Naïve individuals promote collective exploration in homing pigeons

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Abstract. Group-living animals that rely on stable foraging or migratory routes can develop behavioural traditions to pass route information down to inexperienced individuals. Striking a balance between exploitation of social information and exploration for better alternatives is essential to prevent the spread of maladaptive traditions. We investigated this balance during cumulative route development in the homing pigeon *Columba livia*. We quantified information transfer within pairs of birds in a transmission-chain experiment and determined how birds with different levels of experience contributed to the exploration–exploitation trade-off. Newly introduced naïve individuals were initially more likely to initiate exploration than experienced birds, but the pair soon settled into a pattern of alternating leadership with both birds contributing equally. Experimental pairs showed an oscillating pattern of exploration over generations that might facilitate the discovery of more efficient routes. Our results introduce a new perspective on the roles of leadership and information pooling in the context of collective learning.

Introduction

The coordinated motion of groups is a widespread phenomenon observed in multiple taxa (Vicsek and Zafeiris 2012). Among other adaptive advantages, such as increased energetic efficiency and decreased odds of predation (Krause and Ruxton 2002), collective motion also allows group members to increase their sensory and cognitive capacity (Berdahl et al. 2013; Gelblum et al. 2020) and to acquire valuable social information for navigation (Couzin 2009; Couzin et al. 2011). In many animals, this social information concerns well-established foraging or migratory routes that can, in some species, persist over successive generations (Helfman and Schultz 1984; Sasaki and Biro 2017; Jesmer et al. 2018). Knowledge and skills that accumulate over generations can provide groups with an enhanced ability to solve difficult problems

(Biro, Sasaki, and Portugal 2016). Not only can later generations build on the success of earlier ones, but
the introduction of new members, even those with no prior knowledge, adds diversity that can enhance the
group's behavioural solutions (Mehlhorn et al. 2015). As is often the case (Hills et al. 2015), behavioural
patterns that lead to a search for improvement, whether individually, socially, or over multiple generations,
involve an exploration–exploitation trade-off. In navigation problems, both solitary individuals and groups
have to balance between exploiting previously acquired information necessary to navigate a known route
and exploring for additional information that might allow them to approach the optimal route (Fu and Gray
2006). However, how moving collectives compromise between these tasks has received limited attention.

Understanding the exploration–exploitation trade-off is complicated by ambiguity about group
leadership (Couzin et al. 2005; Garland et al. 2018). Although some collectives (e.g., ants, honeybees) can
allocate certain individuals to spatial exploration while others continue to exploit accumulated information
(Hills et al. 2015), individuals in cohesively moving groups are highly coupled and can only benefit from
compromising between exploring and exploiting if they do so in unison. For example, dominant guineafowls
displace subordinates to monopolize a foraging patch (i.e., exploitation) but, to benefit from the safety of
group cohesion, are then forced to follow subordinates in their exploration for alternative patches
(Papageorgiou and Farine 2020). Baboons can also compromise between movement decisions. When they
travel together, they follow one member's directional preference over another if the angle of disagreement
between conflicting preffered directions is large but compromise by averaging alternative proposed
directions when this angle is small (Strandburg-Peshkin et al. 2015). If individuals are to stay together, the
group must reach consensus between following a known route or departing from it to find better routes,
foraging patches, or temporary resting locations. Elucidating whether different group members contribute
differently to this process is crucial to understanding how groups compromise between exploration and
exploitation.

We investigate this question in the context of navigation through natural landscapes using the homing
pigeon *Columba livia* as our model system. After successive homing journeys from a given release site,
pigeons develop stable idiosyncratic routes that are followed with high fidelity (Meade, Biro, and Guilford
2005; Guilford and Biro 2014). These birds rely on sequences of localized visual landmarks to recapitulate
familiar yet individually distinct routes (Biro, Meade, and Guilford 2004; Meade, Biro, and Guilford 2005).
Each route is learned in a gradual process starting with an exploration phase that samples new landmarks,
from which the bird eventually converges upon a stable sequence of landmarks (Biro, Meade, and Guilford
2004). Experiments with paired birds show that route information can be passed from experienced birds to
naïve individuals through social learning (Pettit, Flack, et al. 2013) and can be modified through information
pooling when individuals with different idiosyncratic routes share information to reach a compromise
between their routes (Biro et al. 2006). Although learning generally improves route efficiency, both social
learning and information pooling tend to reach a plateau beyond which further improvement in efficiency is
not seen. However, for birds flying together, the introduction of a naive individual in place of an experienced
one effectively leads to the resumption of exploratory behaviour and further route improvement (Sasaki and

