## **RESEARCH ARTICLE**

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# Do Ecological Valid Stop Signals Aid Detour Performance? A Comparison of Four Bird Species.

#### Cite as:

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This version of the article has been peer-reviewed and recommended by (https://doi.org/)

## Abstract

Response inhibition, or the stopping of actions, is considered a key component of flexible and adaptive behaviour. Across fields, response inhibition is often treated as a unitary cognitive mechanism. However, we propose that response inhibition consists of a chain of cognitive processes, including the detection of a stimulus, the selection of an appropriate behaviour (go or stop), and the implementation of it (execution or inhibition of a motor response). From this, we propose that individual variation in response inhibition can arise at the early signal detection stage. This idea was tested in a detour barrier task, which is one of the most popular tools to study response inhibition in non-human animals. The role of signal detection in detour tasks has been largely neglected, with a few notable exceptions. We therefore partially replicated two previous studies that manipulated the perceptual characteristics of the barrier, while addressing some conceptual and methodological shortcomings of the original work. Specifically, we compared how detour performance of four bird species (i.e., white leghorn chickens, Japanese quails, herring gulls and domestic canaries) is differently influenced by vertical- and horizontal-barred barriers. In contrast to the previous work, performance on the detour task did not improve when the perceptual characteristics of the barrier matched the ecological niche of the species. However, all species, showed some level of learning, as evidenced by shorter detour latencies (except in herring gulls) and fewer persisting attempts. These findings highlight the need for replication studies and emphasize on the importance of improving methodological and conceptual design factors to further investigate the underlying mechanisms of response inhibition in animals. Preregistered Stage 1 protocol: https://osf.io/qvxgh (date of in-principle acceptance: 20/03/2023).

Keywords: Response Inhibition; Stop-Signal Detection; Comparative Approach; Birds; Detour

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# Introduction

Response inhibition (RI) refers to stopping or cancelling actions that are no-longer relevant, inappropriate, 16 or overly risky (Verbruggen and Logan, 2008b, 2017). It is often regarded as a critical component of flexible 17 and adaptive behaviour (Verbruggen and Logan, 2008b). For example, animals living in urban environments 18 must often inhibit no-longer relevant behaviours when confronted with environmental conditions that differ 19 significantly from their ancestral ones (Lee and Thornton, 2021); lower-ranked animals need to inhibit inap-20 propriate, disobedient behaviour in the presence of dominant animals (Johnson-Ulrich and Holekamp, 2020); 21 and foraging animals must refrain from approaching a food source when this action becomes overly risky 22 due to the emergence of a predator (Tvardíková and Fuchs, 2012). These examples demonstrate that RI (or a 23 lack thereof) can have important fitness consequences (e.g., the animals may be predated if they fail to stop 24 foraging when the predator emerges). 25

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One of the most popular tasks to study RI in animals is the detour task (Kabadayi, Jacobs, et al., 2017; Miller 27 and Tallarico, 1974: Regolin, Giorgio, et al., 1995: Regolin, Vallortigara, et al., 1994: Van Horik, Beardsworth, 28 Laker, Whiteside, et al., 2020; Van Horik, Langley, et al., 2018; Zucca, Antonelli, et al., 2005; Zucca and Sovrano, 29 2008). In this task, the direct path to a motivationally-salient stimulus (e.g., food or a social companion) is 30 blocked by a barrier or cylinder. Animals have to inhibit their prepotent response to go directly for the re-31 ward (as the direct path is blocked), and instead make a detour around the barrier or cylinder to obtain the 32 reward. Detour tasks have been used in non-human animals, such as birds, to study e.g., how the social or 33 physical environment shapes RI. For example, wild Australian magpies (Gymnorhing tibicen) demonstrated 34 superior detour performance when reared in large compared with small social groups (Ashton et al., 2018). 35 Another study found that pheasants (Phasianus colchicus) showed superior detour performance when reared 36 in spatially unpredictable compared with predictable environments (Van Horik, Beardsworth, Laker, Langley, 37 et al., 2019). Combined, these findings suggest that RI development is facilitated in e.g., environments with 38 high social demands or environments that promote the expression of diverse foraging strategies. 39 40

Typically, performance in the detour task has been linked to the variation in the effectiveness of a single 41 cognitive control function, 'response inhibition', or more generally, 'inhibitory control' (which is an umbrella 42 term for various types of inhibition, which may or may not be related to each other; Bari and Robbins, 2013). 43 However, by referring to general ill-defined cognitive constructs such as RI (or even worse, a general umbrella 44 term such as 'inhibitory control'), we do not explain the underlying cognitive mechanisms or building blocks 45 of RI (Verbruggen, McLaren, et al., 2014), as the explanation is 'just as mysterious as the thing it is supposed 46 to explain' (Press et al., 2022). To address this issue in the human RI literature, a theoretical framework of RI 47 has been proposed (Verbruggen, McLaren, et al., 2014). Based on empirical work in humans, primates, and 48 rodents, the authors of the framework proposed that RI involves a chain of processes. More specifically, RI would involve at least three basic processes: the detection of a 'stop signal' (detection), the stochastic accu-50 mulation of information (selection), and suppression of the motoric output (implementation). Furthermore, 51 these core processes can be modulated by a set of processes that take place on shorter (seconds, minutes, 52 hours or days) and longer (months or years) timescales. Depending on the species, this can involve, e.g., out-53 come monitoring, anticipatory adjustments, and both short-term and long-term learning. Here we argue that 54 some of these cognitive processes play a role in RI across species (without assuming a one-to-one mapping 55 for the full processing chain). In particular, in the present study we propose that one of these core processes, 56 namely stop-signal detection, is a crucial (but largely ignored) building block of RI across species, including 57 avian species. 58

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## The Crucial Role of Stop-Signal Detection

Several lines of evidence indicate that signal detection may play a critical role in RI, particularly in humans 60 and non-human primates. (at least in humans). For example, several behavioural studies reveal that RI is 61 impaired when visual distractors occur in the environment (Verbruggen, Stevens, et al., 2014), or when stop 62 signals are hard to perceive (Van Der Schoot et al., 2005). Neurophysiological and computational work also 63 demonstrated that early perceptual processing of potential stop signals (which could be e.g., an obstacle, or 64 in case of humans, a red light) determines to a large extent whether individuals can successfully inhibit a 65 response or not (Bekker et al., 2005; Boehler et al., 2009; Elchlepp and Verbruggen, 2017; Pani et al., 2018; 66 Salinas and Stanford, 2013). 67

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Thus, it appears that RI may largely depend on the outcome of perceptual processes. However, the crucial 69 contribution of these processes to successful RI is rarely acknowledged or studied in the non-human animal 70 cognition domain, with a few notable exceptions. For example, researchers found that avian RI was improved 71 when the visibility of a stop signal (i.e., a predator) was improved (e.g., when the predator occurred against a 72 white background, in bright light, or in short grass) (Devereux et al., 2006; Nebel et al., 2019). Other studies 73 suggested that RI in the detour barrier task is affected by the perceptual characteristics of the barrier (i.e., 74 the type of stop signal). For example, Regolin and colleagues (Exp 1) (1994) exposed 2-day old white leghorn 75 chickens (Gallus gallus domesticus) to a variety of barrier types. These included a barrier with vertical bars or 76 stripes, and a barrier with horizontal bars. The authors found that RI performance was impaired (i.e., the 77 time required to successfully detour around the barrier) when faced with vertical-bar barriers compared with 78 horizontal-bar barriers. Both vertical- and horizontal-bar barriers occluded the reward behind the barrier to a similar degree (i.e., 20% compared with a fully transparent barrier). Thus, the differences between these 80 two barrier types cannot be attributed to differences in physical reward occlusion. Instead, the authors sug-81 gested two alternative potential explanations for this asymmetrical effect, namely (1) the degree of subjective 82 occlusion and (2) the ecological validity of stop signals. 83 84

First, despite equal reward occlusion by each barrier type, the behavioural repertoire of ground-moving an-85 imals consists primarily of horizontal movements (e.g., walking, running). Consequentially, these animals can 86 'subjectively' perceive a reward as less occlusive (i.e., more visible) with vertical- than horizontal-bar barriers 87 (making it harder to inhibit the response to go directly for the reward) (Kabadayi, Bobrowicz, et al., 2018). However, follow-up experiments in which the occlusion of the reward was directly manipulated, were inconsistent 89 with this 'subjective occlusion' account (Regolin, Vallortigara, et al., 1994). Second, differential performance 90 between vertical- and horizontal-bar barriers might be due to the ecological niche of the species. Gallinaceous 91 birds such as chickens are mainly terrestrial animals that have occupied niches that consist of penetrable long 92 grass and twigs. Regolin et al. (1994) therefore argued that it might be harder for gallinaceous birds to detour 93 around vertical- than horizontal-bar barriers, as the former would mimic the penetrable vertical vegetation 94 of their ecological niche (whereas in the detour task, the vertical-bar barrier is of course, not penetrable). 95 96

<sup>97</sup> Zucca and colleagues (2005) further investigated this ecological-niche hypothesis by comparing detour per-<sup>98</sup> formance in another gallinaceous bird species, hybrid (Japanese) quails (*Coturnix coturnix x C. japonica*), with <sup>99</sup> performance in two species with a substantially different ecological niche, namely yellow-legged gulls (*Larus* <sup>100</sup> *michahellis*) <sup>1</sup> and domestic canaries (*Serinus canaria*). They used a variant of the detour task with multiple <sup>101</sup> compartments and again, vertical- and horizontal-bar barriers (14% reward occlusion compared with a trans-<sup>102</sup> parent barrier<sup>2</sup>). In this study, the authors considered both probability of a correct response (i.e., going to the

<sup>&</sup>lt;sup>1</sup>Zucca and colleagues (2005) mention in their paper that they tested herring gulls, but used the scientific name of the Caspian gull *Larus cachinnans*. The taxonomy of the genus Larus has been updated over the years. Given that the authors mentioned that the gulls were from a breeding colony in Trieste (Italy), we assume they tested yellow-legged gulls, which were formerly treated as a subspecies of the herring gull (as was the Caspian gull).

<sup>&</sup>lt;sup>2</sup>Zucca and colleagues (2005) mention that they used a 23 x 26 cm barrier, barrier lines/rods with a width of 0.2 cm, and in-between

correct compartment during their first attempt) and the latency to detour as measures of RI. They found that 103 the detour accuracy for guails was lower (i.e., RI performance was impaired) for vertical- than for horizontal-104 bar barriers. This seems consistent with the findings of Regolin and colleagues (1994), although it should be 105 noted that Zucca et al. (2005) only found a significant effect for detour accuracy but not for the latency to de-106 tour, the measure of RI in the study of Regolin et al. (1994). For yellow-legged gulls, detour accuracy was not 107 influenced by barrier type, but detour latency was. Specifically, the latency to detour was longer (i.e., RI was 108 impaired) for horizontal- than vertical-bar barriers. Again, the authors attributed this to the species' ecologi-109 cal niche. Specifically, Zucca et al. (2005) argued that in the (original) ecological niche of young, yellow-legged 110 gulls, chicks are accustomed to consider the vertical ground vegetation of sand dunes as largely impenetra-111 ble. According to the authors, it might therefore be harder for (young) gulls to detour around horizontal-than 112 vertical barriers, as the latter would be perceived as less penetrable (note that Zucca et al. (2005) tested ju-113 venile gulls that could not fly vet). Lastly, canaries were unable to detour around the barrier, although they 114 made several attempts to fly over the barrier demonstrating that they were sufficiently motivated. Zucca et al. 115 (2005) therefore hypothesized that the detour task is not considered to be a real problem for the two-month 116 old canaries. The authors suggested that, after all, canaries are aerial birds, allowing them to tackle obstacle 117 problems by simply flying over them (but which was not possible in the detour task due to the dimensions of 118 the used apparatuses). 119

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In sum, the results of Regolin et al. (1994) and Zucca et al. (2005) indicate that the characteristics of the 'stop 121 signal' matter in the detour task, potentially shedding new light on RI in avian species. However, some con-122 cerns can be raised about certain features of the previous studies (which are summarized in Table 1). First, 123 the sample size was low (at least for some species), the studies (inconsistently) used within- and between-124 species designs, the number of trials per barrier type differed within and between species and the number 125 of sessions per barrier type fluctuated between species (e.g., yellow-legged gulls received three sessions per 126 barrier type spread over three days, while hybrid quails received one session per barrier type). The latter 127 two issues are also problematic from a conceptual point of view, as previous work indicates that learning 128 will influence RI (Verbruggen, Best, et al., 2014; Verbruggen and Logan, 2008a), including in the detour task 129 (Van Horik, Beardsworth, Laker, Whiteside, et al., 2020). Second, both studies used less-common variants of 130 the detour task, which complicates comparisons with the wider literature. Additionally, they used less com-131 mon, hard to standardize (social) rewards, which complicates between-species comparisons of RI behaviour. 132 Similarly, the large differences in developmental trajectories and the lack of adapting the test apparatus to 133 the morphological differences between species, also complicated the between-species comparisons. 134

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# A Partial Replication of Regolin et al. (1994) and Zucca et al. (2005)

To prove the significance of the previous study, our study investigated the role of stop-signal detection in avian RI by partially replicating the studies of Regolin et al. (1994) and Zucca et al. (2005). The importance of the original studies is indubitable, as they are one of the few studies that aimed to 'deconstruct' avian RI performance by focusing on the underlying cognitive processes (in this case, stop-signal detection). Additionally, Zucca et al. (2005) implemented a comparative approach to investigate whether differences in how the stop signal might be perceived by different species could contribute to variation in RI.

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In our partial replication, we made several changes to address commonly raised concerns in the detour literature (including the concerns raised in the previous section, see Table 1). First, we directly compared four species (white leghorn chickens, Japanese quails, herring gulls (*Larus argentatus*<sup>3</sup>) and domestic canaries), in a well-powered mixed design analysis with *Species* as between-species factor, and *Barrier* (vertical-bar vs.

line/rod gaps of 1.25 cm. Given these measurements, only a combination of 18 rods/lines and 18 in-between gaps is possible, which is the equivalent of a vertical- or horizontal-bar barrier with a 14% reward occlusion.

<sup>&</sup>lt;sup>3</sup>Given the high availability of the herring gull at the North Sea coast (Belgium), herring gulls were selected to represent the gull species in this comparative study. But see also footnote 1.

Source	Regolin (1994)		Zucca (2005)		Current Study
1. Methodology					
Species	White leghorn chicken	Hybrid quail	Yellow- legged gull	Canary	All four species
Design	Between	Between	Within	Within	Mixed
Total Sample	750 (250) <sup>1</sup>	90	5	26	240 (60/species)
Sample per Barrier Type	102 (34) <sup>1</sup>	18	5	26	60/species
Trials per Barrier Type	1	10	10	1 or 10 $^{2}$	3
Sessions per Barrier Type	1	1	3	1	1
2. Detour Task	Two compart.	Four compart.	Four compart.	Four compart.	Simple
3. Reward	Cagemates	Reflection	Reflection	Reflection	Food
4. Baseline covariate	No	No	No	No	Yes
5. DV's	Latency	Latency Accuracy	Latency Accuracy	Latency Accuracy	Latency Persistence
6. Enclosure					
Social density	3	1	5	5	$\pm$ 10
Fence	NA	Vertical	Bricks	Vertical	Mesh netting
7. Test Age	2 days	1 M	10-25 days	4-6 M	Species specific
8. Apparatus					
Test box: L x W	120 x 35	150 x 75	150 x 75	150 x 75	Scaled/species
Test box: H	60	40	40	40	Barrier H
Barrier-Entry Distance	27	27	27	27	Scaled/species
Barrier: L x H	10 x 20	23 x 26	23 x 26	23 x 26	Scaled/species
Barrier line: W	0.3	0.2	0.2	0.2	Scaled/species
Gap between barrier lines	1.2	1.25	1.25	1.25	Scaled/species

Table 1. Methodological features of the Regolin et al. (1994), the Zucca et al. (2005) and the present study.

