fast-adapting species would adapt their behaviour on a weekly basis.
287 However, during the lockdown, all species had enough time to adapt to the long-lasting new
288 conditions. In fact, as we have demonstrated, all of them modified their daily patterns of
289 detectability. Maybe the most urbanite species have benefited the least of this lockdown as their
290 boldness and higher human tolerance was no longer an advantage in empty cities.

291 In addition to the birds' rapid behavioural response to the anomalous environmental
292 conditions during the lockdown, observers had certainly enhanced opportunities to detect birds
293 during this period. Urban areas were quieter than usual [27-29], improving the chances of listening
294 the birds [32-34,53]. Moreover, absence of people outdoors allowed for the display of shy and
295 distrustful behaviours [6], facilitating bird observations, especially for those less singing species, as
296 the magpie *Pica pica* or the yellow-legged gull *Larus michahellis*. However, these improved
297 conditions for urban birdwatching were heavily constrained by the fact that observers were forced
298 to stay at their homes and their sampling area was reduced to what they could see from their
299 balconies, yards or rooftops. Therefore, improved detection was to some extent counterbalanced
300 by the limited scope from the survey sites. The observed effect in increased sampling time would
301 support this hypothesis, as we demonstrated that the discovery rate in most species was slower
302 during the lockdown surveys than in the historical urban ones.

303 The differences observed between urban and non-urban environments were expected as
304 habitat configuration and bird densities are patently different between them. In fact, populations of
305 urban exploiter birds show usually higher densities in cities than in rural or natural close areas [6-
306 8], facilitating their detection in urban areas. Such differences may have serious consequences for
307 monitoring schemes aiming to quantify wildlife occurrence and abundance by standardized
308 protocols, as the assumption of equal detectability under similar circumstances is usually violated
309 [33,44,45,63]. For instance, one sampling hour at dawn is not equivalent in terms of chances to
310 detect a species in urban and non-urban habitats. Traditional protocols assume that the best
311 moment to detect birds is early morning [53,64], which is actually true, but apparently only in natural
312 conditions without human disturbance, as we have demonstrated here (see figure 2). If the
313 detectability peak in most urban populations is reached at mid-morning, their abundance would be
314 systematically underestimated by usual sampling protocols based on early morning bird surveys.
315 As there is an increased awareness about the importance of urban bird populations [40], it is
316 necessary to ensure its accurate quantification, which may imply a redefinition of the most popular

317 current census techniques [8]. Additionally, in this work we demonstrated the utility of occupancy
318 models and the necessity to account for imperfect detection [45,46].

319 The COVID-19 lockdown is revealing the stress, noise and pollution present in urban areas
320 [25,26,28,29]. Under normal conditions, bird behaviour is altered and the possibility to enjoy the
321 natural values of our cities is notably diminished [8]. Our society should reflect on our urban lifestyle
322 and how it affects welfare of urban fauna and jeopardizes its conservation. As the world is becoming
323 more urbanized and animals will be forced to live more often in anthropogenic environments [6-8],
324 one way to ensure their adaptation as urban dwellers would be by reducing our noisier and more
325 disturbing activities. Most importantly, not only urban populations of non-human animals would be
326 benefited, but also ourselves from quieter, more peaceful and less polluted cities.

327

328

329 **Authors' contributions**

330 G.G. has the original idea and designed the monitoring program. O.G. and G.G. formulated formal
331 hypotheses, and collected and arranged the necessary data for the study. O.G. analysed data.
332 O.G. wrote the manuscript, with contributions from all the authors.