Biro 2017). Yet, it remains unclear to what extent a bird's prior experience influences the balance between
exploration and exploitation and how birds with potentially different route preferences jointly shape a route.
Indeed, the mechanisms underlying how different individual preferences are combined into a collective
outcome is one of the key foci in studies of collective animal behaviour. Broadly, group decisions can range
from despotic with a single leader to democratic in which input from different individuals is aggregated to
reach consensus (Conradt and Roper 2003). Some animal groups make both despotic and democratic
decisions, and researchers have been investigating what determines reliance on one collective decision-
making strategy over the other (King and Cowlishaw 2009). For example, baboons live in despotic societies
where the alpha male is most often responsible for group decisions between alternative foraging
destinations (King et al. 2008), but they can also decide democratically in certain situations, such as during
daily ranging activities (Strandburg-Peshkin et al. 2015). Evidence of both despotic and democratic
decisions also exists in homing pigeons (Biro et al. 2006; Nagy et al. 2010; Jorge and Marques 2012).
When leadership is defined as disproportionate input into collective navigational decisions, either through
spatial position (Pettit, Perna, et al. 2013), route similarity (Flack et al. 2012), or directional correlation delay
(Nagy et al. 2010), a number of different factors have been shown to play a role in it. Leadership dynamics
are influenced by individual differences among birds in fidelity to their own routes (Freeman et al. 2011),
their typical flight speed (Pettit et al. 2015), their personality (Sasaki et al. 2018), as well as their level of
experience (Flack et al. 2012). Moreover, equally experienced birds are known to come to a compromise
by averaging their idiosyncratic routes so long as the pair's route remains within a threshold distance from
each bird's favoured one – a low level of conflict. Higher levels of conflict lead instead to a splitting of the
pair or to the emergence of a single leader (Biro et al. 2006). Nonetheless, experience alone is unable to
fully recover the leadership structure characteristic of larger flocks (Watts et al. 2016). Spatial position offers
some insight into leadership; on average, birds flying closer to the front of the flock have a stronger influence
on the flock's directional choices than birds flying at the back (Nagy et al. 2010; Pettit, Perna, et al. 2013).
Even so, the moment-to-moment relationship between leadership and level of experience remains unclear.

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Figure 1. Illustration of the methodological approach. The spatial trajectories of an experienced (E) and a naïve (N) bird
(point 1) are encoded as clockwise and counter-clockwise rotations (point 2) which we represent as discrete time series
(point 3). The combination of rotations encoded in both series (point 4) is used to estimate the probabilities required to
compute transfer entropy (point 5) and to determine the influence of one individual over the future behavior of the other

(point 6). This example illustrates transfer entropy from experienced to naïve, but we also computed it for the opposite
direction.