Note. DV's = Dependent variables, 1 = Animals were reared and tested in groups of three. The means of each trio was used as individual data for the subsequent analysis; 2 = 6/26 canaries received 1 trial/barrier, 20/26 canaries received 10 trials/barrier. Measurements are in cm.

horizontal-bar barrier) as within-species factor. Hereafter, each species will be referred to by its common 147 name for clarity: chicken (excluding 'White Leghorn'), quail (excluding 'Japanese'), gull (excluding 'herring'), and 148 canary (excluding 'domestic'). All species were given an equal amount of trials and sessions per barrier type 149 (see below). Second, the perceptual characteristics of the barrier (i.e., vertical-bar vs. horizontal-bar barriers) 150 were manipulated in a simple detour barrier task (which is the most common variant of the detour problem; 151 (Kabadayi, Bobrowicz, et al., 2018), rather than a four- (Zucca, Antonelli, et al., 2005) or two-compartment 152 (Regolin, Vallortigara, et al., 1994) detour task). See Figure 1, for an overview of the designs. Third, the un-153 conditional reward was food instead of a social stimulus (as in Regolin, Vallortigara, et al., 1994 and Zucca, 154 Antonelli, et al., 2005). Food is a common reward in laboratory tests and has a high incentive value across 155 species and individuals. Furthermore, it's subjective value can be better standardised both within and be-156 tween species compared with social rewards. Fourth, non-cognitive, motivational states can influence detour 157 performance (Kabadayi, Krasheninnikova, et al., 2017; Van Horik, Langley, et al., 2018). Therefore, we col-158 lected for each individual a 'multi-baseline' measure of their general motivational state (which could be a 159 combination of, e.g., non-transparent obstacle neophobia, test box neophobia, food motivation, or motiva-160 tion to explore). This 'multi-baseline' measure was obtained with an opaque barrier during habituation (see 161

below). We included this as a covariate in our statistical models to increase the likelihood of detecting barrier

type effects within species conditional on/adjusted for the 'multi-baseline' measure of an individual's general 163 motivational state<sup>4</sup>. Fifth, our study considered two measures of interest, namely the latency to detour (Regolin, Vallortigara, et al., 1994; Zucca, Antonelli, et al., 2005) and the time spent in proximity to the barrier 165 (persistence). The last variable was not included in the original studies, but adds substantial information 166 about variation in (un)successful inhibition following the different barrier types. Note that this measure also 167 captures 'accuracy', as all birds that did not peck at the barrier (i.e., an accurate response) got a score of 0, 168 whereas all birds that pecked, got a score > 0. Sixth, for all species, (fledged) chicks were raised in groups of 169 approximately 10 individuals, as variations in detour performance had already been reported with fluctuating 170 social group sizes (Ashton et al., 2018). Furthermore, mesh netting was used for the enclosures of all species 171 (preventing variation in experiences with (im)prenetrable vertical- or horizontal-bar objects in the enclosures). 172 Seventh, detour performance of the different species was compared when they are on similar levels in their 173 developmental trajectory (see e.g., Kabadayi, Jacobs, et al. (2017), Kabadayi, Krasheninnikova, et al. (2017) 174 and Verbruggen, McLaren, et al. (2014) for the influence of cognitive maturation on RI), and again, with simi-175 lar experiences in the enclosure, keeping in mind the precocial-altricial spectrum (see below). Finally, the size 176 of the test apparatuses and barriers (including the width and the in-between line gaps) was scaled based on 177 the morphological characteristics of each species (see below). For example, Zucca et al. (2005) used the same 178 test box for all three species, and argued that this was appropriate because they tested species at different 179 ages. Nevertheless, one could still expect substantial differences in body size (e.g., a one-month old quail is 180 substantially larger than a two-month old canary). As (relative) distance towards the reward influences detour 181 performance (Kabadayi, Bobrowicz, et al., 2018), it was therefore important to control for this as well.



**Figure 1.** A display of the three detour task variants: the two-compartment detour task (left, Regolin, Vallortigara, et al., 1994) the four-compartment detour task (middle, Zucca, Antonelli, et al., 2005) and the simple detour barrier task (right, current study). Double line: barrier; full arrows: correct responses; dashed arrows: incorrect responses.

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## Predictions

First, we predicted better RI performance for ecologically valid compared with non-valid stop signals, as 184 should be reflected in shorter latencies to detour and less time spent persisting. As the (original) ecologi-185 cal niche of our species substantially differs (chicken and quails: penetrable vertical terrestrial vegetation; 186 gulls: impenetrable vertical vegetation of sand dunes; canaries: aerial environment), the ecological validity of 187 stop signals would be species-specific. Specifically, for chickens and quails, we expected better detour perfor-188 mance for horizontal- (HB) compared with vertical-bar barriers (VB; thus, detour performance HB > VB). We 189 expected the opposite pattern for gulls (i.e., detour performance HB < VB). Finally, based on the findings of 190 Zucca et al. (2005), we did not expect differences between vertical- and horizontal-bar barriers for canaries 191 (i.e., detour performance HB = VB). Overall, this pattern should result in a statistical interaction between Bar-192 rier and Species (Prediction 1). 193

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<sup>&</sup>lt;sup>4</sup>The original studies administered a variety of barrier types, including a transparent barrier, which was then also used to interpret performance with the other barrier types. Yet, we opted to exclude transparent barriers from our design for three reasons. First, the comparison of transparent and barred barriers tests another hypothesis, namely the role of reward occlusion, rather than the ecological validity of the barrier type (which can only be tested by comparing vertical- and horizontal-bar barriers). Second, performance in the detour task can be influenced by differential individual experiences with transparent obstacles (Kabadayi, Krasheninnikova, et al., 2017; Van Horik, Langley, et al., 2018). Last, omitting the transparent barrier reduces the contact of wild gulls to humans, which is important as gulls are wild animals and will be raised as such, prior to their post-test release back to nature.

Second, as each session consisted of three trials (of the same barrier type), we also looked at how detour
 performance improved within each session. Based on previous studies, we predicted that detour performance should improve across trials within a session (**Prediction 2**).

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Furthermore, we explored if the learning effect (i.e., improved detour performance across trials) interacts 199 with the ecological validity of the stop signals. There are two possible patterns that would result in a three-way 200 interaction between Species, Barrier (horizontal- vs. vertical-bar barriers), and Trial (1-3) (Explorative Predic-201 tion 3). First, detour performance might be better for ecologically valid compared with non-valid stop signals 202 at the beginning, but this pattern might diminish over time as individuals learned to stop (i.e., the differences 203 between barrier types would decrease). Second, detour performance might be poor at the beginning for 204 both barrier types, but learning to stop might be easier for ecologically valid signals compared with non-valid 205 stop signals (i.e., the differences between barrier types would increase). Both patterns would be theoretically 206 meaningful, but we did not have *a-priori* predictions about the direction of the three-way interaction. 207

# 208

# Methods

We report how we determined our sample size, all data exclusions, all inclusion/exclusion criteria, whether inclusion/exclusion criteria were established prior to data analysis, all manipulations, and all measures in the study.

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# **Subjects and Housing**

Quails, gulls and chickens were raised and tested at the avian research facilities of Ghent University (Lab number LA1400452), located at the Wildlife Rescue Center (WRC) in Ostend (Belgium). The canaries were raised and tested at the avian research facilities of the University of Antwerp (Lab number LA1100161) in Wilrijk (Belgium).

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# Sample size

We originally registered to test 60 individuals per species. A-priori power sensitivity analyses done in 219 G\*Power (Faul et al., 2009) indicated that this was sufficient to detect small effects; it was also the largest 220 number that was practically feasible<sup>5</sup>. For the sensitivity analysis, we used a mixed ANOVA model with one 221 between-subjects factor (4 levels; corresponding to our Species factor) and two within-subjects factor (one 222 with 2 levels - Barrier - and one with 3 levels - Trial). This indicated that our sample size of 60 animals per 223 species (240 in total) was sufficient to detect a Species x Barrier interaction effect (Prediction 1) with a small 224 effect size (Cohen's f effect size of 0.12; Cohen, 1988; Power = .80; cor. among RM = 0.5; we used an alpha 225 of .025 to correct for the fact that we had two dependent variables measuring (slightly) different aspects of 226 detour performance). Second, our sample size was sufficient to detect a small effect of Trial (Prediction 2; Co-227 hen's f effect size of 0.09; Cohen, 1988). Third, our sample size was sufficient to explore a small effect (Cohen's 228 f effect size of 0.09; Cohen, 1988) for the Species x Barrier x Trial interaction effect (Explorative Prediction 229 3). Due to higher than expected post-hatch mortality, the quail sample size was slightly reduced to 58. This 230 reduction did not result in meaningful differences in effect size for our three predictions <sup>6</sup>. 231

Our sensitivity analyses were based on mixed ANOVAs (fixed-effects models with between- and withinspecies factors). However, as discussed below, we analysed our data with (G)LMMs, which are currently not covered by G\*Power or most other power-estimation tools. These mixed-effect models are more flexible in assigning variance as they allow for the specification of both fixed and random effects. However, by accounting for unexplained variance (see below), our proposed mixed effect models are more powerful than the fixed-effect model ANOVAs used in our sensitivity analyses (and than the models used in the studies of

<sup>&</sup>lt;sup>5</sup>Farrar et al. (2020) mention in their paper on replications in comparative cognition that power analyses are not the golden standard in this research domain, and 'in many cases comparative cognition researchers could be better off performing design or sensitivity analyses based on their resource constraints.'

<sup>&</sup>lt;sup>6</sup>This adjustment was made after data collection but prior to any data analysis. These changes are still considered part of Stage 1 IPA and were pre-approved.

Regolin, Vallortigara, et al., 1994, and Zucca, Antonelli, et al., 2005). Thus, the sensitivity analyses discussed
 here are a conservative estimate.

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## White leghorn chickens and Japanese quails

Chicken and quail eggs were obtained from local breeders in Belgium. At the WRC, the eggs were incubated 241 in Brinsea Ova-Easy incubators (temperature = 37.5°C; humidity = 45% for first 15 [quail] or 17 [chicken] days, 242 after which humidity = 70% till hatching). Once hatched, chicks received a unique colour ring combination 243 prior to being housed in groups of  $\pm$  10 chicks<sup>7</sup> per indoor enclosure (size = 1m x 1m x 2m; L x W x H; 244 ambient temperature =  $15-25^{\circ}$ C; humidity = 40%-80%; photoperiod = 14:10 L:D; type of wire fencing = mesh 245 netting). Birds were ad libitum provided with a chicken meal mixture (Aveve Chicken Start Mash) and water. 246 Shelter, additional heating panels (30 x 30 cm; till Day 7), and pecking objects (pine cones) were available. The 247 (precocial) chickens and quails were tested at ±3 weeks (see below for justification of species-specific test age). 248 After testing, the individuals were euthanized by certified staff. 249

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#### **Herring Gulls**

Gull eggs were collected by the 'Agentschap voor Natuur en Bos' and the gull patrol team in Ostend (https: 251 //www.oostende.be/meeuwen) who are authorized to remove gull eggs along the Belgium coast for various 252 reasons. The eggs were collected prior to pipping, and were safely transported to the WRC. At the WRC, the 253 eggs were further incubated in Brinsea Ova-Easy incubators (temperature =  $37.5^{\circ}$ C; humidity = 45%) and 254 checked twice per day for signs of pipping. When gull embryos reached the pipping stage, they were placed 255 in a hatchery (temperature = 37.2°C; humidity = 50%). Once hatched, the semi-precocial gull chicks received 256 a unique colour ring combination prior to being placed in boxes with netting bottoms (size = 1.20m x 0.60m 257 x 0.60m; L x W x H) within heated rooms (ambient temperature = 15-25°C; humidity = 40%-80%; typical pho-258 toperiod = the natural photoperiod at the latitude of Belgium; type of netting = grid) for 5 days (and till their 259 body mass exceeded 60 grams). During this period, the gulls were hand-fed small pieces of fish and dog 260 pellets (soaked in water), supplemented with Akwavit (Kiezebrink Focus on Food, The Netherlands). We also 261 provided one heating panel per box. After this initial indoor period in the boxes, the gull chicks were housed 262 in groups of  $\pm$  10 individuals<sup>7</sup> per outdoor enclosure (size = 5 m x 1,95 m x 2,65; L x W x H), type of wire fencing 263 = mesh netting), with an extra heating panel for the first couple of days (note that the exact number of days 264 depended on the weather conditions). Food (a mixture of 75% dog food soaked in water and 25% defrosted 265 fish, supplemented with Akwavit) was provided four times per day (the default policy at the WRC); water was 266 provided *ad libitum*. The (semi-precocial) gulls were tested when they were approximately 3 weeks old (see 267 below). After testing, gulls were moved to large flight cages to dehabituate them from human handling (and 268 hence improve their survival rates). They were released into the wild when they were approximately 8-10 269 weeks old. 270

## **Domestic canaries**

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# 272 Domestic canaries (of the Fife Fancy type) were obtained from long-term, breeding populations at the De-

partment of Biology ('Behavioural Ecology and Ecophysiology' research group) of Antwerp University. Canaries
 are altricial species, and nestlings are thus highly dependent of their parents for food. Therefore, chicks were

only separated from their parents at the end of the nestling period (i.e., when they were  $\pm 25$  days old) (Garcia-

<sup>&</sup>lt;sup>276</sup> Co et al. 2024)<sup>8</sup>. At this point, the canaries were moved in groups of  $\pm$  10 individuals<sup>7</sup> to indoor aviaries of

<sup>&</sup>lt;sup>7</sup>This adjustment was made after data collection but prior to data analysis. These changes are still considered part of Stage 1 IPA and were pre-approved. Additionally, an analysis incorporating group size as an extra random variable in the statistical models is provided in the supplementary materials (Dewulf, Garcia-Co, et al., 2023).