333 **Competing Interest**

334 The authors declare no competing interests.

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342

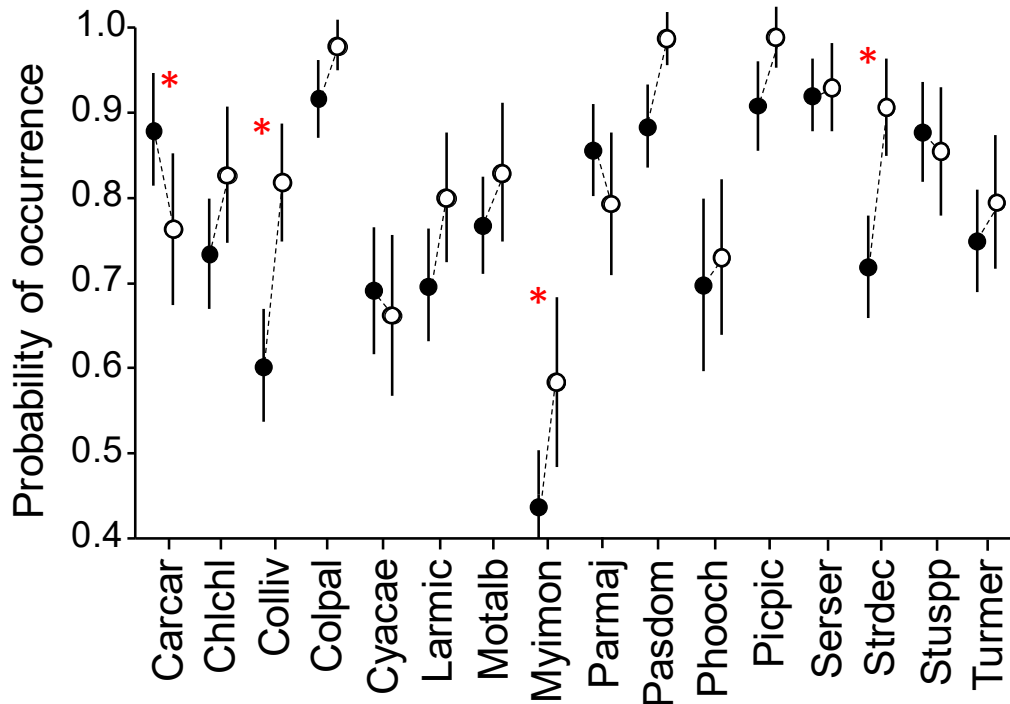
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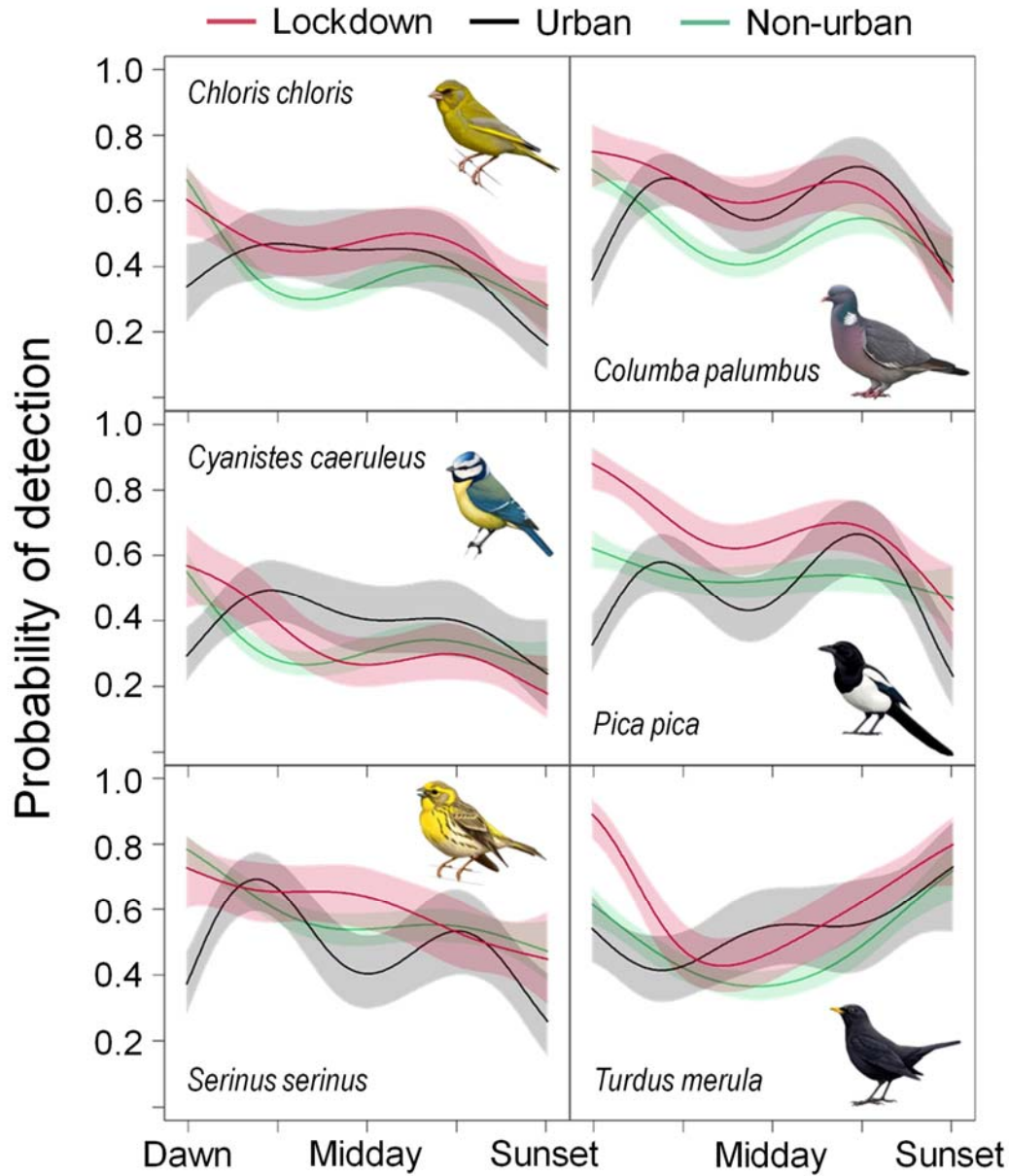


487

488 **Figure 1.** Probability of occurrence of birds in urban areas before (2015-19, black dots) and during
489 (2020; white dots) the COVID-19 lockdown. Asterisks indicate significant differences (p -value <
490 0.05). Error bars denote 95% confidence intervals. Acronyms for the species: Carcar *Carduelis*
491 *carduelis*, Chlchl *Chloris chloris*, Colliv *Columba livia*, Colpal *Columba palumbus*, Cyacae
492 *Cyanistes caeruleus*, Larmic *Larus michahellis*, Motalb *Motacilla alba*, Myimon *Myiopsitta*
493 *monachus*, Parmaj *Parus major*, Pasdom *Passer domesticus*, Phooch *Phoenicurus ochruros*,
494 Picpic *Pica pica*, Serser *Serinus serinus*, Strdec *Streptopelia decaocto*, Stuspp *Sturnus* spp.,
495 Turner *Turdus merula*.