Leader–follower interactions of this sort can be accurately captured using information-theoretic
measures that quantify causal relations in terms of predictive information (Butail, Mwaffo, and Porfiri 2016;
Kim et al. 2018; Crosato et al. 2018; Ray et al. 2019; Valentini et al. 2020). This methodological approach,
which generally requires large amounts of data (but see (Porfiri and Ruiz Marín 2020)), is gaining popularity
among behavioural ecologists (Strandburg-Peshkin et al. 2018; Pilkiewicz et al. 2020) as tools for automatic
monitoring and extraction of the necessary volumes of behavioural data become increasingly available
(Egnor and Branson 2016). One of these measures, *transfer entropy,* quantifies information about the future
behaviour of a focal individual that can be obtained exclusively from knowledge of the present behaviour of
another subject (Schreiber 2000). Transfer entropy measures information transferred from the present of
the sender to the future of the receiver (Lizier and Prokopenko 2010). It explicitly accounts for
autocorrelations characteristic of individual birds' trajectories (Mitchell et al. 2019) by discounting predictive
information available from the sender's present that is already included in the receiver's past (see Figure
1). Furthermore, it does not require a model of how sender and receiver interact, and it is well suited to
study social interactions both over space and time (Lizier, Prokopenko, and Zomaya 2008; Strandburg-
Peshkin et al. 2018). This aspect of transfer entropy encompasses traditional methods to quantify collective
movement that are based on modelling an individual's behaviour as a combination of three motional
tendencies (Couzin et al. 2002) – alignment of direction to nearby group members, attraction towards
sufficiently distant members, and repulsion from sufficiently close members – that allow an individual to
maintain proximity to the group. In this context, transfer entropy is advantageous as it can capture causal
interactions due not only to alignment forces (Nagy et al. 2010) but also to attraction and repulsion forces
that result in temporarily unaligned states (Pettit, Perna, et al. 2013).

We study collective decision making and the exploration–exploitation trade-off using an experimental
analysis of cumulative route development in homing pigeons conducted by Sasaki and Biro (2017). In their
experiments, pairs consisting of a naïve and an experienced bird were required to successively solve the
same homing task a total of 12 times. This set of paired flights, which represents a single generation of
cumulative route development, allowed the naïve bird to acquire knowledge of localized visual landmarks
necessary for homing. At the end of each generation, the more experienced bird was then replaced with a
new naïve individual and the learning process was repeated with the newly formed pair. This transmission-
chain design, where experienced individuals were repeatedly replaced with naïve ones, lasted five
generations and was replicated in ten independent transmission chains. Route efficiency was measured as
the ratio of the beeline distance between the release site and the home loft (i.e., the ideal optimum) and the
actual distance travelled by birds. Sasaki and Biro's (2017) results showed that, although homing efficiency
dropped considerably every time a new naïve bird was introduced, transmission-chain pairs continued to
improve within and over generations, eventually outperforming both solo and fixed-pair controls
(respectively, 0.92 efficiency versus 0.83 and 0.85). In contrast, the efficiency of solo and fixed pairs

plateaued after they had first established their idiosyncratic routes (around the 9th–10th release for the
former and the 7th–8th release for the latter).

The continued improvement seen in transmission chains might result from a variety of decision-making
mechanisms ranging from fully despotic to increasingly democratic. A simplified perspective of this
continuum allows us to consider four alternative hypotheses. In two of these alternatives, H_1 and H_2, a
single despotic leader, either the naïve (H_1) or the experienced (H_2) bird, determines the entire homing
route. Whereas evidence of social learning (Sasaki and Biro 2017) suffices to dismiss the possibility of
leadership by the naïve individual (H_1), leadership by the experienced individual (H_2) could still be the only
process in place if social learning is unidirectional and the naïve individual merely triggers the experienced
bird to resume and lead exploration. Under the other two hypotheses, H_3 and H_4, birds pool their personal
information by means of democratic processes based on moment-by-moment integration of individual
preferences or transient, alternating leadership (Conradt 2012). The third hypothesis (H_3) entails the
experienced bird contributing only its past route information and relying instead on the naïve individual for
the discovery of route innovations. If this hypothesis holds, we expect the naïve bird to disproportionally
lead phases of exploration. Otherwise (fourth hypothesis, H_4), both experienced and naïve birds might
contribute through exploration to the discovery of new information.

We discriminated between these alternative hypotheses by using transfer entropy to reveal the extent
to which birds influence each other and to investigate if relative spatial position can accurately predict
leader–follower dynamics. On this basis, we studied the contribution of each bird to the exploration–
exploitation trade-off over different stages of route development. This exploration–exploitation perspective
of homing route development allowed us to characterize the efficiency of choices made by birds over the
course of the experiment and to shed light on the superior performance of experimental pairs with respect
to solo and fixed pairs controls.