<sup>&</sup>lt;sup>8</sup>During the nestling period, birds were kept in cages with vertical wires (the default policy at the University of Antwerp). However, we believe that this does not affect the interpretation of our results. First the birds did not interact with the cage during the nestling period, so they could not learn whether the bars or strips were impenetrable or not. Second, Zucca et al. (2005) argued that detour performance is improved when the perceptual characteristics of the stop signal match the ecological niche (rather than the prior experience) of the

Antwerp University (size : 1m x 2m x 2m; L x W x H; ambient temperature = 15-25°C; humidity = 40% - 80%, 277 photoperiod = the natural photoperiod at the latitude of Belgium; type of wire fencing = mesh netting). The canaries were marked with a permanent marker for individual recognition at hatching, and ringed with a 279 closed metal ring when their body mass exceeded the predetermined threshold of 7 g. Upon arrival at the 280 indoor aviaries, canaries received a unique number-color ring combination (the default policy at the Univer-281 sity of Antwerp). In the aviaries, canaries were provided with canary seed mixture and egg food (van Camp, 282 Belgium), water, shell grit, and cuttlefish bone ad libitum. They were tested at 7 weeks (approximately three 283 weeks after fledging; see below). After testing, canaries returned to their local breeding population in the 284 University of Antwerp. 285

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## A Comparative Testing Age

Our previous work (Troisi et al., 2024) indicates that 3 weeks is an ideal testing age for large gulls (incl. 287 herring gulls) in detour tasks and other related cognitive tests. Gulls are semi-precocial, but only require hand-288 feeding for the first couple of days (and most start eating independently after two/three days). Furthermore, 289 the gull chicks can already move around (and explore their environment) from Day 1. Chickens and quails are 290 precocial, which implies that they can feed independently and explore their environment from Day 1. Given 291 the overall similarities, we therefore tested chickens, quails and gulls when they were approximately 3 weeks 202 old (i.e., habituation happened  $\approx$  on day 16-18; testing happened on  $\approx$  day 19-20). By contrast, canaries only 293 become independent when they are approximately 25 days old (see previous subsection). At this point they 294 were moved to larger enclosures and housed in groups. To ensure that the altricial canaries had a similar 295 (15-day) experience with their enclosure and their cagemates as the (semi-)precocial species, habituation and 296 testing of canaries was delayed with 25 days (i.e., habituation happened  $\approx$  on day 41-43; testing on  $\approx$  day 297 44-45) 298

## Apparatus

For each species, the test apparatus consisted of a two door start box, a test box, a barrier, and a feed-300 ing bowl. Performance of the birds was monitored using a camera placed centrally at the top of the testing 301 arena (Sony Action Cam HDR-AS50). In the test box, a vertical- or horizontal-bar barrier blocked the direct 302 path to the unconditional reward (i.e., the food in the bowl) that was immediately placed behind the barrier. 303 The species-specific unconditional food reward (chickens and quails: chicken meal, gull: dog pellets and fish, 304 canaries: canary seed mixture and egg food) consisted of clearly visible food, placed in a coloured bowl. For 305 chickens and quails, these were coloured green and yellow (brand: Junai, The Netherlands); for gulls and 306 canaries, these were coloured orange-brown (brand: Elho, Belgium)<sup>9</sup>. To avoid satiation after the first trial 307 on test trials (see next section), the pile of food was largely covered with a transparent perspex cover, with 308 only a small bit of accessible food placed on top of the cover. The vertical- and horizontal-bar barriers were 309 made of transparent perspex on which 18 vertical and horizontal lines, respectively, were painted per species 310 (see below). To prevent canaries from flying over the barrier (as as alternative way of avoiding the barrier), 311 floor-to-ceiling barriers were used for all species. 312

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The size of the test apparatus was adjusted per species. In a recent study from our lab, we tested gulls in a detour task (akin to the task proposed here, but with transparent and non-transparent barriers). In this study, the starting box was  $35 \times 20 \times 26$  cm (L x W x H), and the test box was  $145 \times 88 \times 132$  cm (L x W x H). The barrier was  $40 \times 40$  cm (L x H), and was placed 50 cm from the start box entrance (with approximately 24 cm between the edges of the barrier and the sides of the test box). In the present study we used the

species. Notably, although the canaries in Zucca et al. (2005) were reared in cages with vertical bars from two months until tested at four to six months, performance did not differ between barrier types.

<sup>&</sup>lt;sup>9</sup>Potential variations in colour perceptions and sensitivities across species are mitigated by the developed preference for their respective food bowls prior to the start of the experiment through repeated exposure (i.e., 10-day habituation to the food bowl in the enclosure) and learning (pairing of the coloured bowl with food during these 10 days).

same set-up for the gulls, and re-scaled all values based on tarsus length at testing age (see Table 2 for the 319 values for each species). For the chickens, quails and gulls we used the growth curves (Figure 2) for tarsus 320 length reported in previous studies (Dudusola and Bashiru, 2020; Troisi et al., 2024; Yeasmin and Howlider, 321 2013). For canaries, no such growth curves were available. However, in a recent study, Garcia-Co et al. (2024) 322 measured tarsus length at day 25. Given that morphological traits (incl. tarsus) seem to plateau at a similar 323 moment in the lifespan of a canary, we used this tarsus measure at day 25 as our measure for the tarsus 324 length at testing age. In addition, the black painted barrier lines (18 in total per species) occluded the food 325 reward by approximately 14% (Zucca, Antonelli, et al., 2005). As a consequence, the width of the barrier 326 lines (and of the in-between gaps) as described in the study Zucca and colleagues (2005) was adjusted to the 327

re-scaled barrier size dimensions per species (see Table 2).

**Table 2.** The upper table shows the tarsus length and species-specific re-scaled test apparatuses based on the herring gull detour set-up of Troisi et al. (2024). The lower table shows the re-scaled width of the barrier lines (and of the in-between gaps) for each species based on the detour set-up of Zucca et al. (2005)

	White leghorn	Japanese	Herring	Fife Fancy
	chicken	quail	gull	Canary
Source	Yeasmin (2013)	Dudusola (2020)	Troisi (2024)	Garcia-Co (2024)
Test age (days)	19	19	19	44
Sample size	130	2591	42	69
Mean Tarsus (mm)	30.71	25.32	55.56	17.81
Ratio	1.81	2.19	1	3.12
Apparatus (rescaled)				
Start box: L x W	19.35 x 11.05	15.95 x 9.11	35 x 20	11.22 x 6.41
Test box: L x W	80.15 x 48.64	66.08 x 40.10	145 x 88	46.48 x 28.21
Barrier-Entry Distance	27.64	22.79	50	16.03
Barrier: L x H	22.11 x 22.11	18.23 x 18.23	40 x 40	12.82 x 12.82
	White leghorn	Japanese	Herring	Fife Fancy
	chicken	quail	gull	Canary
Barrier line: W	0.17	0.14	0.31	0.10
Gap between barrier lines	1.06	0.87	1.91	0.61

Note. The estimated tarsus length at testing age for (semi-)precocial species, is derived from a linear equation using the two nearest measuring points for a mixed-sex sample (except for white leghorn chickens, where an additional average was calculated over pullets and cockerels). For re-scaling the test apparatuses, the unrounded factor per species was utilized. Unless specified otherwise, measurements are in cm.



**Figure 2.** Full lines: reported tarsus growth during early life for white leghorn chickens (Yeasmin and Howlider, 2013), Japanese quails (Dudusola and Bashiru, 2020) and herring gulls (Troisi et al., 2024). Dashed lines: hypothetical tarsus length for canaries based the assumption that tarsus length does not change (much) after fledging (Garcia-Co et al., 2024)

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## Procedure

Prior to the start of the experiment, birds were habituated for 10 days in their enclosure to feed from a 330 coloured food bowl, which was identical to the bowl used during both habituation and testing in the test box. 331 For all species, the feeder was placed on the ground, to simulate ground feeding during the test. When they 332 reached the appropriate age (see above), animals were tested for five consecutive days (i.e., three habituation 333 and two testing days). Food in the enclosures was provided ad libitum, but the evening before an individual's 334 habituation or testing day, the feeders were removed from the enclosures at 6PM (after the last feeding time). 335 This created a non-feeding period during the night (which is normal and also happens in non-experimental 336 conditions), followed by (shortly) delayed feeding in the morning to prevent birds from overindulging prior 337 to habituation or testing. This is in line with other studies using the same species (chicken: e.g., Bollweg and 338 Sparber, 1998; quail: e.g., Ueno and Suzuki, 2014 and unpublished data from our lab; gulls: e.g., Dewulf, 339 Knoch, et al., 2025; canaries: e.g., Müller et al., 2008). After all individuals of one enclosure completed the 340 habituation or testing trials for the day, food was again provided *ad libitum*. 341 342

On the three habituation days (08:00 AM - 10:30 AM), each bird received 1 trial per day where it could 343 freely explore the test box and feed from a centrally placed coloured food bowl. During the second and 344 third habituation day, an opaque barrier was placed just behind the coloured food bowl. This allowed us to 345 obtain a 'multi-baseline' measure of an individual's general motivational state (which could be a combination 346 of e.g., non-transparent obstacle neophobia, test box neophobia, food motivation, motivation to explore; see 347 below). The current habituation set-up (i.e., the food bowl in front of the barrier) was designed in such a 348 way that acquiring a motor routine during habituation was unnecessary and could not confound subsequent 349 detour performance with the barred barriers (Van Horik, Beardsworth, Laker, Whiteside, et al., 2020). 350 Each bird participated in one session per day on the two testing days (10:30AM - 02:30 PM), each bird 351

<sup>351</sup> Latri bird participated in one session per day on the two testing days (10.50AM - 02.50 PM), each bird
 <sup>352</sup> performed one session. Each session consisted ing of 3 trials with one barrier type. The order of barrier type
 <sup>353</sup> (i.e., horizontal-bar or vertical-bar barrier) was pseudo-randomized within and between species, across the
 <sup>354</sup> two testing days.

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Due to the natural breeding season of the wild gull and the canary, birds hatched non-simultaneously. In order to guarantee an appropriate test age (see above), we grouped individuals of a similar age per enclosure; and then habituated or tested birds per enclosure (by taking into account the average age of the enclosure). Although there was no fixed breeding season for quails and chickens, incubation happened in 'batches' (due to reduced egg production/supply). As a result, an identical grouping procedure within these species was applied.

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At the beginning of each trial, each bird was gently placed in the dark two-door start box. The trial started 363 when the researcher opened the first non-transparent cardboard door of the start box. This permited the 364 bird to see the test arena but not access it. After 15 seconds, the second, transparent door of the start box 365 was opened and the bird could enter the test box. If the bird did not exit the start box within 30 seconds, it 366 was gently pushed forward (by sliding the back of the starting box forward; Troisi et al., 2024). The habituation 367 trials ended when the individual ate from the food bowl for 30 seconds or when the maximum trial time had 368 been reached (i.e., 5 min 15 seconds). The test trials ended immediately when the individual ate from the 369 food bowl (to avoid food satiation on subsequent trials) or when the maximum trial time had been reached 370 (i.e., 2 min 15 seconds). Maximum trial times during habituation were longer than during testing, as the main 371 goal of the habituation was to familiarize each bird with the test material (and obtain a 'multi-baseline' mea-372 sure of an individual's general motivational state). The maximum duration of a test trial was 2 minutes (after 373 an additional 15 seconds inside the start box with the second, transparent door), which is in line with other 374 studies (e.g., Kabadayi, Krasheninnikova, et al. 2017 and Vernouillet at al. 2016). Two minutes should be sufficient, especially because our barriers were not entirely transparent (hence, partially occluded the food 376 reward), making it easier to execute a detour response (Kabadayi, Bobrowicz, et al., 2018). 377

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Gulls were tested during the second half of June in 2023 and 2024 (restricted to breeding season), quails in November 2023 (autumn), chickens in February 2024 (late winter), and canaries in May 2024 (late spring, again, restricted to breeding season).

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# **Data Processing and Analysis**

# **Video Recording and Analysis**

The videos of the second and third habituation trial and the three test trials per test session were coded 384 using the free, open-source 'Behavioural Observation Research Interactive' Software (BORIS, v.7.13.6) (Friard 385 and Gamba, 2016). We coded five (types of) events (see Table 3 and Figure 3): latency to leave the start box 386 (for habituation trials 2 and 3, as well as the six test trials), persisting (test trials only), moment of detouring 387 the barrier (test trials only), interacting with the food bowl (for habituation trials 2 and 3, as well as the six test 388 trials) and leaving the species-specific 'test box zone of interest' (test trials only). All videos were coded by the 389 first author. A second person, blind to the hypotheses coded 10% of the videos per species. An average Co-390 hen's Kappa (McHugh, 2012) was calculated for these videos to provide a measure of inter-rater, cross-species 391 reliability. We had registered that, in the case no perfect inter-rater, cross species agreement (0.81 < Cohen's 392 Kappa < 1) had been reached, discrepancies in inter-rater reliability would be investigated by calculating the 393 average Cohen's kappa (McHugh, 2012) for each species, separately. By doing so, a species-specific or overall 394 low Cohen's Kappa would reveal whether the videos have to be recoded for one or all four species. However 395 this was not needed, as the average cross-species Cohen's kappa value indicated a strong level of inter-rater, 396 cross species agreement (k = 0.927; McHugh, 2012). 397

To compare detour performance between species, we extracted our two response variables from the behavioural events coded in BORIS. First, the latency to detour (in seconds) was determined as the time between

leaving the start box and the moment the individuals detoured the barrier. A maximum trial duration of 135 401 seconds for detour latency was assigned to the 20 trials (1.40% of the data set) in which birds did not failed 402 to detour but entered the species-specific 'barrier zone of interest'. Second, the time spent persisting (in 403 seconds) was calculated as the cumulative time that the individual spent in the species-specific 'barrier zone 404 of interest' (size = Barrier L x 25% of the Barrier-Entry Distance; L x W, see Table 2 for the species-specific 405 dimensions). A minimum trial duration of 0 seconds for persisting was assigned to the 483 trials (33.82% 406 of the data set) in which birds detoured without entering did not enter the species-specific 'barrier zone of 407 interest' first. Third, a 'multi-baseline' measure of an individual's general motivational state (in seconds) was 408 calculated, by averaging the time between leaving the start box and touching the food (bowl) placed in front of 409 the opaque barrier on habituation trial 2 and 3. Note that if a bird was unsuccessful on trial 2, a non-averaged 410 'multi-baseline' score was calculated based on habituation trial 3 only. 411



**Figure 3.** Visualisation of the species-specific dimensions (left) and the behaviours (right) that were coded in BORIS. Double line: Barrier; hatched area: species-specific 'barrier zone of interest'; dashed lines: (fictitious) lines that needed to be crossed by the bird, see Table 3.

Behaviour	Description
Leaving start box	<ul> <li>When the bird voluntarily leaves the start box: when both feet of the bird are visibly inside the test box, or (when the feet are not visible) when the front body half of the bird is inside the test box.</li> <li>When the bird needs to be pushed: When the bird's entire body is inside the test box <sup>1</sup>.</li> </ul>
Persisting	At least the bird's whole head crosses the (fictitious) lines of the rectangular-shaped, species-specific 'barrier zone of interest' <sup>2</sup> .
Detour	At least the bird's whole head crosses the (fictitious) straight line from the barrier to the side of the test box (with a modifier whether they detour on the right or left side)
Interaction with food bowl	Bird touches the food or food bowl with its beak.
Leaving the 'test box zone of interest'	At least the bird's whole head crosses the (fictitious) straight line at $\approx$ 2/3 of the test box length <sup>2</sup> .

Table 3. The description of the behaviours that were coded in BORIS.

Note. <sup>1</sup> = When a bird needed to be gently pushed, it was most likely that the individual was lying down in the start box. As a result, a gentle push put the bird forward and resulted in the bird ending up in a standing position in the test box. As a result, we coded whether these individuals had left the start box when the bird's entire body (vs. front half body) was inside the test box. <sup>2</sup> = The fictitious lines that marked a zone of interest were defined by two wooden sticks attached to each side of the test box.