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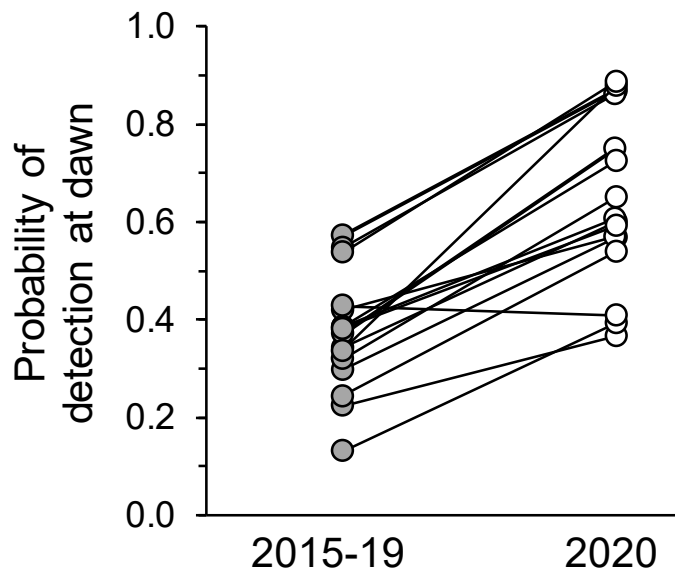
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499 **Figure 2.** Variation in the probability of detection along the day for each group of data (collected
500 during the lockdown, collected historically in urban sites, and collected in non-urban environments).
501 Shaded areas represent the 95% confidence intervals. See fig. S2 in the electronic supplementary
502 material for the rest of species. Bird illustrations by Martí Franch/Catalan Ornithological Institute.
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507 **Figure 3.** Probability of detection in urban environments at dawn for the 16 studied bird species
508 before (2015-19) and during (2020) COVID-19 lockdown.

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510

511 **Electronic supplementary information for this manuscript include the following:**

512

513 **Figure S1.** Variation in people's visiting habits to six categories of places in Catalonia between
514 March 1st and July1st.

515 **Figure S2.** Probability of detection of birds in urban areas before and during the COVID-19
516 lockdown

517 **Figure S3.** Variation in the probability of detection along the day for each group of data in 10 studied
518 species.

519 **Figure S4.** Effect of sampling time on the probability of detection for each group of data.

520

521 **Table S1.** Percentage variation in people's visiting habits to six categories of places between March
522 15th and April 14th in Catalonia and its provinces.

523 **Table S2.** Results of the occupancy part of the occupancy models.

524 **Table S3.** Results for group effects on the detection part of the occupancy models.

525 **Table S4.** *P*-values for the effects of hour, the interaction between hour and group, and the
526 interaction between time and group in the occupancy models

527 **Table S5.** Results for the effects of the interaction between time and group on the detection part of
528 the occupancy models.

529

ELECTRONIC SUPPLEMENTARY INFORMATION for:

RAPID BEHAVIOURAL RESPONSE OF URBAN BIRDS TO COVID-19 LOCKDOWN

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Figures S1 to S5

Tables S1 to S4

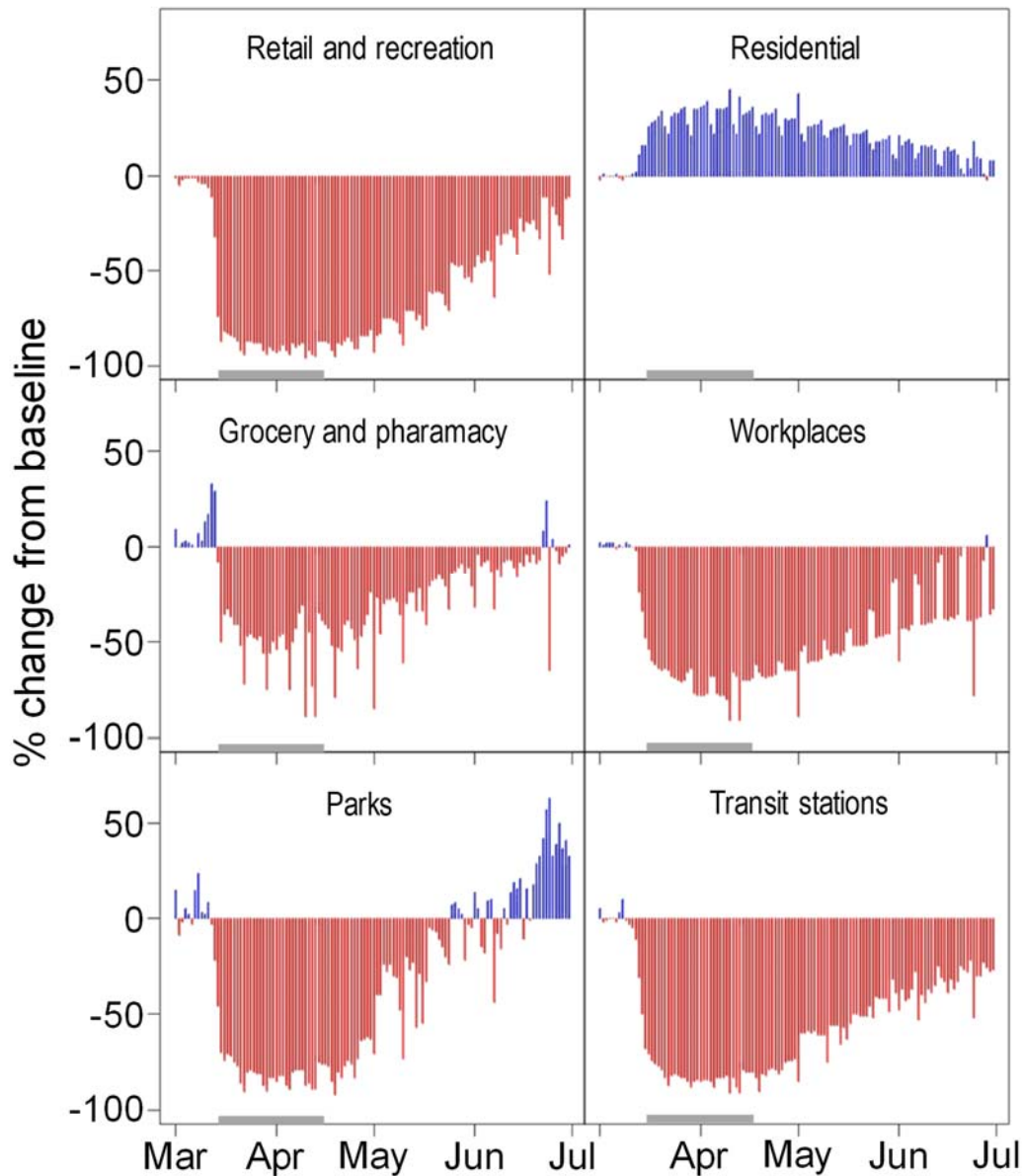


Figure S1. Variation in people's visiting habits to six categories of places in Catalonia between March 1st and July 1st. These graphs can be interpreted as a proxy for human activity. To improve visualization, negative deviations are shown in red and positive in blue. The grey band on to the x axis shows the period of bird data collection. The emergency state in Spain due to COVID-19 lasted from March 15th to June 21st. Note the forced strict lockdown during the first two months of this period and the following progressive recovery of activities. Figure data are freely available from "Google COVID-19 Community Mobility Reports" (<https://www.google.com/covid19/mobility/>; Accessed: August 25th 2020). Baseline level (i.e., 0%) has been calculated as the average from January 3rd to February 6th 2020 (visit the previous link for further technical details on data calculations).

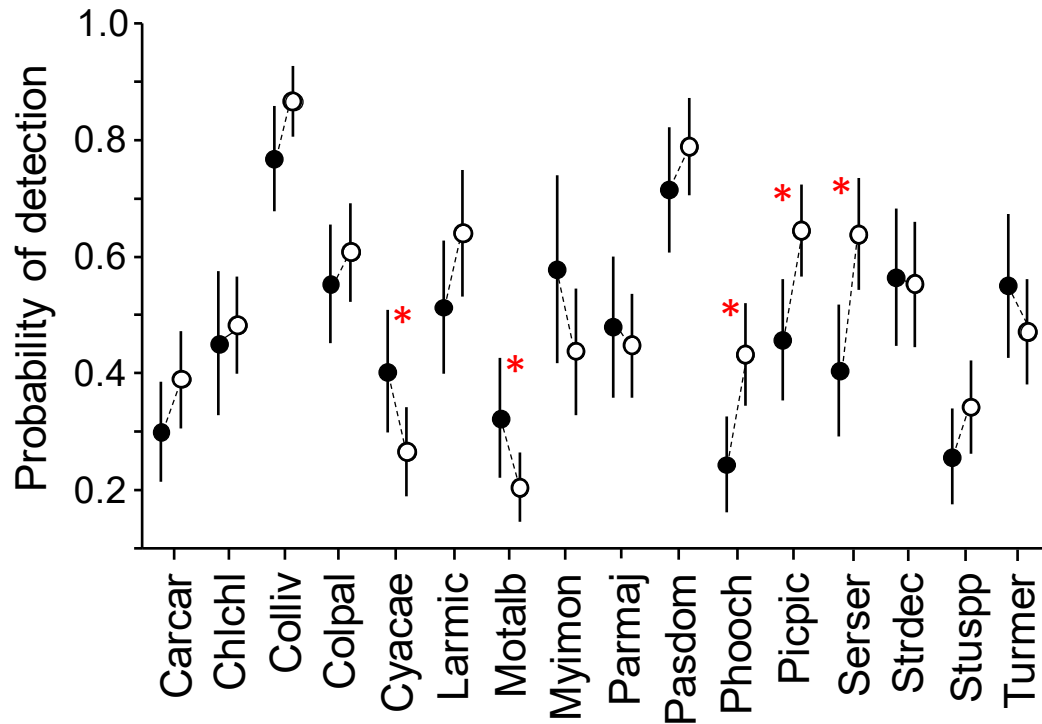


Figure S2. Probability of detection of birds in urban areas before (2015-19, black dots) and during (2020; white dots) the COVID-19 lockdown. Asterisks indicate significant differences (p -value < 0.05). Error bars denote 95% confidence intervals. Acronyms for the species: Carcar *Carduelis carduelis*, Chlchl *Chloris chloris*, Colliv *Columba livia*, Colpal *Columba palumbus*, Cyacae *Cyanistes caeruleus*, Larmic *Larus michahellis*, Motalb *Motacilla alba*, Myimon *Myiopsitta monachus*, Parmaj *Parus major*, Pasdom *Passer domesticus*, Phooch *Phoenicurus ochruros*, Picpic *Pica pica*, Serser *Serinus serinus*, Strdec *Streptopelia decaocto*, Stuspp *Sturnus* spp., Turner *Turdus merula*.