Results

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Birds pool information. Whereas previous evidence of social-learning (Sasaki and Biro 2017) suffices to
dismiss the possibility of naïve individuals behaving in a despotic manner (H_1), the despotic approach
remains a possible option for experienced birds (H_2,). Indeed, the social-learning hypothesis under which
the naïve bird passively copies the idiosyncratic route of the experienced one (i.e., H_2, the despotic leader)
entails a transfer of information that is unidirectional – from the experienced to the naïve bird. Instead, under
the two alternative hypotheses based on democratic decision-making (H_3 and H_4), the two birds rely on
bidirectional information transfer to pool information and increase the efficiency of their route (Pettit, Flack,
et al. 2013; Sasaki and Biro 2017). We rejected the unidirectional social-learning hypothesis H_2 by finding
causal evidence of information pooling; the naïve bird actively influenced the behaviour of the experienced
one for a large portion of the parameter space (Figure S1). As is common practice with these measures
(Porfiri 2018), we selected the parameter configuration that maximized the total transfer of information
between the two birds (one sample every 0.2 seconds, history length of 10 samples). This was maximal for

the shortest sampling period (i.e., prediction interval) of 0.2 seconds and progressively decreased towards
0 for larger periods up to 4 seconds, indicating that the effect of an interaction between birds was transitory
and lasted for a limited period of time. Using this configuration, we compared measurements of information
transfer against those of a surrogate dataset created by pairing trajectories of birds that were not flown
together. We found that levels of mutual influence between birds that flew together were significantly higher
than those observed in the surrogate dataset both overall and for each generation separately (Mann–
Whitney–Wilcoxon, columns 2 and 3 of Table S1).

(a)   (b)

Figure 2. Panel (a) shows the net predictive power of the two birds over generations; it measures the excess predictive
information within the pair and highlights which of the two bird is more informative (purple for the experienced bird,
green for the naïve bird). The red line corresponds to a linear fit over generations using the Theil–Sen estimator. Panel
(b) shows the predictive power of naïve and experienced birds separately from each other as a function of generations.
The predictive power of a bird with respect to the other is measured in terms of the percentage reduction in uncertainty
and it has been computed on the basis of transfer entropy as detailed in Materials and Methods.

During the first two generations of paired flights (Figure 2a, paired analysis), when there was a large margin
to improve the efficiency of the pair's trajectory, the naïve bird was more informative than the experienced
one, evidenced by a stronger influence of the former over the latter. At generation 4, there was a balance
between the two birds, whereas the experienced bird eventually became the better source of predictive
information in the last generation. A linear fit over generations of the paired comparison (Figure 2a, red line)
showed an increasing influence of the experienced bird over the naïve one (Theil–Sen estimator, slope
0.534, $p < 0.001$). Additionally, a non-paired comparison of the same results revealed that, although the
behaviour of the naïve bird resulted, on average, in a marginally higher predictive power than that of the
experienced bird (18.7–21.4% versus 17.7–20.5%), variations in each bird's route explained a large portion
of the other bird's behaviour (Figure 2b) suggesting non-trivial leadership dynamics. These results do not
show whether different levels of experience within the pair led to asymmetric contributions of birds to route
development, with the experienced bird providing only its past route information and the naïve bird in charge
of discovering route innovations, or if both birds contributed to the exploration for possible route alternatives.

To discriminate between the remaining hypotheses H_3 and H_4, we first developed the means to evaluate

leadership on a moment-to-moment basis.

(a) **(b)**

Figure 3. Panel (a) shows sample flight trajectories for a number of different releases, r, of the same pair of birds.
Colours highlight which bird is ahead of the other during different segments of the route. Panel (b) shows the local
transfer of information (mean and 95% confidence interval) between the experienced bird and the naïve one as a
function of their relative distance (colours represent the direction of information transfer) estimated from generations
2–5 using smoothed conditional means.