# **Data Exclusion Criteria**

Individuals that failed to visit the food bowl at the third habituation day, were excluded from subsequent
test trials (pre-test criterion). This exclusion criterion guaranteed a similar within- and between-species proficiency with the basic task demands (e.g., the perceptual, motoric and motivational requirements to retrieve
a food reward; for a similar pre-test exclusion criterion see, MacLean et al., 2014).

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Birds that did not detour around the barrier nor entered the species-specific 'barrier zone of interest' in 418 a test trial, were excluded from subsequent test trials (and data of that individual was excluded from all 419 statistical analyses). This mid-test exclusion criterion 1 was applied for two reasons. First, birds that did not 420 obtain a measure for one of the two dependent variables within 2 minutes were likely to be unmotivated or 421 be in distress. Furthermore, observations from similar RI test paradigms in our lab demonstrate that such 422 individuals are unlikely to eat at all with a prolonged test time or on subsequent test trials (within the same 423 day).<sup>10</sup> In addition, removing birds from subsequent trials (rather than assigning a maximum trial limit for 424 both dependent variables) reduced the risk of data skewing. 425

Individuals that left the species-specific 'test box zone of interest' (size = 2 times the Barrier-Entry Distance, 426 see Table 2 for the species-specific dimensions) without touching the food (bowl) were also excluded from 427 further testing and all analyses. This mid-test exclusion criterion 2 assured that we avoid confusing general ex-428 ploration behaviour (without initial interest in the food) with successful detour performance (which assumes 429 interest in the food). Thus, by excluding birds with differential trial experiences (due to e.g., demotivation, dis-430 tress, distraction or exploration; for a similar mid-test exclusion criterion see, Van Horik, Beardsworth, Laker, 431 Langley, et al., 2019), we aimed to ensure that each barrier orientation is standardized within- and between 432 species. We registered that we would test all individuals of each species in a single 'season', as we incubated 433 per season 20% more eggs than the number of individuals required for the testing; we expected that this 434 surplus would allow us to replace all excluded individuals. For an overview of the birds excluded per criterion, 435 see Table 4. However, due to the fearful and stress responses of gulls during testing (we come back to this in 436 the general discussion), the exclusion rate was higher than expected so we had to include a second breeding 437 season. 438

Species	White leghorn	Japanese	Herring	Fife fancy	Total
	chicken	quail	gull	Canary	
Initial sample	90	85	147	170	492
Excluded for:					
Pre-test criterion	10 (11%)	2 (2%)	35 (24%)	1 (1%)	48 (10%)
Mid-test criterion 1	2 (2%)	4 (5%)	39 (26%)	14 (8%)	59 (12%)
Mid-test criterion 2	7 (8%)	21 (25%)	0 (0%)	31 (18%)	59 (12%)
Technical issues/Sick birds	2 (2%)	0 (0%)	0 (0%)	9 (5%)	11 (2%)
Remaining sample	69 (77%)	58 (68%)	73 (50%)	115 (68%)	315 (64%)

**Table 4.** Number (and %) of birds excluded per criterion.

Note. All raised birds were subjected to habituation and (part of) testing. As can be seen, the total number of birds tested was higher than registered for all species (apart from the quails). This was due to the fact that these individuals were <del>also</del> reused for other studies, with different sample size requirements. Reusing individuals in other behavioral studies is possible when they share similar prior experiences (Van Horik, Langley, et al., 2018), and facilitates future analyses, such as exploring correlations between different tasks and making comparisons across studies. The first 60 individuals (58 for quails) that did not fail any exclusion criteria were selected for this study, ensuring a balanced design and minimizing group variation.

<sup>&</sup>lt;sup>10</sup>In a continuous RI task with a sample size of 80 herring gulls, birds that failed on the first trial, were likely to fail again on the second trial of the same test day (Dewulf, Knoch, et al., 2025)

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# **Statistical Analysis**

Statistical analyses were performed using R. v. 4.2.2 (R Core Team, 2021). Models were fitted by means of 440 the Ime4 package (D Bates, Mächler, et al., 2015) and parameter estimation and p-values for the generated 441 models were provided by means of the ImerTest package (Kuznetsova et al., 2017) via the Satterthwaite's 442 degrees of freedom method (linear mixed model, LMM) or via carData (Fox, Weisberg, and Price, 2022) and car 443 (Fox and Weisberg, 2019) packages, which are suited for both linear mixed models (LMM) with temporal corre-444 lation structures and generalized linear mixed models (GLMM). For the (G)LMM, we used partial eta-squared 445  $(\eta_{e}^{2})$  as effect sizes for the relevant statistical models (linear mixed model, LMM) and they were calculated by 446 means of the *effectsize* (Ben-Shachar et al., 2020) package. 447 118

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# **Registered model**

We registered that we would perform a (G)LMM with Type III sum of squares on the latency to detour and 450 the cumulative time spent in the species-specific 'barrier zone of interest' (persisting). These registered Both 451 models (model specification 1) included the between-species factor: Species (i.e., chickens, quails, gulls and ca-452 naries) and both within-species factors: Barrier (i.e., vertical- and horizontal-bar) and Trial (i.e., 1-3), and their 453 interactions. In addition, they included we add two extra explanatory variables to the model: a 'multi-baseline' 454 measure of an individual's general motivational state (and its interaction with Species, as we mean-centered 455 this *'multi-baseline'* measure within *Species*, see Chen et al., 2014 for an example of within-group centering): 456 and Barrier Order (with two levels: the individual received the horizontal-bar barrier on the first test day 1 457 and the vertical-bar barrier on the second test day; or vice versa), as species might demonstrate superior 458 performance with the last encountered barrier, irrespective of its type and ecological validity. Bird identity 459 and enclosure (social group) were included as a random intercept in the models, with bird identity nested in 460 enclosures. In addition, we included by-individual (nested in enclosures) random slopes that varied for the 461 levels of Species (corresponding with species-specific intercepts). The registered model is presented in model 462 specification 1. 463

Log(Outcome, s)~Species×(Barrier×Trial+Baseline)+BarrierOrder+(Species|Id:Enclosure)) (1)

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We registered that we would generate plots by means of the package *performance* (Lüdecke et al., 2021) to 466 inspect for violations of the model assumptions: 1) heteroscedasticity (plotting the square root of the residu-467 als (y-axis) and fitted values (x-axis)), 2) non-normality of residuals (plotting the sample quantiles (y-axis) on 468 the standard normal distribution quantiles) and 3) outliers (plotting standard residuals (y-axis) and leverage). 469 Additionally, the multicollinearity between fixed main factors (via the variance inflation factor, VIF) and the au-470 tocorrelation between residuals (via a Durbin-Watson-Test) would be calculated via functions provided by the 471 performance (Lüdecke et al., 2021) package. Potential violations of model assumptions would be addressed 172 by transforming the (in)dependent variables (i.e., via log-transformation) or by changing the error distribution 473 (family) or the link function of the model (switching a default LMM that will be fitted to a GLMM). Fixed main 474 effects with a VIF of >5 were planned to be removed and logical outliers (i.e., recording/entry errors) would 475 be inspected and corrected (if possible). In the case that the outlier could not be corrected, all data of that 476 individual was planned to be excluded from all statistical analyses. 477 478

# **Applied model**

Following the registered inspections and analyses, changes were made to the registered models to address
 model complexity, violations of certain assumptions, and issues with model convergence.

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For detour latency, the registered model (model specification 1) was simplified by removing the random 483 slope for Species in order to reduce the model's complexity. This decision was based on the presence of per-484 fect or near-perfect correlations among random effects, indicating redundancy and boundary singularity. In 485 order to address violations of the model assumptions (i.e., heteroscedasticity and non-normality of residuals), 486 the dependent variable was log-transformed (see supplementary Figure 1, Dewulf, Garcia-Co, et al., 2023). To 487 address autocorrelation in the residuals <sup>11</sup>, the model was further extended with a temporal correlation struc-188 ture using the *nlme* package (Pinheiro and DM Bates, 2000). This temporal correlation structure accounts for 489 the correlation in residuals from repeated measurements across Time (i.e., 1-6 trials; for each bird, nested 490 within enclosures). Specifically, each bird participated in two sessions, with one session per barrier type and 491 three trials per session, resulting in six interdependent trials. The autocorrelation parameter ( $\phi$ ), estimated 492 by the model at lag 1, was 0.319. Explicitly modeling this autocorrelation properly accounts for the residuals' 493 temporal dependencies (see supplementary Figure 2, Dewulf, Garcia-Co, et al., 2023), leading to an improved 494 model fit (AIC = 4063,716 with the correlation structure vs. AIC 4275,634 without) and more accurate parame-495 ter estimates. Adding this correlation structure was required as adjusting the error distribution (e.g., gamma 496 or inverse Gaussian) did not resolve the autocorrelation issue, as the models with adjusted error distributions 497 encountered convergence problems. VIF scores were all below 5, and no logical outliers were detected, so we 498 did not have to remove any outliers. The applied model for detour latency is presented in model specification 499 2, for an overview of the evolution of the model structure, see supplementary materials (Dewulf, Garcia-Co, 500 et al., 2023). 501

Log(Detour Latency, s)~Species×(Barrier×Trial+Baseline)+BarrierOrder+(1|Id:Enclosure)+corAR1(Time|Id:Enclosure) (2)

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For persisting, the registered model (model specification 1) was simplified by removing the random slope 503 for Species, for the same reasons as in the statistical model for detour latency. The simplified model demon-504 strated violations of model assumptions (i.e., heteroscedasticity and non-normality of residuals; see supple-505 mentary Figure 3 Dewulf, Garcia-Co, et al., 2023), which could not be addressed by log-transforming the de-506 pendent variable due to the presence of zeros in the data. To meet model assumptions, various models with 507 different error distributions were explored, including Poisson, negative binomial (ZI) and zero-inflated nega-508 tive binomial (ZINB), with the selection guided by the data characteristics. The dependent variable, persisting, 509 was therefore also converted to integer counts by scaling the original data to frames (30 frames per second), 510 which was necessary to meet the model's requirements while preserving the precision of short latencies that 511 would otherwise be rounded to zero. The selected models are designed to account for overdispersion and 512 excess zeros, reducing the need for explicit tests of heteroscedasticity and non-normality of residuals. How-513 ever, additional diagnostics were conducted using the DHARMa package (Hartig, 2022), assessing: 1) residual 514 uniformity (Kolmogorov-Smirnov test), 2) over/under-dispersion, 3) outliers, 4) zero-inflation, and 5) autocor-515 relation (via residual plots <sup>11</sup>). Ultimately, the ZINB model, implemented via the glmmTMB package <sup>12</sup> satisfied 516 the final model assumptions (see supplementary Figure 4 Dewulf, Garcia-Co, et al., 2023). The ZINB model in-517 cluded a negative binomial component to capture variability in persisting and a zero-inflated part to account 518

<sup>&</sup>lt;sup>11</sup>The Stage 1 registered Durbin-Watson statistic (using *performance* Lüdecke et al., 2021 package) and the ACF plot (using the base R package) are not suitable for (G)LMM (due to the random effect structure) an issue identified during the data analysis. Consequently, we opted for alternative methods to assess autocorrelation in the residuals. Specifically, we employed ACF plots (for LMM; using the package *nlme*, Pinheiro and DM Bates, 2000) or simulation-based residual plots (GLMM; using the package *DHARMa*, Hartig, 2022) for inspecting autocorrelation between residuals.

<sup>&</sup>lt;sup>12</sup> The *glmmTMB* (Brooks et al., 2017) package depends on both the *Matrix* (D Bates, Maechler, et al., 2023) and *TMB* (Thygesen et al., 2017) packages. To fit a (ZI)NB model with random effects, it is essential to reinstall the *Matrix* (D Bates, Maechler, et al., 2023) package from source (version 1.6.2) followed by reinstalling *TMB* (Thygesen et al., 2017) and *glmmTMB* (Brooks et al., 2017) from source.

for the excess of zeros in persisting. The best-fitting zero-inflation component was determined using AIC comparisons across models with different combinations of main and interaction effects in the zero-inflated model's formula, which resulted in the inclusion of *Barrier, Baseline, Species, Trial* and the interaction between *Species* and *Trial* as zero-inflated effects. VIF scores were all below 5, and no logical outliers were detected. The applied model for persisting is presented in model specification 3, see supplementary materials (Dewulf, Garcia-Co, et al., 2023).

NB2(Persisting,frames)~Species×(Barrier×Trial+Baseline)+BarrierOrder+(1|Id:Enclosure)+zi~Barrier+Baseline+Species+Trial+Species:Trial)
(3)

525

We had preregistered that, in case of (a) significant Barrier x Species interaction effect(s) (Prediction 1), 526 and/or (b) a main effect of Trial (Prediction 2), further post-hoc Bonferroni-holm corrected linear contrasts 527 upon the model will be performed to compare respectively, the performance with different ecological (non-) 528 valid barrier per Species and performance over Trials (within one session). In case we find or (c) a signifi-529 cant three-way Species x Barrier x Trial interaction effect(s) (Explorative Prediction 3), further exploratory 530 Bonferroni-Holm corrected linear contrasts upon the model will be performed to compare Trial performance 531 of Species on different types of the Barrier follow-up tests would be performed. We only found a significant 532 effect of Trial (Prediction 2). In line with the preregistration, we performed post-hoc Bonferroni-Holm cor-533 rected (Holm, 1979) linear contrasts upon the model to compare performance over trials (within one session). 534 Follow-up linear contrasts along with the corresponding and Cohen's d effect sizes (observed Cohen's d for 535 LMMs, Incidence Rate Ratio (IRR) for the negative binomial part of the ZINB model and Odds Ratios (OR) for 536 the zero-inflated part of the ZINB model) were calculated by means of the emmeans (Lenth, 2023) and Isr 537 (Cohen's d, Navarro, 2015) packages. 538

539	Results								
540	Detour Latency								
541	Registered comparisons Main Analyses								
542	<sup>2</sup> Descriptive statistics appear in Figure 4 and Tables 6, 7 and 8; inferential statistics appear in Table 5 and								
543	supplementary Table	es 1, 2 and 3 (Dewulf, Garcia	-Co, et al., <mark>202</mark>	<mark>3</mark> ). Th	ne Species	x Barrie	er ( <b>Prediction 1</b> ) and		
544	the Species x Barrier	x Trial ( <b>Prediction 3</b> ) intera	ction effects we	ere no	ot significa	nt. How	ever, there was a sig-		
545	nificant main effect o	f Trial ( <b>Prediction 2</b> ), as sho	wn in Table <mark>5</mark> .						
546	<b>Table 5.</b> Output: LMM with temporal correlation structure on detour latency (s)								
		Parameter	$ ilde{\mathrm{X}}^2$	Df	Р	np <sup>2</sup>			
		(Intercept)	1322.498	1	<0.001				
		Species	77.015	3	<0.001	0.249			
		<b>B</b>	4 2 4 2	4	0 2 4 6	0.004			

(incercept)	1522.450		-0.001	
Species	77.015	3	<0.001	0.249
Barrier	1.343	1	0.246	0.001
Trial	64.249	2	<0.001	0.051
Baseline_centered	0.568	1	0.451	0.000
BarrierOrder	2.852	1	0.091	0.013
Barrier:Trial	0.151	2	0.927	0.000
Species:Barrier	0.147	3	0.986	0.000
Species:Trial	56.035	6	<0.001	0.045
Species:baseline_Centered	5.452	3	0.142	0.023
Species:Barrier:Trial	8.228	6	0.222	0.007

Note. Significant effects are indicated with bold p-values

<sup>547</sup> Follow-up contrasts upon the model for the main effect of Trial showed that performance improved over

trials, with significantly slower detour latencies on Trial 1 compared to Trial 2 and Trial 3. There was no signif-

icant difference in detour latencies on Trial 2 compared to Trial 3 (see Table 6 and Figure 4 panel A). Further

<sup>550</sup> inferential statistics are provided in the supplementary Table 1.