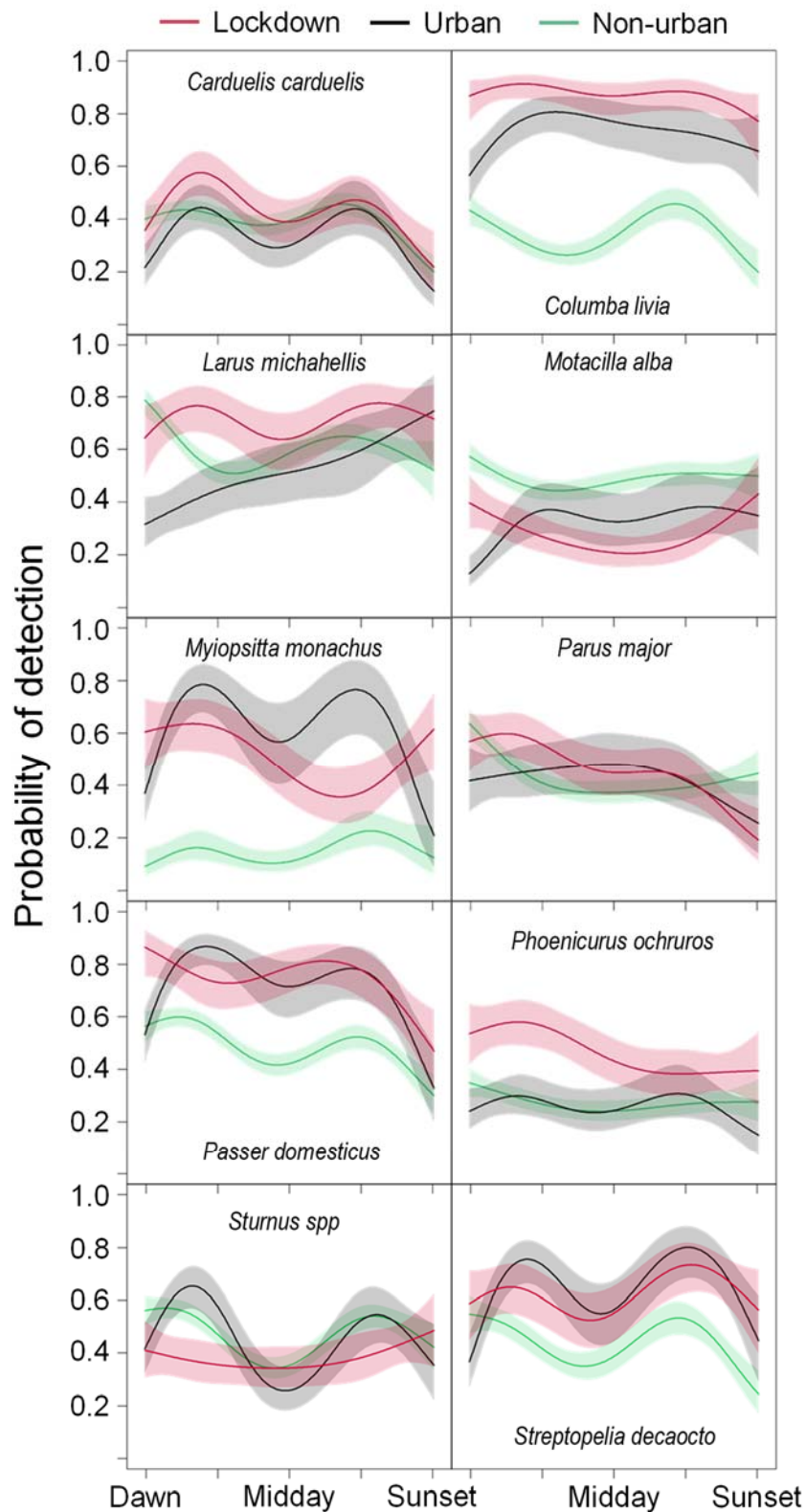


Figure S3. Variation in the probability of detection along de day for each group of data (collected during the lockdown, collected historically in urban sites, and collected in non-urban environments). Shaded areas represent the 95% confidence intervals.