Relative position determines temporary leadership. Consistent with information sharing within each

pair, we found that experienced and naïve birds repeatedly switched their positions at the front and back of

the pair (Figure 3a). Previous studies found evidence that birds that spent, on average, more time at the

front of the flock had a tendency to assume leadership roles (Nagy et al. 2010; Pettit, Perna, et al. 2013).

To see whether this average relationship between leadership and position holds at each point in time, we

investigated the spatiotemporal dynamics of information transfer (Lizier, Prokopenko, and Zomaya 2008).

We did so by considering the amount of predictive information obtained by each bird as a function of the

distance from the experienced to the naïve bird projected over their mean direction of motion (Figure 3b).

We found that within a distance of up to 30 meters, the bird flying ahead was consistently more informative

than that flying in the back. This is not only further evidence that the bird flying ahead acts as the leader,

influencing the path of the follower behind it, but, because of its finer grain, it also enables relative distance

between birds to be used as a (more parsimonious) moment-to-moment measure of causal influence within

the pair.

The experienced and the naïve bird alternated leading segments of the route, where leadership

durations were consistent with a log-normal distribution (see Supplementary Material). Although the naïve

bird flew at the front of the pair for longer segments (Whitney–Mann–Wilcoxon, $p < .047, W = 56865638$),

the difference was very small (0.3 seconds) and largely driven by the flights of one generation. Indeed, for

all generations but the third ($p < .001, W = 6588674$), the distribution of consecutive time spent at the front

of the pair by the experienced bird cannot be distinguished from that of the naïve individual (Table S2). The

tails of these distributions approach that of an exponential distribution and suggest that temporary leadership might be decided randomly (Biro et al. 2006) instead of using deterministic rules such as fixed periods of time. Moreover, with the exception of generation 3 where 54% of the route was led by the naïve bird (Wilcoxon signed rank-test, $p = .03, V = 1851$), there was no significant difference in the proportion of a flight spent by each bird at the front of the pair (Table S3) suggesting a relatively egalitarian relation between birds despite differing levels of experience.

(a) (b) (c)

Figure 4. Panel (a) shows the proportion of exploration over releases for the experimental group (the red dotted vertical line separates solo flights at generation 1 from paired flights at generations 2-5), the solo control, and the fixed-pairs control. Smoothed lines are computed using generalized additive models using shrinkage cubic regression splines (mean and standard error); points represent averages for individual releases. Panel (b) shows the proportion of a flight led by each bird during phases of exploration and exploitation in the experimental treatment. Darker colours correspond to exploitation, lighter colours to exploration; purple represents the experienced bird, green the naïve one; red lines represent linear fits to data pooled from both birds using the Theil–Sen estimator (slopes and p-values: respectively, 0.0203, $p < 0.001$ for exploitation and -0.0171, $p = 0$ for exploration). Panel (c) shows the probability for the naïve individual to initiate phases of exploration over releases (exact binomial test, $p < .01$ for release 1). The dotted vertical line separates the first release, where the naïve bird was significantly more likely to initiate, from the successive releases 2–12 showing no significant difference between the two birds; the red line provides a visual reference indicating an equal likelihood between the two birds to initiate exploration.

Exploration–exploitation dynamics explain flight performance. Sasaki and Biro (2017) previously showed that flight efficiency varies across treatments with experimental pairs eventually outperforming both fixed pairs of birds and solo individuals. The discovery of route innovations and, in particular, how birds with different levels of experience contribute to this task, is the key to the superior performance of experimental pairs. To understand this phenomenon and thus shed light on the pair's information-pooling mechanism, we investigated how pigeons balance between exploitation of known information – closely following (< 300 meters) their most recent route – and exploration for possible route improvements. To do so, we labelled segments of flight trajectories as a function of the point-to-point distance from each point of a focal trajectory to the closest point of the immediately preceding trajectory (i.e., baseline) and compared the exploration–exploitation dynamics both across treatments and between experienced and naïve birds.