<sup>551</sup> **Table 6.** The model predicted means (on the log scale), the back-transformed model-predicted means (on the original scale, in seconds) and the observed means (also on the original scale, in seconds) for detour latency across different Trial levels.

Trial         Mean(SE)         CI         Mean         CI         Mean (           Trial1         2.005 (0.057)         1.892 - 2.118         7.428         6.635 - 8.316         15.276 (2.116)           Trial2         1.664 (0.058)         1.550 - 1.777         5.279         4.712 - 5.914         11.636 (2.116)	observed	p(model)	ex	Model		
Trial1         2.005 (0.057)         1.892 - 2.118         7.428         6.635 - 8.316         15.276 (2)           Trial2         1.664 (0.058)         1.550 - 1.777         5.279         4.712 - 5.914         11.636 (2)	Mean (SD)	CI	Mean	CI	Mean(SE)	Trial
Trial2 1.664 (0.058) 1.550 - 1.777 5.279 4.712 - 5.914 11.636 (2)	.276 (22.114)	6.635 - 8.316	7.428	1.892 - 2.118	2.005 (0.057)	Trial1
	.636 (21.879)	4.712 - 5.914	5.279	1.550 - 1.777	1.664 (0.058)	Trial2
<b>Trial3</b> 1.617 (0.057) 1.504 - 1.730 5.038 4.500 - 5.640 10.418 (15)	.418 (18.523)	4.500 - 5.640	5.038	1.504 - 1.730	1.617 (0.057)	Trial3

552

# Additional Exploratory Analyses

<sup>553</sup> Further examination of the model revealed an unexpected significant main effect of Species (see Table 5).

All pairwise comparisons were statistically significant (see Table 7 and Figure 4 panel B). Further inferential

- statistics are provided in the supplementary Table 2).
- <sup>556</sup> **Table 7.** The model predicted means (on the log scale), the back-transformed model-predicted means (on the original scale, in seconds) and the observed means (also on the original scale, in seconds) for detour latency across different Species levels.

	Мо	del	ex	p(model)	Observed
Species	Mean(SE)	CI	Mean	CI	Mean (SD)
Canary	2.002 (0.097)	1.810 - 2.194	7.404	6.110 - 8.972	14.591 (22.203)
Chicken	1.531 (0.097)	1.339 - 1.724	4.625	3.817 - 5.604	8.245 (13.867)
Gull	2.313 (0.097)	2.121 - 2.505	10.100	8.335 - 12.239	18.517 (27.911)
Quail	1.202 (0.099)	1.007 - 1.397	3.327	2.736 - 4.044	8.280 (14.771)

There was also an unexpected interaction between Species x Trial (see Table 5). Post-hoc linear contrasts showed that, performance improved for Canaries, Chickens and Quails, but not for Gulls. Specifically, Quails exhibited slower detour latencies on Trial 1 compared to Trial 2 and Trial 3. Quails were also significantly faster on Trial 3 than Trial 2. Canaries and Chickens exhibited slower detour latencies on Trial 1 compared to Trial 2 and Trial 3. There was no significant effect between Trial 2 and Trial 3 for either species. For Gulls detour latencies were significantly faster on Trial 1 compared to Trial 3. Similarly, detour latencies on Trial 2 were significantly faster than on Trial 3. No significant effect was observed between Trial 1 and Trial 2 (see

- Table 8 and Figure 4 panel C). Further inferential statistics are provided in the supplementary Table 3.
- <sup>565</sup> **Table 8.** The model predicted means (on the log scale), the back-transformed model-predicted means (on the original scale, in seconds) and the observed means (also on the original scale, in seconds) for detour latency across different Trial nested within Species levels.

	Model		exp(model)		Observed
Condition	Mean(SE)	CI	Mean	CI	Mean (SD)
Canary					
Trial1	2.417 (0.114)	2.192 - 2.642	11.209	8.952 - 14.035	20.890 (26.208)
Trial2	1.874 (0.115)	1.647 - 2.100	6.511	5.193 - 8.165	13.992 (24.149)
Trial3	1.716 (0.114)	1.491 - 1.941	5.561	4.441 - 6.963	8.892 (12.004)
Chicken					
Trial1	1.888 (0.114)	1.664 - 2.113	6.608	5.278 - 8.274	11.423 (17.141)
Trial2	1.395 (0.115)	1.169 - 1.621	4.035	3.218 - 5.060	6.510 (9.563)
Trial3	1.311 (0.114)	1.086 - 1.536	3.709	2.963 - 4.644	6.801 (13.399)
Gull					
Trial1	2.220 (0.114)	1.995 - 2.445	9.206	7.352 - 11.527	16.243 (22.341)
Trial2	2.213 (0.115)	1.986 - 2.439	9.140	7.289 - 11.461	18.424 (31.530)
Trial3	2.505 (0.114)	2.280 - 2.730	12.247	9.781 - 15.335	20.885 (29.086)
Quail					
Trial1	1.496 (0.116)	1.268 - 1.725	4.465	3.552 - 5.612	12.452 (20.766)
Trial2	1.174 (0.117)	0.944 - 1.404	3.234	2.569 - 4.071	7.479 (12.313)
Trial3	0.936 (0.116)	0.707 - 1.164	2.549	2.028 - 3.204	4.909 (6.770)



**Figure 4.** Visualization of model-predicted means (back-transformed to detour latency in seconds) along with their 95% CI across different Trial levels (Panel A), Species (Panel B), and Species by Trial interactions (Panel C). Significant effects are indicated as follows: \* p < 0.05, \*\* p < 0.01, \*\*\* p < 0.001.

## Persisting

566 567

## Registered comparisons Main Analyses: Negative Binomial Part

Descriptive statistics for the negative binomial part of the model appear in Figure 5 and Tables 10, 11 and 12; inferential statistics appear in Table 9 and supplementary Tables 4, 5 and 6 (Dewulf, Garcia-Co, et al., 2023).

The Species x Barrier (**Prediction 1**) and the Species x Barrier x Trial (**Prediction 3**) interaction effects were

not significant. However, there was a significant main effect of Trial (**Prediction 2**), as shown in Table 9.

**Table 9.** Output: GLMM on persisting (frames)

Parameter	$ ilde{\mathrm{X}}^2$	Df	р
	Negative bi	nomi	al part
(Intercept)	9266.6 <mark>59</mark> 80	1	<0.001
Species	24.031	3	<0.001
Barrier	0.005	1	0.942
Trial	97.22 <mark>5</mark> 2	2	<0.001
baseline_centered	1.506	1	0.220
BarrierOrder	2.30 <mark>1 2</mark>	1	0.129
Barrier:Trial	8.514	2	0.014
Species:Barrier	5.292	3	0.152
Species:Trial	3.949	6	0.684
Species:baseline_centered	4.20 <mark>1 0</mark>	3	0.241
Species:Barrier:Trial	4.150	6	0.656
	Zero-infl	ated	part
(Intercept)	153.73 <mark>2</mark> 1	1	<0.001
Barrier	11.758	1	< 0.001
baseline_centered	12.73 <mark>0</mark> 3	1	<0.001
Species	174.5 <mark>49</mark> 52	3	<0.001
Trial	35.177	2	<0.001
Species:Trial	12.57 <mark><del>2</del> 3</mark>	6	0.050

Note. Significant effects are indicated with bold p-values

<sup>572</sup> Follow-up contrasts upon the negative binomial part of the model to further investigate the main effect of

<sup>573</sup> Trial showed that performance improved over trials, with significantly more persisting on Trial 1 compared

to Trial 2 and Trial 3. There was also significantly more persisting on Trial 2 compared to Trial 3 (see Table 10

and Figure 5 panel A). Further inferential statistics are provided in the supplementary Table 4.

<sup>576</sup> **Table 10.** The model predicted means (on the log scale), the back-transformed model-predicted means (on the original scale, in frames) and the observed means (also on the original scale, in frames) for persisting across different Trial levels.

	Model			exp(model)	Observed
Trial	Mean(SE)	CI	Mean	CI	Mean (SD)
Trial1	5.594 (0.070)	5.456 - 5.732	268.7 <mark>72</mark> 69	234.14 <mark>7 4</mark> - 308.51 <mark>8 4</mark>	295.861 (389.194)
Trial2	5.025 (0.077)	4.875 - 5.176	152.233	130.941 - 176.987	204.441 (396.460)
Trial3	4.664 (0.083)	4.501 - 4.827	106.03 <mark>6</mark> 7	90.09 <mark>4 5</mark> - 124. <del>799</del> 800	148.066 (288.298)

## Additional Exploratory Analyses Negative Binomial Part

Further examination of the negative binomial part of the model revealed an unexpected significant main effect of Species (see Table 9). Post-hoc linear contrasts showed that Canaries and Chickens persisted less compared to Gulls and Quails. All pairwise comparisons were statistically significant, except for the comparisons between Canaries and Chickens, and between Gulls and Quails (see Table 11 and Figure 5 panel B). Further inferential statistics are provided in the supplementary Table 5. The main effect of Species will be addressed further in the discussion.

Table 11. The model predicted means (on the log scale), the back-transformed model-predicted means (on the original scale, in frames) and the observed means (also on the original scale, in frames) for persisting across different Species levels.

Model				exp(model)	observed
Species	Mean(SE)	CI	Mean	CI	Mean (SD)
Canary	4.827 (0.114)	4.604 - 5.051	124.88 <mark>5</mark> 6	99.898 - 156.123	190.718 (337.446)
Chicken	4.878 (0.092)	4.698 - 5.059	131.41 <mark>7 8</mark>	109.747 - 157.36 <mark>7 8</mark>	152.094 (171.974)
Gull	5.396 (0.093)	5.214 - 5.579	220.62 <mark>5</mark> 7	183.83 <mark>6 8</mark> - 264.77 <mark>5 8</mark>	305.363 (505.505)
Quail	5.275 (0.120)	5.039 - 5.511	195.42 <mark>4 0</mark>	154.33 <mark>4 1X</mark> X - 247.4 <mark>54</mark> 48	238.181 (345.849)

There was also an unexpected interaction between Barrier x Trial (see Table 9). Post-hoc linear contrasts showed that overall performance improved with both types of barriers. For the Horizontal-Bar Barrier, individuals persisted significantly more on Trial 1 compared to Trial 2 and Trial 3. There was no significant difference between Trail 2 and Trial 3. For the Vertical-Bar Barrier, individuals persisted significantly more on Trial 1 compared to Trial 2 and Trial 3. In addition, there was significantly more persisting on Trial 2 compared

- to Trial 3 (see Table 12 and Figure 5 panel C). Further inferential statistics are provided in the supplementary
- 591 Table 6).

577

Table 12. The model predicted means (on the log scale), the back-transformed model-predicted means (on the original scale, in frames) and the observed means (also on the original scale, in frames) for persisting across Trial nested in Barrier levels.

	Мо	del	ex	exp(model)			
Condition	Mean(SE)	CI	Mean	CI	Mean (SD)		
Horizontal							
Trial1	5.635 (0.102)	5.435 - 5.835	<del>280.005</del> 279.997	229.29 <mark>8 3</mark> - 341.9 <mark>24</mark> 14	18.442 (23.536)		
Trial2	4.862 (0.107)	4.653 - 5.072	129.346	104.915 - 159.466	14.458 (24.518)		
Trial3	4.776 (0.119)	4.543 - 5.009	118.6 <mark>09</mark> 10	93.96 <mark><del>6</del> 8</mark> - 149.71 <mark>3</mark> 5	16.337 (26.444)		
Vertical							
Trial1	5.553 (0.092)	5.372 - 5.734	257.99 <mark>0</mark> 1	215.250 - 309.218	19.378 (24.638)		
Trial2	5.188 (0.102)	4.989 - 5.388	179.170	146.77 <mark><del>3</del> 4 - 218.71<del>6</del> 7</mark>	16.757 (26.925)		
Trial3	4.552 (0.108)	4.340 - 4.763	94.79 <mark>6</mark> 7	76.74 <mark><del>3</del> 4</mark> - 117.097	12.214 (17.545)		



**Figure 5.** Visualisation of the negative binomial model-predicted means (back-transformed log estimates to the original scale, representing, persisting in frames) along with their asymptotic CI across different Trial levels (Panel A), Species (Panel B) and Trial by Barrier interactions (Panel C). Significant effects are indicated as follows: \* p < 0.05, \*\* p < 0.01, \*\*\* p < 0.001.

## **Additional Analyses Zero-Inflated Part**

In addition to the negative binomial component, the statistical model for persisting also included a zeroinflated component that accounted for excess zeros in persisting (i.e., capturing all birds that did not persist). Descriptive statistics for the zero-inflated part of the model appear in Figure 6 and Tables 13, 14, 15 and 16; inferential statistics appear in Table 9 and supplementary Tables 7, 8 and 9 (Dewulf, Garcia-Co, et al., 2023).

- Examination of the zero-inflation part of the model revealed a significant main effect of Trial (see Table 9). Post-hoc linear contrasts showed that overall there was an increase in the probability of zeros for persisting in later trials, indicating improved accuracy. Specifically, Trial 3 showed a significant higher probability of zeros for persisting compared to Trial 2 and Trial 1. Trial 2 also had a significantly higher probability of zeros for persisting compared to Trial 1 (see Table 13 and Figure 6 panel A). Further inferential statistics are provided in the supplementary Table 7.

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Table 13. The model predicted means (on the log odds ratio scale), the back-transformed zero probabilities (on the original scale, in frames) and the observed zero probabilities (also on the original scale, in frames) for persisting across Trial levels.

	Мс	odel	Pro	ob(model)	Observed
Trial	Mean(SE)	CI	prob	CI	prob
Trial1	-1.438 (0.143)	-1.7181.158	0.192	0.153 - 0.239	0.244
Trial2	-0.832 (0.116)	-1.0580.605	0.303	0.258 - 0.353	0.342
Trial3	-0.388 (0.107)	-0.5970.180	0.404	0.355 - 0.455	0.429

Again, we had not predicted a main effect of Species (see Table 9). However, Canaries and Quails had a

<sup>607</sup> higher probability of zeros for persisting (indicating higher accuracy) compared to Chickens and Gulls. All pair-

<sup>608</sup> wise comparisons were statistically significant, exception for the comparisons between Canaries and Quails,

- and between Chickens and Gulls (see Table 14 and Figure 6 panel B). Further inferential statistics are provided
- in the supplementary Table 8). The main effect of Species will be addressed further in the discussion.
- Table 14. The model predicted means (on the log odds ratio scale), the back-transformed zero probabilities (on the original scale, in frames) and the observed zero probabilities (also on the original scale, in frames) for persisting across Species levels.