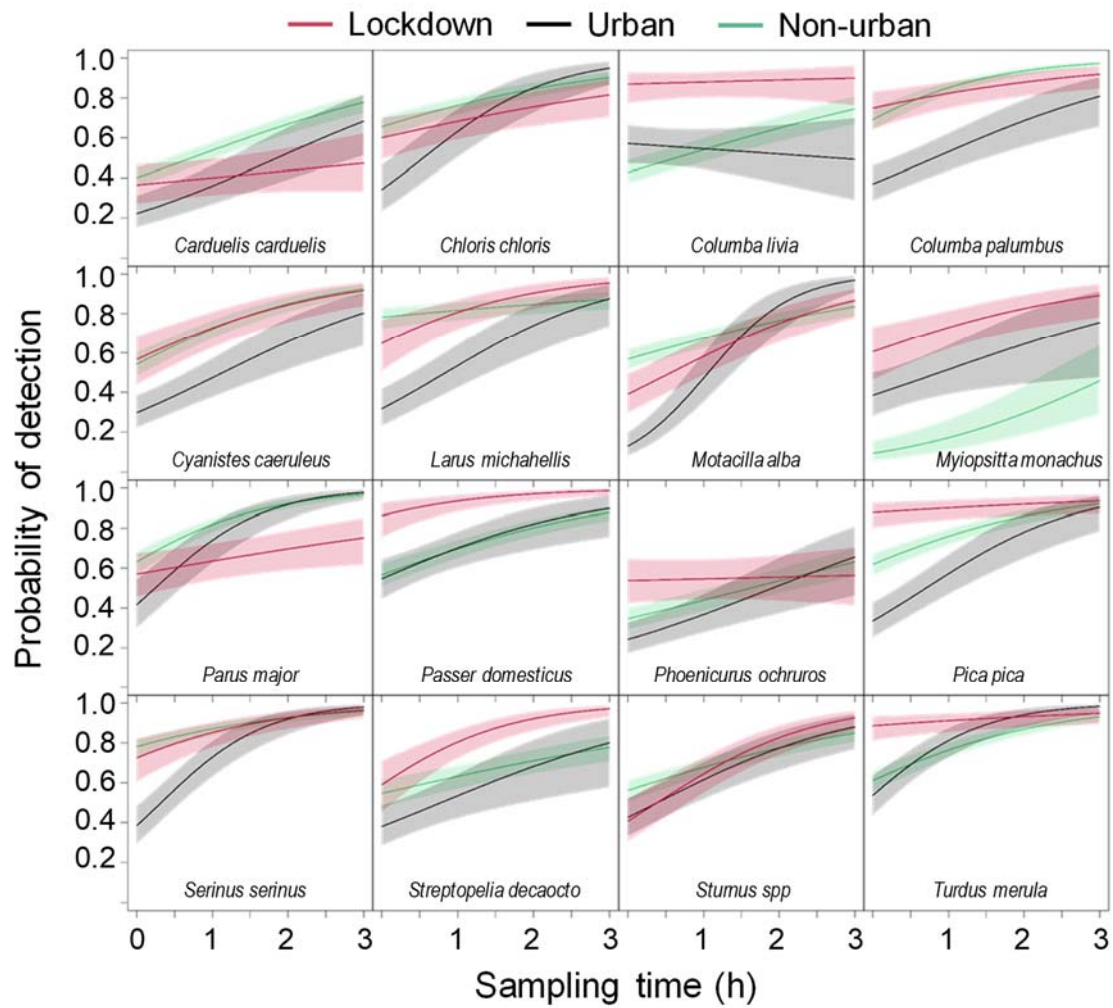


Figure S4. Effect of sampling time on the probability of detection for each group of data (collected during the lockdown, collected historically in urban sites, and collected in non-urban environments). Predictions done for surveys started at dawn. Shaded areas represent the 95% confidence intervals.

Table S1. Percentage variation in people's visiting habits to six categories of places between March 15th and April 14th in Catalonia and its provinces. The values show the average of daily deviations from the baseline level. Outdoor sites experienced a severe decline, while people stayed more at home. See figure S1 for details.

| Place | Catalonia | Barcelona | Girona | Lleida | Tarragona |
|-----------------------|------------------|------------------|---------------|---------------|------------------|
| Retail and recreation | -89.56 | -90.41 | -88.13 | -86.19 | -88.56 |
| Grocery and pharmacy | -51.28 | -53.13 | -51.59 | -52.72 | -53.06 |
| Parks | -81.34 | -83.59 | -74.22 | -62.63 | -76.63 |
| Transit Stations | -82.41 | -82.44 | -80.41 | -78.88 | -83.00 |
| Workplaces | -70.41 | -72.22 | -64.22 | -61.81 | -65.69 |
| Residential | 31.25 | 32.03 | 28.41 | 26.00 | 27.84 |

Table S2. Results of the occupancy part of the occupancy models. *Naïve* columns show the proportion of sites where the species was found during the surveys, i.e. the raw occurrence without correction for the imperfect detection of birds. *Estimate* columns show the estimated occurrence of the studied species once the detectability was accounted for. As expected, estimated occupancy was always higher than the observed (i.e., *naïve*), demonstrating both imperfect and variable detection of birds. For the urban and non-urban groups of historical data (2015-19), a column with the differences (Δ) between their estimates and the lockdown group (2020 data) estimates is provided. The *p*-values for these differences are also shown. Values <0.05 are in bold. At the bottom of the table, the average and standard deviation of the studied species are given.

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| Species | Lockdown (2020) | | Urban (2015-19) | | Δ | <i>p</i> | Non-urban (2015-19) | | Δ | <i>p</i> |
|--|-----------------|---------------|-----------------|---------------|--------------------------|---------------|---------------------|---------------|------------------------------|-------------------|
| | naïve | estimate | naïve | estimate | lockdown-urban estimates | | naïve | estimate | lockdown-non-urban estimates | |
| European goldfinch, <i>Carduelis carduelis</i> | 0.5973 | 0.7635 | 0.5634 | 0.8795 | -0.1160 | 0.0457 | 0.5401 | 0.8435 | -0.0800 | 0.0677 |
| Greenfinch, <i>Carduelis chloris</i> | 0.6779 | 0.8263 | 0.5537 | 0.7342 | 0.0921 | 0.1020 | 0.4055 | 0.6759 | 0.1504 | 0.0056 |
| Rock pigeon, <i>Columba livia</i> | 0.7718 | 0.8176 | 0.4683 | 0.6024 | 0.2152 | 0.0001 | 0.2501 | 0.4582 | 0.3593 | <0.0001 |
| Wood pigeon, <i>Columba palumbus</i> | 0.8859 | 0.9781 | 0.7463 | 0.9158 | 0.0623 | 0.0695 | 0.6645 | 0.8511 | 0.1270 | 0.0044 |
| Blue tit, <i>Cyanistes caeruleus</i> | 0.5235 | 0.6616 | 0.4693 | 0.6906 | -0.0290 | 0.6364 | 0.3792 | 0.6144 | 0.0472 | 0.3670 |
| Yellow-legged gull, <i>Larus michahellis</i> | 0.7481 | 0.8001 | 0.5428 | 0.6970 | 0.1031 | 0.0565 | 0.4929 | 0.6851 | 0.1150 | 0.0159 |
| White wagtail, <i>Motacilla alba</i> | 0.6376 | 0.8292 | 0.6049 | 0.7675 | 0.0618 | 0.2520 | 0.4899 | 0.7377 | 0.0915 | 0.0735 |
| Monk parakeet, <i>Myiopsitta monachus</i> | 0.5041 | 0.5833 | 0.3580 | 0.4373 | 0.1460 | 0.0187 | 0.1300 | 0.3243 | 0.2590 | <0.0001 |
| Great tit, <i>Parus major</i> | 0.6443 | 0.7924 | 0.6642 | 0.8548 | -0.0624 | 0.2060 | 0.6126 | 0.8221 | -0.0296 | 0.4840 |
| House sparrow, <i>Passer domesticus</i> | 0.9195 | 0.9863 | 0.7677 | 0.8837 | 0.1025 | 0.0633 | 0.5480 | 0.7649 | 0.2213 | 0.0092 |
| Black redstart, <i>Phoenicurus ochruros</i> | 0.5705 | 0.7306 | 0.3585 | 0.6975 | 0.0330 | 0.6350 | 0.2965 | 0.6426 | 0.0879 | 0.1090 |
| Magpie, <i>Pica pica</i> | 0.8658 | 0.9879 | 0.7311 | 0.9076 | 0.0803 | 0.1770 | 0.5328 | 0.7040 | 0.2839 | 0.0215 |
| Serin, <i>Serinus serinus</i> | 0.8389 | 0.9295 | 0.7732 | 0.9198 | 0.0097 | 0.7790 | 0.6200 | 0.7851 | 0.1444 | 0.0017 |
| Collared dove, <i>Streptopelia decaocto</i> | 0.8322 | 0.9052 | 0.5980 | 0.7189 | 0.1863 | 0.0005 | 0.2842 | 0.4578 | 0.4474 | <0.0001 |
| Starling, <i>Sturnus spp</i> | 0.6980 | 0.8538 | 0.6479 | 0.8774 | -0.0235 | 0.6250 | 0.5010 | 0.7480 | 0.1058 | 0.0342 |
| Eurasian blackbird, <i>Turdus merula</i> | 0.6846 | 0.7941 | 0.6073 | 0.7494 | 0.0447 | 0.3890 | 0.5754 | 0.7853 | 0.0088 | 0.8380 |
| <i>mean</i> | <i>0.7125</i> | <i>0.8275</i> | <i>0.5909</i> | <i>0.7708</i> | <i>0.0566</i> | | <i>0.4577</i> | <i>0.6813</i> | <i>0.1462</i> | |
| <i>SD</i> | <i>0.1308</i> | <i>0.1139</i> | <i>0.1316</i> | <i>0.1332</i> | <i>0.0882</i> | | <i>0.1521</i> | <i>0.1516</i> | <i>0.1399</i> | |