During the initial part of the experiment (Figure 4a, first 12 releases), exploration decreased steadily in all conditions with birds that flew individually in the experimental group (i.e., generation 1) performing similarly to those of the solo control (respectively, 36.7% and 34.2%) whereas fixed pairs of birds explored significantly more (51.7%, Whitney–Mann–Wilcoxon, $p < .001$, Table S4). However, while exploration steadily decreased for solo and fixed pairs of birds in the successive 48 releases, experimental pairs

showed a markedly different pattern of exploration oscillating over generations (Figure 4a, releases 13–
60). Each time a new naïve individual was paired with an experienced one (dotted vertical lines), exploration
increased for about 5 to 6 releases, reaching values well beyond those of both solo birds and fixed pairs;
then exploration decreased within a few releases (2 to 4) to the same levels as those of fixed pairs. On
average during generations 2 to 5, experimental pairs explored (32.9%) significantly more than both solo
(15.7%, $p < .001$) and fixed pairs of birds (29.3%, $p = .0456$). These results also held when exploration and
exploitation were defined with respect to the last release of the previous generation. Under this model,
differences between experimental and fixed pairs were even more pronounced, with the former
characterized by 46.6% exploration and the latter by only 32.4% ($p < .001$, Table S5 and Figure S6). The
inferior flight efficiency of solo and fixed pairs of birds might thus be explained, at least in some measure,
by a lower likelihood to discover route improvements due to limited exploration.

The superior homing performance of experimental pairs is suggested to be rooted in their ability to
select novel portions of a route introduced by the naïve individual that are more efficient while discarding
inefficient ones (Sasaki and Biro 2017). Under this hypothesis, which is an extension of H_3, we expect to
observe not only increasing homing efficiency over generations but also an asymmetric pattern of
leadership in which the naïve individual leads periods of exploration and the experienced one leads periods
of exploitation. We found instead no significant difference between the contributions of the experienced bird
and those of the naïve one both overall and within generations (Figure 4b and Table S7). The sole exception
is represented by generation 3 during which the naïve bird contributed more than the experienced one to
exploitation (Wilcoxon signed-rank test, $p = .035, V = 1871$). In agreement with our expectations for H_4,
experienced and naïve birds led the pair with approximately the same frequency in both exploration and
exploitation, suggesting that deviations from established routes were not caused only by the naïve bird (see
also Figure S4, inset). We did find evidence of behavioural asymmetries, in that transitions from exploitation
to exploration were marginally more likely to be initiated by naïve birds (exact binomial test, $p = .042, n =$
964, Table S8); however, this result was driven by those of generation 3 ($p = .02, n = 301$) whereas no
difference was detected in other generations. Transitions from exploration to exploitation were equally likely
to be initiated by the two birds both overall and within each generation. However, when transitions are
considered over the 12 releases composing each generation (Figure 4c), the naïve individual was more
likely to initiate phases of exploration during the first release ($p < .01, n = 44$, Table S9) doing so 70.5% of
the time compared to 29.5% for the experienced bird. After the first release, transitions that initiate phases
of exploration were about as likely to be initiated by either of the two birds independently of their level of
experience.
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Discussion

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For many group-living animals, searching for optimal travel routes can be a complex task as social
information about routes can persist over generations regardless of its quality (Helfman and Schultz 1984;

Sasaki and Biro 2017; Jesmer et al. 2018; Laland and Williams 1998). This search is inherently subject to
a trade-off between the exploitation of well-established route information accumulated over time and
exploration for innovations that constitute potential improvements (Hills et al. 2015). Striking a balance is
fundamental as a pronounced reliance on exploitation of learned information can hinder productive
innovations (Davies, Krebs, and West 2012) and thus promote the maintenance of potentially suboptimal
behaviour and even of maladaptive behavioural traditions (Laland and Williams 1998). Equally, an over-
reliance on exploration without exploiting the rewards of beneficial innovations eventually impedes
improvements in performance over time (Fu and Gray 2006; Mehlhorn et al. 2015).