	Мо	del	Pro	ob(model)	Observed
Species	Mean(SE)	CI	prob	CI	prob
Canary	-0.139 (0.114)	-0.362 - 0.085	0.466	0.410 - 0.521	0.478
Chicken	-1.831 (0.175)	-2.1731.489	0.138	0.102 - 0.184	0.169
Gull	-1.785 (0.157)	-2.0931.477	0.144	0.110 - 0.186	0.158
Quail	0.211 (0.111)	-0.006 - 0.428	0.553	0.499 - 0.605	0.555

- There was also an unexpected significant main effect of Barrier (see Table 9). Post-hoc linear contrasts showed a significant higher probability of zeros for persisting (indicating higher accuracy) for the Horizontal-Bar Barrier than Vertical-bar Barrier (see Table 15 and Figure 6 panel C). Further inferential statistics are
- provided in the supplementary Table 9)
- <sup>616</sup> Table 15. The model predicted means (on the log odds ratio scale), the back-transformed zero probabilities (on the original scale, in frames) and the observed zero probabilities (also on the original scale, in frames) for persisting across Barrier levels.

	Мс	odel	Pro	ob(model)	Observed
Barrier	Mean(SE)	CI	Prob	CI	Prob
Horizontal	-0.665 (0.092)	-0.8460.485	0.339	0.300 - 0.381	0.377
Vertical	-1.106 (0.100)	-1.3020.910	0.249	0.214 - 0.287	0.300

An unexpected main effect of Baseline was also observed (see Table 9). Follow-up analyses indicated that the probability of zeros for persisting was estimated at 0.292, when birds had an average motivation score

- (Baseline at zero, due to within-species mean-centering). Descriptive statistics appear in Table 16; inferential
- statistics appear in Table 9.
- <sup>621</sup> **Table 16.** The model predicted means (on the log odds ratio scale), the back-transformed zero probabilities (on the original scale, in frames) and the observed zero probabilities (also on the original scale, in frames) for the average value of Baseline.



**Figure 6.** Visualisation of the zero-inflated model-predicted means (back-transformed to the probability zeros for persisting in frames) along with their asymptotic CI across different Trial (Panel A), Species (Panel B) and Barrier levels (Panel C). Significant effects are indicated as follows: \* p < 0.05, \*\* p < 0.01, \*\*\* p < 0.001.

# Additional Analysis: Group Size as a Random effect

Due to post-hatch mortality (in canaries, gulls, and quails), group size deviated slightly from the intended 623 10 individuals per group (Table 17) as batch incubation limited the ability to replace these losses with agematched individuals. Note that group sizes greater than 10 resulted from the merging of two high mortality 625 enclosures where birds were of the same age. For the canaries, this was further complicated by the need 626 to introduce 'tutors' for the juveniles (i.e. adult demonstrators to teach independent feeding), which meant 627 that each group of juveniles (about 10) had at least one extra adult for a few days. To examine the possible 628 impact of variation in group size on performance, additional analyses were conducted using 'Group Size' as 629 a random variable. For both detour latency and persistence, the extended model did not outperform the 630 above-mentioned reported models. As a result, these additional analyses are discussed in the supplementary 631 materials (Table 10, Table 11). 632

**Table 17.** Visualization of the number of individuals that met our exclusion criteria, in relation to the enclosure raised in each group size and the across species.

	Group Size								
Species	#6	#8	#9	#10	#11	#12	#14	Mean	Range
Canary	0	0	15	20	14	0	11	10.717	9 - 14
Chicken	0	0	0	60	0	0	0	10	10 - 10
Gull	0	22	3	35	0	0	0	9.217	8 - 10
Quail	3	2	7	39	0	7	0	9.845	6 - 12

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# Discussion

We argue that stop-signal detection is a critical cognitive component of response inhibition (RI) across 634 species, including birds. This study explored this idea further by investigating whether RI is improved when the 635 perceptual characteristics of the stop signal (i.e., barrier) in the detour barrier task correspond to the species' 636 ecological niche, as shown by Regolin et al. (1994) and Zucca et al. (2005). However, we failed to replicate this 637 earlier work, as RI was not significantly improved when the barrier type supposedly matched the ecological 638 niche of the species (Prediction 1). However, we did find that performance generally improved over trials 639 (**Prediction 2**) for both detour latency and persisting, but again, this did not interact with the species-specific 640 ecological validity of the stop signal vary between species (Prediction 3). The latter indicates that learning 641 occurs independently of the ecological validity of the stop signal. 642

643

Most importantly, we were unable to replicate the finding that barrier type had a species-specific influence 644 on detour performance, even though our study has several methodological and conceptual strengths, includ-645 ing a well-powered design, standardised experimental procedures, controlled prior experience (through pre-646 and mid-test exclusion criteria), and baseline measures (to minimise confounding by non-cognitive, motiva-647 tional traits). Thus, our findings do not support the ecological-niche hypothesis as proposed by Regolin et al. 648 (1994) and Zucca et al. (2005), suggesting that the adaptation to a specific ecological niche cannot account 649 for variation in stop-signal detection (at least, not in the detour task). This does not necessarily imply that 650 stop-signal detection is not important at all for RI, but it does indicate that differences between the four bird 651 species tested here are not caused by variation in how they perceive or interpret vertical and horizontal-bar 652 barriers. 653

654

As trials progressed, most individuals became faster in detouring (except for gulls) and made fewer attempts to persist at interacting with the barrier, regardless of the barrier type (confirming Prediction 2 but disconfirming Prediction 3). In the habituation set-up (or training phase) of our study, the food bowl was placed *in front* of the opaque barrier; this ensured that birds had no prior experience of retrieving food from behind a barrier (which standardised baseline performance). But without this experience, in the test phase,
individuals had to learn both to inhibit their prepotent response to go directly for the reward (as the direct
path is blocked) their initial behaviour and to navigate around the barrier (Van Horik, Beardsworth, Laker,
Whiteside, et al., 2020), explaining the observed improvements over trials.

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The learning pattern observed for the gulls was unexpected, as it appears that gulls learned to inhibit inter-664 acting with the barrier itself but without an overall improvement in detour latency, whereas the other species 665 became faster at detouring and interacted less over time with the barrier. At present, we have no explana-666 tion as to why, for gulls, learning was only observed for the persistence measure and not detour latency as 667 was seen in the other three species. However, this pattern demonstrates the value of looking at detour la-668 tency and time spent interacting with the barrier. One might assume that lower persistence scores should 669 automatically result in shorter detour latencies but for gulls, this was not the case. This indicates that overall 670 task performance (i.e., detour latency) captures additional behaviours, potentially unrelated to response in-671 hibition (e.g., the time taken to approach the barrier, time spent not interacting, time needed to navigate the 672 barrier, etc.). The observed differences in learning also highlight two further issues. First, the fact that gulls 673 showed evidence of learning in measures of persistence but not in the measure of detour latency suggests 674 that, at least for some species, tasks include several subcomponents and that some of these are not equally 675 influenced by learning across species certain task components are more influenced by learning (inhibiting an 676 unrewarded repetitive response) than others (inhibiting the response to go straight for the food or navigating 677 around a barrier, which are both captured by detour latency). Speculatively, this could be related to the ecological niche adaptations of the species as well. Certain behaviours, such as inhibition of unrewarded responses, 679 may be more critical than others in certain ecological niches, making them easier to learn. In contrast, other 680 behaviours, such as navigating obstacles, may be more influenced by context-specific factors, and therefore, 681 harder to learn for certain species (although follow-up work is required to test this idea). Second, the learning 682 differences stress the need to take the role of learning in RI (and cognition in general) into account when aim-683 ing to interpret the variation in RI between species. For example, while canaries and gulls were initially slow 684 at detouring (compared to chickens and quails; Figure 4), detour latencies of canaries gradually decreased. 685 while those of gulls did not. This suggests that the differences between these two species in a putative test of 686 response inhibition could at least partly reflect variations in learning rather than inhibition, with interaction effects between species and trial potentially explaining these findings (Willcox et al., 2024).

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Alongside the effects of trial, we also identified general latency differences between species. Even though 690 such differences are hard to interpret, one notable finding stands out, namely that gulls appeared to 'under-691 perform' compared to the other species, as they were generally slower (compared to the three other species; 692 Figure 4), more likely to peck (compared to canaries and quails: Figure 6), and when they did peck, they pecked 693 for longer (compared to canaries and chickens; Figure 5). As noted above, the gulls' detour latencies also did 694 not decrease over trials. We consider two (not mutually exclusive) hypotheses. First, the gulls are a wild 695 species, whereas the three other species are domesticated. Domesticated species are generally less fearful 696 and stressed than wild species (Kaiser et al., 2015). For example, Giøen and colleagues (2023) compared the 697 behavioural responses of white leghorn chickens with their wild counterparts, red junglefowls (Gallus gallus), 698 in risk-taking situations, such as the encounter of a novel object during food retrieval. They found that red 699 junglefowls were more stressed and fearful of the object and reached the food later than white leghorns. If 700 gulls were indeed more fearful and stressed than the other species, this could have influenced their detour 701 performance. Consistent with this idea, there was a high number of drop-outs among gulls (compared to the 702 three other species: Table 4) due to the pre-test (i.e., a failure to interact with the food bowl in the presence 703 of a novel barrier in a new test environment) or mid-test 1 (i.e., a failure to obtain a measure for one of the 704 two dependent variables during a test trial, indicating little interaction with the experimental task) exclusion 705 criteria. Second, even though we standardised the testing age in terms of number of days, the developmental

trajectories of RI (and cognition in general) may have differed between species. Gulls have a much longer maximum lifespan (49 years) compared to canaries (24 years), chickens (15-20 years), and quails (6 years; Tacutu
et al., 2018); based on life history theories, one could speculate that neuro-cognitive development would
be protracted in the semi-precocial and long-living gulls compared with e.g. the precocial and shorter-living
quails and chickens (Bunge, 2012). However, this idea should be further tested.

Finally, it is noteworthy that canaries successfully solved (and learned) the detour problem (irrespective of 712 barrier type). In contrast, while in the study of Zucca et al. (2005), canaries were unable to solve the detour 713 problem and repeatedly attempted to fly over the barrier (again, irrespective of barrier type). The authors 714 attributed this inability to the canaries' adaptation to an aerial environment, which enables them to navigate 715 obstacles by simply flying over them in natural environments. However, several other studies have already 716 shown that species, adapted to an aerial lifestyle, such as ravens (Kabadayi, Jacobs, et al., 2017), ring doves 717 (Streptopelig risoria, Miller, 1974; Miller and Tallarico, 1974) and pigeons (Columbia livia, Miller, 1974; Miller 718 and Tallarico, 1974), are capable of solving the detour barrier task as well. We speculate that canaries were 719 able to solve the detour problem in our study, but not in the original work, due to the exclusion criteria we 720 implemented, which ensured proficiency with the basic task demands (e.g., the perceptual, motoric, and mo-721 tivational requirements for retrieving a food reward; MacLean et al., 2014). Specifically, our pre-test exclusion 722 criterion ensured that all included birds visited and ate from a food bowl placed in front of a barrier (novel 723 object) in the habituation phase before access to the food bowl was restricted by moving the barrier in front 724 of it in the test phase. We believe that experience with retrieving the reward may be critical for measuring 725 detour performance, potentially more so in aerially adapted birds. After all, Zucca et al. (2005) found that, 726 even after prolonged exposure to the test situation, a large proportion of canaries were unable to solve the 727 detour problem. This suggest, that the problem was not a lack of familiarity with the test itself, but rather a 728 lack of experience with retrieving the reward. However, this explanation is speculative and requires further 729 investigation. 730

In summary, we failed to provide support for the 'ecological niche hypothesis', as proposed by (Regolin, 731 Vallortigara, et al., 1994; Zucca, Antonelli, et al., 2005). Our study adds to the growing body of evidence for the 732 critical need for replication studies (Farrar et al., 2020), and highlights the need to consider methodological 733 and conceptual design factors, as these can significantly impact results. Although our study did not provide 734 strong evidence for the idea that interspecies differences in the perception of barrier types influence detour 735 performance (and cause species differences), this does not negate the need for further research into the 736 influence of the characteristics of the stop signal or other underlying mechanisms of RI. More generally, future 737 research should focus on the cognitive mechanisms underlying RI. Understanding these mechanisms will 738 help explain inter-individual variation such as in decision-making in dynamic environments (Johnson-Ulrich 739 and Holekamp, 2020), predator avoidance and foraging optimization (Tvardíková and Fuchs, 2012), as well as 740 responses to broader ecological pressures (Lee and Thornton, 2021). 741

#### 742

# Acknowledgements

We thank Dries Debeer for statistical support, and two anonymous reviewers and the specialist recommender for their helpful comments on the paper. We also thank Sophia Knoch for assisting with data validation, as well as to the staff of the ECoBird Centre, the Department of Biology at the University of Antwerp and
the Wildlife Rescue Centre Ostend for general project support. Preprint Stage 1 version 2 of this article has
been peer-reviewed and recommended by Peer Community In Registered Reports (https://osf.io/qvxgh).

# **Fundings and Ethics**

This work is supported by a FWO (Flemish Research Foundation) PhD Fellowship grant (No. 11F0823N, awarded to Anneleen Dewulf), an ERC Consolidator grant (European Union's Horizon 2020 research and innovation programme, grant agreement No. 769595, awarded to Frederick Verbruggen), and a Methusalem grant (Methusalem Project 01M00221, Ghent University, awarded to Frederick Verbruggen, Luc Lens and An Martel). This experiment has been approved by the Ethical Committee for Animal Experimentation, Faculty of Sciences, Flemish Institute for Biotechnology (VIB) of Ghent University (EC2022-091, EC2024-023).

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# **Conflict of interest disclosure**

The authors declare that they comply with the PCI rule of having no financial conflicts of interest in relation
to the content of the article.

# <sup>759</sup> Data, script, code, and supplementary information availability

Data, script, codes and supplementary information is available at the OSF repository (Dewulf, Garcia-Co,
 et al., 2023, https://osf.io/u63w2/?view\_only=db4c1a620de841c28d7fc9a52e326cfd)

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# **Authors' Contributions**

'Conceptualization' – AD, CG, WM, JM, AM, LL, FV ; 'Data Curation' – AD; 'Formal Analysis' – AD; 'Funding
 Acquisition' – AD, AM, LL, FV; 'Investigation' - AD, CG; 'Methodology' - AD, CG, WM, JM, AM, LL, FV; 'Project
 Administration' – AD, FV; 'Software' – AD; 'Resources' – AD, WM, FV; 'Supervision' – JM, LL, FV; 'Validation' – AD,
 JM, LL, FV; 'Visualization' – AD; 'Writing – Original Draft' – AD; 'Writing – Review & Editing' – AD, CG, WM, JM, AM,
 LL, FV.