Table S3. Results for group effects on the detection part of the occupancy models. For the urban and non-urban groups, a column with the differences (Δ) between their estimates and the lockdown group estimates is provided. The p -values for these differences are also shown. Values <0.05 are in bold. At the bottom of the table, the average and standard deviation of the studied species are given.

| Species | Lockdown (2020) | Urban (2015-19) | Δ lockdown- urban | p | Non-urban (2015-19) | Δ lockdown- non-urban | p |
|--|--------------------|--------------------|-----------------------------|---------------|------------------------|---------------------------------|-------------------|
| European goldfinch, <i>Carduelis carduelis</i> | 0.3897 | 0.3002 | 0.0895 | 0.1444 | 0.3948 | -0.0051 | 0.9133 |
| Greenfinch, <i>Carduelis chloris</i> | 0.4820 | 0.4517 | 0.0303 | 0.6938 | 0.3469 | 0.1351 | 0.0049 |
| Rock pigeon, <i>Columba livia</i> | 0.8665 | 0.7688 | 0.0978 | 0.0716 | 0.3289 | 0.5376 | <0.0001 |
| Wood pigeon, <i>Columba palumbus</i> | 0.6070 | 0.5535 | 0.0534 | 0.4289 | 0.4302 | 0.1768 | 0.0005 |
| Blue tit, <i>Cyanistes caeruleus</i> | 0.2658 | 0.4035 | -0.1377 | 0.0364 | 0.3036 | -0.0378 | 0.4140 |
| Yellow-legged gull, <i>Larus michahellis</i> | 0.6405 | 0.5131 | 0.1274 | 0.1182 | 0.5865 | 0.0540 | 0.3912 |
| White wagtail, <i>Motacilla alba</i> | 0.2052 | 0.3233 | -0.1181 | 0.0419 | 0.4716 | -0.2664 | <0.0001 |
| Monk parakeet, <i>Myiopsitta monachus</i> | 0.4378 | 0.5773 | -0.1395 | 0.1640 | 0.1097 | 0.3281 | <0.0001 |
| Great tit, <i>Parus major</i> | 0.4479 | 0.4803 | -0.0324 | 0.6730 | 0.3717 | 0.0762 | 0.1290 |
| House sparrow, <i>Passer domesticus</i> | 0.7891 | 0.7150 | 0.0740 | 0.2820 | 0.4224 | 0.3667 | <0.0001 |
| Black redstart, <i>Phoenicurus ochruros</i> | 0.4321 | 0.2437 | 0.1885 | 0.0028 | 0.2414 | 0.1907 | <0.0001 |
| Magpie, <i>Pica pica</i> | 0.6452 | 0.4573 | 0.1879 | 0.0058 | 0.5259 | 0.1192 | 0.0169 |
| Serin, <i>Serinus serinus</i> | 0.6385 | 0.4040 | 0.2345 | 0.0027 | 0.5393 | 0.0993 | 0.0810 |
| Collared dove, <i>Streptopelia decaocto</i> | 0.5525 | 0.5649 | -0.0124 | 0.8790 | 0.3876 | 0.1648 | 0.0092 |
| Starling, <i>Sturnus spp</i> | 0.3426 | 0.2574 | 0.0852 | 0.1510 | 0.3534 | -0.0108 | 0.8140 |
| Eurasian blackbird, <i>Turdus merula</i> | 0.4717 | 0.5500 | -0.0783 | 0.3100 | 0.3683 | 0.1034 | 0.0394 |
| <i>mean</i> | <i>0.5134</i> | <i>0.4728</i> | <i>0.0406</i> | | <i>0.3864</i> | <i>0.1270</i> | |
| <i>SD</i> | <i>0.1789</i> | <i>0.1508</i> | <i>0.1182</i> | | <i>0.1167</i> | <i>0.1830</i> | |

Table S4. *P*-values for the effects of hour, the interaction between hour and group, and the interaction between time and group in the occupancy models. *f* refers to the basis function used to model the non-linear effect of hour on bird detectability.