We studied the causal structure of this process in flights of the homing pigeon *C. livia*, as this species
is capable of both social learning and information pooling (Biro et al. 2006; Pettit, Flack, et al. 2013). Of
particular interest for our study is the increase in route efficiency that results from the pairing of naïve
individuals with experienced ones (Pettit, Flack, et al. 2013), including when this happens iteratively over
multiple generations (Sasaki and Biro 2017). Previous work has proposed information pooling as the
underlying mechanism driving this increase in flight performance (Sasaki and Biro 2017). Using transfer
entropy to measure predictive information (Schreiber 2000; Pilkiewicz et al. 2020), we found quantitative
evidence that supports the information-pooling hypotheses H_3 and H_4 in the strength of causal interactions
within pairs of birds. Experienced and naïve birds influence each other's behaviour; about 20% of the future
directional choices of any individual in a pair is explained by the behaviour of the other individual. These
results contrast with our expectations for unidirectional social learning (H_2) that entails an asymmetric
pattern of leadership with a pronounced role for experienced individuals.

Our analysis showed that, in a multi-generational transmission-chain design, the naïve bird has a higher
influence than the experienced one during the early generations. In later generations, however, the
experienced bird becomes the better source of predictive information. We hypothesize that, over
generations, as birds have explored more search space and thus exhausted many alternative routes, a
newly introduced bird becomes less likely to contribute productive innovations. As a result, innovations can
lead to errors instead of improvements in later generations. Rather than leading to additional exploration
bouts, these erroneous innovations could be suppressed by the contributions of the experienced individual
that, as generations progress, has access to increasingly better information. This theoretical reasoning is
analogous to the diminishing marginal value of returning to a previous location (i.e., continuing to rely on
the naïve bird for innovations) when searching for an object in space (Stone 1976). From an information-
foraging perspective (Stephens and Krebs 1986; Pirolli 2007), the time invested in harvesting innovations
on the introduction of a naïve bird corresponds to the time invested in attending to a newly discovered
patch; at some point, the opportunity cost of further harvesting becomes too high to justify remaining in the
patch. Thus, after the introduction of a naïve bird, the information-foraging pair shifts over successive
releases from information-harvesting exploration back toward information-preserving exploitation, as would
be expected in an optimal search problem (Stone 1976). Over generations, as route information within the

pair becomes better, the balance between exploration for route improvement and exploitation of the known
route changes in favor of the latter.

How do experimental pairs improve their homing routes over generations? Previous studies where
leadership was defined on the basis of route similarity showed that, in pairs with a large difference in
experience between birds, experienced individuals were more likely to assume leadership (Flack et al.
2012). Still using route similarity, Sasaki and Biro (2017) found evidence of social learning with naïve
individuals learning routes from their experienced partners. Moreover, because newly formed pairs in the
transmission-chain design also improved performance generation after generation, they proposed that
naïve individuals could contribute innovations that pigeons evaluate in terms of route efficiency and prune
away when inefficient (i.e., H_3). However, defining leadership in terms of causal interactions instead of route
similarity allowed us to show that there is an asymmetric relation between innovators and exploiters only
during the initial flight of a newly formed pair. Although leadership is ephemeral and equally shared between
birds during exploration and exploitation independently of their level of experience, over the course of this
first flight, the naïve individual disproportionally initiates phases of exploration, attracting the experienced
bird to unfamiliar areas and triggering it to also resume the search. After that, both birds are equally likely
to initiate transitions between exploration and exploitation as expected from our fourth hypothesis H_4.
Moreover, as it is unlikely that experimental pairs were merely better than control groups at evaluating
efficiency, we believe that their superior performance is rooted instead in their complex exploration–
exploitation dynamics that allowed them to better cover the search space.