Question	Hypothesis	Sampling plan	Analysis Plan	Rationale for	Interpretation given	Theory that could be
				deciding the	different outcomes	shown wrong by the
				sensitivity of the		outcomes
				test for confirming		
				or disconfirming		
				the hypothesis		
1	As we will test four bird	We will test 60 individuals per	A (G)LMM with type III	We performed a-	This study is a partial	We propose that stop-
Does detour	species with	species (total N = 240). A-priori	sum of squares will be used	priori power	replication of the	signal detection
performance	substantially different	power sensitivity analyses done	to analyse our two	sensitivity analyses	study of Regolin et	(hence, barrier
improve when	ecological niches, the	in G*Power (Faul et al.,	dependent variables,	(alpha corrected =	al., (1994) and the	detection) is a crucial,
the perceptual	perception of a match	2009) indicate that this is	namely 1) the latency to	.025, Power = $.80$ ,	comparative study of	cognitive building
characteristics of	between the perceptual	sufficient to detect small effects;	detour and the 2)	cor. among RM =	Zucca et al., (2005).	block of RI across
the barrier (in a	characteristics and the	it is also the largest number that	cumulative time	0.5) with G*Power	Across research fields,	species (Verbruggen
detour task)	ecological niche will be	is practically feasible given our	spent in the species-	(Faul et al., 2009)	there is a clear need	et al., 2014), including
match the	species-specific (see	resource constraints and study	specific 'barrier zone of	for the different	for these replication	birds.
ecological niche	below). We therefore	design (see Farrar et al., (2020))	interest'(persisting).	effects of interest	studies (see Farrar et	
of a bird species?	predict an interaction	For the sensitivity analysis, we		and confirmed that	al., 2020, for the	Here we will take this
	between barrier type	used a mixed ANOVA model	Both models will include	our sample size is	comparative cognition	idea one step further
	and species.	with one between-subjects	the between-species factor:	sufficient to detect	field), as they are "a	and propose that
		factor (4 levels; corresponding	Species (i.e., 4 levels) and	small effects (with a	central part of the	ecologically valid
	1:1	to our Species factor) and two	both within-species factors:	Cohen's f effect size	iterative maturing	signals are easier to
	For white leghorn	within-subjects factor (one with	Barrier (i.e., 2 levels) and	varying between	cycle of description,	detect (or to perceive
	chickens (Gallus gallus	2 levels $- Barrier - and$	Trial (i.e., 3 levels), and	0.09 -0.12 ; Cohen,	prediction, and	as a stop signal), and
	domesticus) and Japanese	one with 3 levels – <i>Trial</i> ).	the two control variables	1988).	explanation", and as	this will enhance
	quails (Coturnix		(as covariates), namely (a)		such, play an	stopping.
	<i>japonica</i> ): better detour	This indicated that our sample	a 'muti-baseline' measure	See sampling plan	"important, exciting,	
	performance for	size of 60 animals per species	of an individual's	for more	and generative role in	The role of stop-signal
	horizontal- (HB)	(240 in total) is sufficient	motivational state (and its	information.	scientific progress"	detection in avian
	compared with vertical-	to detect a Species x Barrier	interaction with Species, as		(Nosek & Errington,	response inhibition,
	bar barriers (VB; thus,	interaction effect (Question 1)	we will mean-center this		2020).	and in particular, the
	detour performance HB >	with a small effect size (Cohen's	'multi-baseline' measure			interaction with the
	VB).	f effect size of 0.12; Cohen,	within <i>Species</i> ), and (b)		As a consequence, the	ecological niche of the
		1988; Power = .80; cor. among	Barrier Order (i.e., 2		absence of a Barrier x	species, should be
	1:2	RM = 0.5; we used an alpha of	levels). Individual birds		Species interaction	revised if we cannot
	For herring gulls (Larus	.025 to correct for the fact that	and enclosure (social		effect (which we	replicate the previous
	argentatus): better detour	we will have two dependent	group) will be included as		predict on the basis of	work (Regolin et al.,
	performance for vertical-	variables measuring (slightly)	random effects in the		the previous studies)	1994; Zucca et al.,
	compared with	different aspects of detour	models, with individual		informs us about the	2005).

horizonal-bar barriers (Ins., detour performance)performance).birds nested in enclosures In addition, we will include by-individual crested in andoms loops of Trial Question 2; Cohen's f of Cohen's f <th></th> <th></th> <th></th> <th></th> <th>4</th> <th>1</th>					4	1
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discussed below, we will analyse our data with (G)LMMs, which are currently ot covered by G*Power or most other power-estimation tools. These mixed-effectquantiles (y-axis) on the standard normalmost other power-estimation tools. These mixed-effect3) outliers (plotting standard residuals (y-axis) and leverage).3) outliers (plotting multicollinearity between fixed and random effects.However, by accounting for unexplained variance (see below), our proposed mixed effect models are more powerful than the fixed-effect model subscinction between residuals (via a Durbin- Watson-Test) will be calculated via functions provided by the performance package (Lidecke et al., 2021).Withe standard normal standard normal		species factors). However, as	(plotting the sample			
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unexplained variance (see below), our proposed mixed effect models are more powerful than the fixed-effect model analyses (and than the models used in the studies of Regolin,autocorrelation between residuals (via a Durbin- Watson-Test) will be calculated via functionsMarch 1March 2Calculated via functions provided by the performance package (Lüdecke et al., 2021).Image: Calculated via functions		However, by accounting for	VIF) and the			
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analyses (and than the models used in the studies of Regolin,performance package (Lüdecke et al., 2021).		ANOVAs used in our sensitivity	provided by the			
used in the studies of Regolin, (Lüdecke et al., 2021).		analyses (and than the models	performance package			
		used in the studies of Regolin,	(Lüdecke et al., 2021).			

	Vallortigara, et al., 1994, and Zucca, Antonelli, et al., 2005). Thus, the sensitivity analyses discussed here are a conservative estimate. <u>Remark 2:</u> We will incubate 20% more eggs than the number of individuals required for testing (to account for possible drop outs and guarantee statistical power during the whole study)	Potential violations of model assumptions will be addressed by transforming the (in)dependent variables (i.e., via log- transformation) or by changing the error distribution (family) or the link function of the model (switching a default LMM that will be fitted to a GLMM). Fixed main effects with a VIF of >5 will be removed and logical outliers (i.e., recording/entry errors) will be inspected and corrected (if possible). In the case that the outlier cannot be corrected, all data of that individual will be excluded from all statistical analyses		
	drop outs and guarantee	that will be fitted to a		
	statistical power during the	GLMM). Fixed main		
	whole study)	effects with a VIF of $>5$		
		will be removed and		
		logical outliers (i.e.,		
		recording/entry errors) will		
		be inspected and corrected		
		(if possible). In the case		
		that the outlier cannot be		
		corrected, all data of that		
		individual will be excluded		
		from all statistical analyses.		
		In case we find (a)		
		significant Barrier x		
		Species interaction -		
		effect(s) (Question 1)		
		further post-hoc		
		Bonferroni-Holm corrected		
		linear contrasts upon the		
		model will be performed to		
		compare performance with		
		different (ecological valid)		
		barriers per species (1:1,		
		1:2, 1:3).		

2	2:1	In case we find (a)	The absence of a main	We propose that
Does detour	Detour performance will	significant main effect of	effect of trial would	detour performance
performance	gradually improve over	<i>Trial</i> (Question 2) further	demonstrate that	improves over trials.
improve over	trials (thus, detour	post-hoc Bonferroni-Holm	detour performance is	Extensive work on
trials?	performance trial 1 < trial	corrected linear contrasts	consistent over trials.	skill acquisition in
	2 < trial  3).	upon the model will be		humans has shown
		performed to compare		that performance
		performance over trials		generally improves
		(2:1).		rapidly at first and
				then more slowly over
				time (see e.g., Logan,
				1988, Thorndike,
				1913). If we do not
				find a difference
				between trials, this
				would indicate that
				detouring cannot be
				learned easily by
				avian species.
3 (explorative)	3:1 (explorative)	In case we find (a)	Our design enables us	We will <i>explore</i> if the
Does the learning	We will <i>explore</i> if the	significant three-way	to investigate	learning effect (i.e.,
effect (i.e.,	learning effect (i.e.,	Species x Barrier x Trial	potential interaction	improved detour
improved	improved detour	interaction effect(s)	effects with Trial	performance across
detour	performance across trials)	(Question 3 explorative),	(e.g., the three-way	trials) will depend on
performance	interacts with the	further exploratory	Species x Barrier x	the ecological validity
across trials)	ecological validity of the	Bonferroni-Holm corrected	<i>Trial</i> interaction	of signals.
depend on the	stop signals. There are	linear contrasts upon the	effect, Question 3	If we do not find such
ecological	two possible patterns that	model will be performed to	explorative), with	a three-way
validity of the	would result in a three-	compare <i>Trial</i> performance	which we can further	interaction effect, we
barrier?	way interaction between	of <i>Species</i> on different	explore whether	can conclude (a) that
	Species, Barrier	types of the Barrier (3:1	learning effects are	superior detour
	(horizontal- vs. vertical-	explorative).	completely absent in	performance with
	bar barriers), and <i>Trial</i>		our design or are	ecological valid than
	(1-3) (Explorative		influenced by the	non-valid trials is
	<b>Prediction 3</b> ). First,		(ecological validity)	independent of trial
	detour performance might		of the different barrier	number (in case we do
	be better for ecologically		types.	find a Species x
	valid compared with non-			Barrier interaction

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valid stop signals at the			effect) or (b) that the
beginning, but this			interaction between
pattern might diminish			the stop signal and the
over time as individuals			ecological niche of the
learn to stop (i.e., the			species should be
differences between			revised (in case we do
barrier types would			not find a <i>Species</i> x
decrease). Second, detour			Barrier interaction
performance might be			effect).
poor at the beginning for			
both barrier types, but			
learning to stop might be			
easier for ecologically			
valid signals compared			
with non-valid stop			
signals (i.e., the			
differences between			
barrier types would			
increase). Both patterns			
would be theoretically			
meaningful, but we do			
not have <i>a-priori</i>			
predictions about the			
direction of the three-way			
interaction.			
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# **Guidance Notes**

- Question: articulate each research question being addressed in one sentence.
- **Hypothesis**: where applicable, a prediction arising from the research question, stated in terms of specific variables rather than concepts. Where the testability of one or more hypotheses depends on the verification of auxiliary assumptions (such as positive controls, tests of intervention fidelity, manipulation checks, or any other quality checks), any tests of such assumptions should be listed as hypotheses. Stage 1 proposals that do not seek to test hypotheses can ignore or delete this column.
- Sampling plan: For proposals using inferential statistics, the details of the statistical sampling plan for the specific hypothesis (e.g power analysis, Bayes Factor Design Analysis, ROPE etc). For proposals that do not use inferential statistics, include a description and justification of the sample size.

- Analysis plan: For hypothesis-driven studies, the specific test(s) that will confirm or disconfirm the hypothesis. For non-hypothesis-driven studies, the test(s) that will answer the research question.
- Rationale for deciding the sensitivity of the test for confirming or disconfirming the hypothesis: For hypothesis-driven studies that employ inferential statistics, an explanation of how the authors determined a relevant effect size for statistical power analysis, equivalence testing, Bayes factors, or other approach.
- Interpretation given different outcomes: A prospective interpretation of different potential outcomes, making clear which outcomes would confirm or disconfirm the hypothesis.
- Theory that could be shown wrong by the outcomes: Where the proposal is testing a theory, make clear what theory could be shown to be wrong, incomplete, or otherwise inadequate by the outcomes of the research.

# References

Ashton BJ, AR Ridley, EK Edwards, and A Thornton (2018). Cognitive performance is linked to group size and

affects fitness in Australian magpies. en. Nature 554.7692, 364-367. ISSN: 0028-0836, 1476-4687. https:

//doi.org/10.1038/nature25503. 777 Bari A and TW Robbins (2013). Inhibition and impulsivity: Behavioral and neural basis of response control, en. 778 Progress in Neurobiology 108, 44–79. ISSN: 03010082. https://doi.org/10.1016/j.pneurobio.2013.06.005. 779 Bates D, M Mächler, B Bolker, and S Walker (2015). Fitting Linear Mixed-Effects Models Using Ime4. Journal of 780 *Statistical Software* 67.1, 1–48. https://doi.org/10.18637/jss.v067.i01. 781 Bates D, M Maechler, and M Jagan (2023). Matrix: Sparse and Dense Matrix Classes and Methods. R package 782 version 1.6-2. 783 Bekker EM, JL Kenemans, MR Hoeksma, D Talsma, and MN Verbaten (2005). The pure electrophysiology of 784 stopping. en. International Journal of Psychophysiology 55.2, 191-198. ISSN: 01678760. https://doi.org/10. 785 1016/j.jpsycho.2004.07.005. 786 Ben-Shachar MS. D Lüdecke, and D Makowski (2020), effectsize: Estimation of Effect Size Indices and Stan-787 dardized Parameters. Journal of Open Source Software 5.56. Publisher: The Open Journal, 2815. https://doi. 788 org/10.21105/joss.02815. 789 Boehler CN, TF Munte, RM Krebs, H Heinze, MA Schoenfeld, and J Hopf (2009). Sensory MEG Responses Predict 790 Successful and Failed Inhibition in a Stop-Signal Task. en. Cerebral Cortex 19.1, 134–145. ISSN: 1047-3211, 791 1460-2199. https://doi.org/10.1093/cercor/bhn063. 792 Bollweg G and S Sparber (1998). Relationships Between Midembryonic 5-HT2 Agonist and/or Antagonist Ex-793 posure and Detour Learning by Chickens. en. Pharmacology Biochemistry and Behavior 60.1, 47-53. ISSN: 794 00913057. https://doi.org/10.1016/S0091-3057(97)00555-8. 795 Brooks ME, K Kristensen, KI van Benthem, A Magnusson, CW Berg, A Nielsen, HJ Skaug, M Maechler, and BM 796 Bolker (2017). glmmTMB Balances Speed and Flexibility Among Packages for Zero-inflated Generalized 797 Linear Mixed Modeling. The R Journal 9.2, 378-400. https://doi.org/10.32614/RI-2017-066. 798 Bunge SA (2012). The Developing Human Brain: A Frontiers Research Topic. en. Frontiers in Human Neuroscience 6. ISSN: 1662-5161. https://doi.org/10.3389/fnhum.2012.00047. 800 Chen G, NE Adleman, ZS Saad, E Leibenluft, and RW Cox (2014). Applications of multivariate modeling to neu-801 roimaging group analysis: a comprehensive alternative to univariate general linear model. eng. NeuroIm-802 age 99. Place: United States, 571-588. ISSN: 1095-9572 1053-8119. https://doi.org/10.1016/j.neuroimage. 803 2014.06.027. 804 Cohen | (1988). Statistical power analysis for the behavioral sciences. en. 2nd ed. Hillsdale, N.I: L. Erlbaum Asso-805 ciates. ISBN: 978-0-8058-0283-2. 806 Devereux CL, MJ Whittingham, E Fernández-Juricic, JA Vickery, and JR Krebs (2006). Predator detection and 807 avoidance by starlings under differing scenarios of predation risk. en. Behavioral Ecology 17.2, 303-309. 808 ISSN: 1465-7279, 1045-2249. https://doi.org/10.1093/beheco/arj032. 809 Dewulf A, C Garcia-Co, W Müller, JR Madden, A Martel, L Lens, and F Verbruggen (2023). Do Ecological Valid Stop 810 Signals Aid Detour Performance? A Comparison of Four Bird Species. Available at: https://osf.io/u63w2/. 811 Dewulf A, S Knoch, JR Madden, A Martel, L Lens, and F Verbruggen (2025). Unpublished manuscript. 812 Dudusola I and H Bashiru (2020). Repeatability estimates of body weight and shank length in Japanese quail 813 (Coturnix coturnix Japonica). Nigerian Journal of Animal Science 22.1, 27–32. 814 Elchlepp H and F Verbruggen (2017). How to withhold or replace a prepotent response: An analysis of the 815 underlying control processes and their temporal dynamics. en. Biological Psychology 123, 250–268. ISSN: 816 03010511. https://doi.org/10.1016/j.biopsycho.2016.10.005. 817 Farrar B, M Boeckle, and N Clayton (2020). Replications in Comparative Cognition: What Should We Expect 818 and How Can We Improve? en. Animal Behavior and Cognition 7.1, 1-22. ISSN: 23725052, 23724323. https: 819 //doi.org/10.26451/abc.07.01.02.2020. 820