| Species | <i>f</i> (hour) | <i>f</i> (hour)*group | time*group |
|--|-----------------|-----------------------|------------|
| European goldfinch, <i>Carduelis carduelis</i> | <0.00001 | 0.14308 | 0.04767 |
| Greenfinch, <i>Carduelis chloris</i> | <0.00001 | 0.00291 | 0.02069 |
| Rock pigeon, <i>Columba livia</i> | 0.00754 | 0.00137 | 0.04995 |
| Wood pigeon, <i>Columba palumbus</i> | <0.00001 | 0.00001 | 0.93144 |
| Blue tit, <i>Cyanistes caeruleus</i> | <0.00001 | 0.00002 | 0.94749 |
| Yellow-legged gull, <i>Larus michahellis</i> | 0.00256 | 0.00001 | 0.00328 |
| White wagtail, <i>Motacilla alba</i> | 0.09982 | 0.00018 | 0.00001 |
| Monk parakeet, <i>Myiopsitta monachus</i> | 0.00455 | 0.00008 | 0.83579 |
| Great tit, <i>Parus major</i> | <0.00001 | 0.00113 | 0.00023 |
| House sparrow, <i>Passer domesticus</i> | <0.00001 | 0.00073 | 0.64000 |
| Black redstart, <i>Phoenicurus ochruros</i> | 0.01681 | 0.11286 | 0.05613 |
| Magpie, <i>Pica pica</i> | 0.00048 | 0.00009 | 0.02690 |
| Serin, <i>Serinus serinus</i> | <0.00001 | 0.00027 | 0.01794 |
| Collared dove, <i>Streptopelia decaocto</i> | 0.00003 | 0.00039 | 0.02776 |
| Starling, <i>Sturnus spp</i> | <0.00001 | 0.00594 | 0.04043 |
| Eurasian blackbird, <i>Turdus merula</i> | <0.00001 | 0.00464 | 0.00649 |

Table S5. Results for the effects of the interaction between time and group on the detection part of the occupancy models. Columns lockdown, urban and non-urban show the probability of detection of the species after 1 hour of survey. A probability of 0.5 means that the observers may or may not detect the species with the same probability (i.e. flat slope). Except the rock pigeon in urban checklists, in all cases, as expected, these probabilities were above 0.5 (i.e., increased probability of detection of a species throughout sampling time). For the lockdown estimates, the *p*-value testing whether or not this slope was different from 0.5 is shown. However, in urban and non-urban groups, the *p*-value shows whether or not these slopes differed from the lockdown group. The difference (Δ) between these slopes is also provided to enhance result interpretation.

| Species | Lockdown (2020) | <i>p</i> | Urban (2015-19) | Δ lockdown- urban | <i>p</i> | Non-urban (2015-19) | Δ lockdown- non-urban | <i>p</i> |
|--|--------------------|-------------------|--------------------|-----------------------------|-------------------|------------------------|---------------------------------|-------------------|
| European goldfinch, <i>Carduelis carduelis</i> | 0.5372 | 0.1908 | 0.6620 | -0.1248 | 0.0019 | 0.6345 | -0.0973 | 0.0012 |
| Greenfinch, <i>Carduelis chloris</i> | 0.5889 | 0.0023 | 0.7675 | -0.1786 | 0.0002 | 0.6282 | -0.0394 | 0.2063 |
| Rock pigeon, <i>Columba livia</i> | 0.5250 | 0.6086 | 0.4735 | 0.0515 | 0.4093 | 0.6112 | -0.0862 | 0.0861 |
| Wood pigeon, <i>Columba palumbus</i> | 0.6805 | <0.0001 | 0.6757 | 0.0048 | 0.9130 | 0.6920 | -0.0115 | 0.7310 |
| Blue tit, <i>Cyanistes caeruleus</i> | 0.6710 | <0.0001 | 0.6800 | -0.0091 | 0.8400 | 0.6863 | -0.0154 | 0.6480 |
| Yellow-legged gull, <i>Larus michahellis</i> | 0.6936 | 0.0001 | 0.7120 | -0.0184 | 0.7438 | 0.5537 | 0.1399 | 0.0061 |
| White wagtail, <i>Motacilla alba</i> | 0.6839 | <0.0001 | 0.8566 | -0.1726 | 0.0001 | 0.6107 | 0.0733 | 0.0130 |
| Monk parakeet, <i>Myiopsitta monachus</i> | 0.6371 | 0.0008 | 0.6298 | 0.0073 | 0.9110 | 0.6685 | -0.0314 | 0.4990 |
| Great tit, <i>Parus major</i> | 0.5682 | 0.0197 | 0.7988 | -0.2306 | <0.0001 | 0.7243 | -0.1560 | <0.0001 |
| House sparrow, <i>Passer domesticus</i> | 0.6983 | 0.0001 | 0.6602 | 0.0381 | 0.5410 | 0.6384 | 0.0599 | 0.2120 |
| Black redstart, <i>Phoenicurus ochruros</i> | 0.5089 | 0.7519 | 0.6446 | -0.1357 | 0.0019 | 0.5965 | -0.0876 | 0.0041 |
| Magpie, <i>Pica pica</i> | 0.5584 | 0.0532 | 0.7258 | -0.1674 | 0.0005 | 0.6599 | -0.1014 | 0.0021 |
| Serin, <i>Serinus serinus</i> | 0.6862 | <0.0001 | 0.8111 | -0.1248 | 0.0114 | 0.6536 | 0.0327 | 0.4020 |
| Collared dove, <i>Streptopelia decaocto</i> | 0.7411 | <0.0001 | 0.6514 | 0.0897 | 0.1240 | 0.5881 | 0.1530 | 0.0006 |
| Starling, <i>Sturnus spp</i> | 0.7244 | <0.0001 | 0.6823 | 0.0421 | 0.3050 | 0.6212 | 0.1032 | 0.0013 |
| Eurasian blackbird, <i>Turdus merula</i> | 0.5705 | 0.0234 | 0.7883 | -0.2178 | <0.0001 | 0.6726 | -0.1021 | 0.0018 |

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