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Faul F, E Erdfelder, A Buchner, and AG Lang (2009). Statistical power analyses using G\*Power 3.1: Tests for cor-821 relation and regression analyses. en. Behavior Research Methods 41.4, 1149-1160. ISSN: 1554-351X, 1554-822 3528. https://doi.org/10.3758/BRM.41.4.1149. 823 Fox J and S Weisberg (2019). An R Companion to Applied Regression. Third. Thousand Oaks CA: Sage. 824 Fox J, S Weisberg, and B Price (2022). carData: Companion to Applied Regression Data Sets. 825 Friard O and M Gamba (2016). BORIS: a free, versatile open-source event-logging software for video/audio 826 coding and live observations. Methods in ecology and evolution 7.11. Publisher: Wiley Online Library, 1325-827 1330. 828 Garcia-Co C, | Morales, F Verbruggen, and W Müller (2024). Unpublished manuscript. Department of Biology, 829 Behavioural Ecology and Eco-Physiology Group, University of Antwerp. 830 Gjøen J, H Jean-Joseph, K Kotrschal, and P Jensen (2023). Domestication and social environment modulate fear 831 responses in young chickens. en. Behavioural Processes 210, 104906. ISSN: 03766357. https://doi.org/10. 832 1016/j.beproc.2023.104906. 833 Hartig F (2022). DHARMa: Residual Diagnostics for Hierarchical (Multi-Level / Mixed) Regression Models. R package 834 version 0.4.6. 835 Holm S (1979). A Simple Sequentially Rejective Multiple Test Procedure. en. Scand / Statist 6.2, 65-70. 836 Johnson-Ulrich L and KE Holekamp (2020). Group size and social rank predict inhibitory control in spotted 837 hyaenas. en. Animal Behaviour 160, 157-168. ISSN: 00033472. https://doi.org/10.1016/j.anbehav.2019.11. 838 020. 839 Kabadayi C, K Bobrowicz, and M Osvath (2018). The detour paradigm in animal cognition. en. Animal Cognition 840 21.1, 21-35. ISSN: 1435-9448, 1435-9456. https://doi.org/10.1007/s10071-017-1152-0. 841 Kabadayi C, I Jacobs, and M Osvath (2017). The Development of Motor Self-Regulation in Ravens. en. Frontiers 842 in Psychology 8, 2100. ISSN: 1664-1078. https://doi.org/10.3389/fpsyg.2017.02100. 843 Kabadayi C, A Krasheninnikova, L O'Neill, I van de Weijer, M Osvath, and AMP von Bayern (2017). Are parrots 844 poor at motor self-regulation or is the cylinder task poor at measuring it? en. Animal Cognition 20.6, 1137-845 1146. ISSN: 1435-9448, 1435-9456. https://doi.org/10.1007/s10071-017-1131-5. 846 Kaiser S. MB Hennessy, and N Sachser (2015). Domestication affects the structure, development and stability 847 of biobehavioural profiles. en. Frontiers in Zoology 12.Suppl 1, S19, ISSN: 1742-9994, https://doi.org/10. 848 1186/1742-9994-12-S1-S19. 849 Kuznetsova A, PB Brockhoff, and RHB Christensen (2017). ImerTest Package: Tests in Linear Mixed Effects 850 Models. Journal of Statistical Software 82.13, 1–26. https://doi.org/10.18637/jss.v082.i13. 851 Lee VE and A Thornton (2021). Animal Cognition in an Urbanised World. en. Frontiers in Ecology and Evolution 852 9, 633947. ISSN: 2296-701X. https://doi.org/10.3389/fevo.2021.633947. 853 Lenth RV (2023). emmeans: Estimated Marginal Means, aka Least-Squares Means. 854 Lüdecke D, MS Ben-Shachar, I Patil, P Waggoner, and D Makowski (2021), performance: An R Package for 855 Assessment, Comparison and Testing of Statistical Models. Journal of Open Source Software 6.60, 3139. 856 https://doi.org/10.21105/joss.03139. 857 MacLean EL, B Hare, CL Nunn, E Addessi, F Amici, RC Anderson, F Aureli, JM Baker, AE Bania, AM Barnard, NJ 858 Boogert, EM Brannon, EE Bray, J Bray, LJN Brent, JM Burkart, J Call, JF Cantlon, LG Cheke, NS Clayton, MM 859 Delgado, LJ DiVincenti, K Fujita, E Herrmann, C Hiramatsu, LF Jacobs, KE Jordan, JR Laude, KL Leimgruber, 860 EJE Messer, AC de A. Moura, L Ostoji, A Picard, ML Platt, JM Plotnik, F Range, SM Reader, RB Reddy, AA 861 Sandel, LR Santos, K Schumann, AM Seed, KB Sewall, RC Shaw, KE Slocombe, Y Su, A Takimoto, J Tan, R 862 Tao, CP van Schaik, Z Viranyi, E Visalberghi, JC Wade, A Watanabe, J Widness, JK Young, TR Zentall, and 863 Y Zhao (2014). The evolution of self-control, en. Proceedings of the National Academy of Sciences 111.20. 864

E2140-E2148. ISSN: 0027-8424, 1091-6490. https://doi.org/10.1073/pnas.1323533111.

McHugh ML (2012). Interrater reliability: the kappa statistic. en. *Biochemia Medica*, 276–282. ISSN: 18467482.

https://doi.org/10.11613/BM.2012.031.

Miller DB (1974). A comparative analysis of detour behavior in ring doves (Streptopelia risoria) and rock doves 868 (Columbia livia) as a function of barrier density. PhD Thesis. University of Miami. Miller DB and RB Tallarico (1974). On the correlation of brain size and problem-solving behavior of ring doves 870 and pigeons. Brain, Behavior and Evolution 10, 265-273. https://doi.org/https://doi.org/10.1159/ 871 000124318. 872 Müller W, | Vergauwen, and M Eens (2008). Yolk testosterone, postnatal growth and song in male canaries. en. 873 Hormones and Behavior 54.1, 125–133. ISSN: 0018506X. https://doi.org/10.1016/j.yhbeh.2008.02.005. 874 Navarro D (2015). Learning statistics with R: A tutorial for psychology students and other beginners. (Version 0.6). 875 R package version 0.5.1. University of New South Wales, Sydney, Australia. 876 Nebel C, P Sumasgutner, A Pajot, and A Amar (2019). Response time of an avian prey to a simulated hawk 877 attack is slower in darker conditions, but is independent of hawk colour morph. en. Royal Society Open 878 Science 6.8, 190677. ISSN: 2054-5703. https://doi.org/10.1098/rsos.190677. 879 Pani P, F Giarrocco, M Giamundo, R Montanari, E Brunamonti, and S Ferraina (2018). Visual salience of the 880 stop signal affects the neuronal dynamics of controlled inhibition. en. Scientific Reports 8.1, 14265. ISSN: 881 2045-2322. https://doi.org/10.1038/s41598-018-32669-8. 882 Pinheiro JC and DM Bates (2000). Mixed-Effects Models in S and S-PLUS. New York: Springer. https://doi.org/10. 883 1007/b98882. 884 Press C, D Yon, and C Heyes (2022). Building better theories. en. Current Biology 32.1, R13–R17. ISSN: 09609822. 885 https://doi.org/10.1016/j.cub.2021.11.027. 886 R Core Team (2021). R: A Language and Environment for Statistical Computing. Vienna, Austria: R Foundation for 887 Statistical Computing. 888 Regolin L, V Giorgio, and Z Mario (1995). Object and spatial representations in detour problems by chicks. en. 889 Animal Behaviour 49.1, 195–199. ISSN: 00033472. https://doi.org/10.1016/0003-3472(95)80167-7. 890 Regolin L, G Vallortigara, and M Zanforlin (1994). Perceptual and motivational aspects of detour behaviour in 891 young chicks. Animal Behaviour 47, 123–131. https://doi.org/https://doi.org/10.1006/anbe.1994.1014. 892 Salinas E and TR Stanford (2013). The Countermanding Task Revisited: Fast Stimulus Detection Is a Key De-893 terminant of Psychophysical Performance. en. Journal of Neuroscience 33.13, 5668–5685. ISSN: 0270-6474, 894 1529-2401. https://doi.org/10.1523/INEUROSCI.3977-12.2013. 895 Tacutu R, D Thornton, E Johnson, A Budovsky, D Barardo, T Craig, E Diana, G Lehmann, D Toren, J Wang, VE Fraifeld, and JP de Magalhães (2018). Human Ageing Genomic Resources: new and updated databases. en. 897 Nucleic Acids Research 46.D1, D1083–D1090. ISSN: 0305-1048, 1362-4962. https://doi.org/10.1093/nar/ 898 gkx1042. 899 Thygesen UH, CM Albertsen, CW Berg, K Kristensen, and A Nielsen (2017). Validation of ecological state space 900 models using the Laplace approximation. Environmental and Ecological Statistics, 1–23. https://doi.org/10. 901 1007/s10651-017-0372-4. 902 Troisi CA, A Vernouillet, R Allaert, S Knoch, A Martel, L Lens, and F Verbruggen (2024). Manuscript in prepara-903 tion. in prep. 904 Tvardíková K and R Fuchs (2012). Tits recognize the potential dangers of predators and harmless birds in 905 feeder experiments. en. Journal of Ethology 30.1, 157-165. ISSN: 0289-0771, 1439-5444. https://doi.org/10. 906 1007/s10164-011-0310-0. 907 Ueno A and K Suzuki (2014). Comparison of learning ability and memory retention in altricial (Bengalese finch, 908 Lonchura striata var. domestica ) and precocial (blue-breasted quail, Coturnix chinensis ) birds using a color 909 discrimination task: Visual Learning and Retention in Birds. en. Animal Science Journal 85.2, 186–192. ISSN: 910 13443941. https://doi.org/10.1111/asi.12092. 911 Van Der Schoot M, R Licht, TM Horsley, and IA Sergeant (2005). Effects of stop signal modality, stop signal in-912 tensity and tracking method on inhibitory performance as determined by use of the stop signal paradigm: 913 Factors affecting inhibitory performance in the stop task. en. Scandinavian Journal of Psychology 46.4, 331-914 341. ISSN: 00365564. https://doi.org/10.1111/j.1467-9450.2005.00463.x. 915

- Van Horik J, CE Beardsworth, PR Laker, EJG Langley, MA Whiteside, and JR Madden (2019). Unpredictable envi-
- ronments enhance inhibitory control in pheasants. en. *Animal Cognition* 22.6, 1105–1114. ISSN: 1435-9448,
   1435-9456. https://doi.org/10.1007/s10071-019-01302-0.
- Van Horik J, CE Beardsworth, PR Laker, MA Whiteside, and JR Madden (2020). Response learning confounds
   assays of inhibitory control on detour tasks. en. *Animal Cognition* 23.1, 215–225. ISSN: 1435-9448, 1435 9456. https://doi.org/10.1007/s10071-019-01330-w.
- Van Horik J, EJG Langley, MA Whiteside, PR Laker, CE Beardsworth, and JR Madden (2018). Do detour tasks provide accurate assays of inhibitory control? en. *Proceedings of the Royal Society B: Biological Sciences* 285.1875,
- 20180150. ISSN: 0962-8452, 1471-2954. https://doi.org/10.1098/rspb.2018.0150.
- Verbruggen F, M Best, WA Bowditch, T Stevens, and IPL McLaren (2014). The inhibitory control reflex. *Neuropsychologia* 65, 263–278. https://doi.org/10.1016/j.neuropsychologia.2014.08.014.
- Verbruggen F and GD Logan (2008a). Automatic and controlled response inhibition: associative learning in
   the go/no-go and stop-signal paradigms. *Journal of Experimental Psychology: General* 137.4, 649–72. ISSN:
- 929 0096-3445. https://doi.org/10.1037/a0013170.
- (2008b). Response inhibition in the stop-signal paradigm. en. *Trends in Cognitive Sciences* 12.11, 418–424.
   ISSN: 13646613. https://doi.org/10.1016/j.tics.2008.07.005.
- 932 (2017). Control in Response Inhibition. en. In: The Wiley Handbook of Cognitive Control. Ed. by Egner T.
- Chichester, UK: John Wiley & Sons, Ltd, pp. 97–110. ISBN: 978-1-118-92049-7 978-1-118-92054-1. https:
   //doi.org/10.1002/9781118920497.ch6.
- <sup>935</sup> Verbruggen F, IPL McLaren, and CD Chambers (2014). Banishing the Control Homunculi in Studies of Action
- Control and Behavior Change. en. *Perspectives on Psychological Science* 9.5, 497–524. ISSN: 1745-6916, 1745 6924. https://doi.org/10.1177/1745691614526414.
- Verbruggen F, T Stevens, and CD Chambers (2014). Proactive and reactive stopping when distracted: An atten-
- tional account. en. *Journal of Experimental Psychology: Human Perception and Performance* 40.4, 1295–1300.
   ISSN: 1939-1277, 0096-1523. https://doi.org/10.1037/a0036542.
- Vernouillet A, J Anderson, D Clary, and DM Kelly (2016). Inhibition in Clark's nutcrackers (Nucifraga columbiana):
- results of a detour-reaching test. en. *Animal Cognition* 19.3, 661–665. ISSN: 1435-9448, 1435-9456. https:
   //doi.org/10.1007/s10071-016-0952-y.
- Willcox K, A Vernouillet, L Lens, and F Verbruggen (2024). Early-Life Group Size Does Not Influence Japanese
- Quails' Learning in a Response Inhibition Task. *bioRxiv*. https://doi.org/10.1101/2024.06.19.599677. eprint:
   https://www.biorxiv.org/content/early/2024/06/19/2024.06.19.599677.full.pdf.
- Yeasmin T and M Howlider (2013). Effects of autosomal dwarf gene on growth and shank length of chicken. en. *Bangladesh Veterinarian* 30.1, 25–32. ISSN: 1012-5949. https://doi.org/10.3329/bvet.v30i1.16282.
- Zucca P, F Antonelli, and G Vallortigara (2005). Detour behaviour in three species of birds: quails (Coturnix
- sp.), herring gulls (Larus cachinnans) and canaries (Serinus canaria). en. *Animal Cognition* 8.2, 122–128.
   ISSN: 1435-9448, 1435-9456. https://doi.org/10.1007/s10071-004-0243-x.
- <sup>952</sup> Zucca P and VA Sovrano (2008). Animal lateralization and social recognition: Quails use their left visual hemi-
- field when approaching a companion and their right visual hemifield when approaching a stranger. en.
- <sup>954</sup> *Cortex* 44.1, 13–20. ISSN: 00109452. https://doi.org/10.1016/j.cortex.2006.01.002.