Personally acquired information allows solo individuals to improve flight performance (Meade, Biro, and
Guilford 2005) but only to discover routes moderately efficient (consistently within 0.8–0.85 efficiency
across a large number of experiments, reviewed in Guilford and Biro 2014) because solo individuals rapidly
reduce their exploration efforts to seek out novel information. Experimental and fixed pairs of birds on the
other hand explore more and thus outperform solo individuals. Together, two birds have superior sensory
and cognitive capacities compared to single birds (Krause, Ruxton, and Krause 2010) which facilitates the
discovery of better routes that are then learned collectively (Biro, Sasaki, and Portugal 2016; Kao et al.
2014). The reasons why pairs explore more than solo individuals might lie partly with the conflicts
characteristic of newly formed pairs (Biro et al. 2006) if the resolution of conflict, e.g., through averaging
individual inputs, indirectly prompts pairs to explore more and discover route innovations. This hypothesis
could also explain why experimental pairs outperform fixed pairs. The introduction of a naïve individual at
the start of each generation repeatedly creates an experience imbalance in the newly formed pair, which
differs from fixed pairs that undergo a process of mutual habituation as they develop a stable route that
likely reduces conflicts. This experience imbalance could be the source of new conflicts, providing a
possible explanation for why experimental pairs reach levels of exploration generally higher than those of
fixed pairs. The process of gradually settling on a joint route over the course of a generation following this
initial perturbation is also reminiscent of the transient effects observed when an ant with outside information

joins a group of nestmates transporting an object; the new information temporarily steers the collective in
the right direction, but its effects on collective motion vanish quickly (Gelblum et al. 2015, 2016, 2020).
The ability of groups to outperform single individuals by pooling information across their members is an
aspect of collective intelligence that has long intrigued researchers. One potential mechanism underlying
this phenomenon, popularly known as the wisdom of crowds (Surowiecki 2005), is averaging many
individuals' estimates independent from each other. Averaging individual decisions is expected to provide
a more accurate group estimate than any individuals' guess. Previous studies have also shown that animals
can average their movement decisions to reach a compromise (Biro et al. 2006; Strandburg-Peshkin et al.
2015). Although the mechanisms by which experienced and naïve individuals pool information during route
development remain unknown, our study points to the importance of naïve group members within the
information-pooling process. Moreover, the wisdom of crowds is known to require personal information to
be independent among group members (Couzin 2018) otherwise group performance can degrade quickly
for increasing group size (Kao and Couzin 2014). Experimental pairs could thus benefit from pooling
information with naïve individuals that, at least at the beginning of each generation, likely provide a source
of information independent from that of the experienced bird. The potentially deleterious effects of losing
independence may provide another pressure to shift over time from innovative exploration to route-
preserving exploitation. It remains to be explored how our results generalize to larger flock sizes. Previous
experiments without generational replacement showed that, even in larger flocks, birds flying ahead of the
flock had a tendency to assume leadership positions (Nagy et al. 2010). However, the repeated introduction
of naïve individuals into larger flocks might complicate the dichotomy between leaders and followers by
inducing turnover dynamics between the front and the back of the flock.
Adopting an explicit exploration–exploitation perspective to study search strategies and doing so
through the use of predictive information to quantify causal interactions has the potential to advance our
understanding of collective navigation in larger flocks. The conceptual framework of exploration and
exploitation as well as the methods we proposed can also benefit researchers studying other taxa that
move in groups with the potential to learn from previous experiences, such as shoaling fish or certain
primates. In principle, this information-theoretic approach could also be applied to the study of information
transfer between the environment and the individuals within a group. The ability to quantify causal
interactions of this sort could shed light on broader questions in ecology involving animals moving in a
group and their environment such as the impact of visual landmarks on navigation or the effects of terrestrial
migration on the environment (Bracis and Mueller 2017; de Guinea et al. 2021).
418

419   **Materials and Methods**
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421   Data and source code are available in (Valentini et al. 2021).
422
Experimental subjects and procedure. Data were taken from a previous study on cumulative route
development in the homing pigeon *Columba livia* (Sasaki and Biro 2017). Pairs of birds composed of an

experienced and a *naïve* bird were released together from the same site and allowed to fly back to the
home loft. Pairs were created and released over 5 successive generations of a transmission chain, each
generation lasting 12 consecutive releases of the same pair, according to the following procedure: initially,
at generation 1, a naïve bird was released alone 12 times, allowing it to develop its own idiosyncratic route
to the home loft. This bird, now experienced, was then paired with a new (naïve) bird at generation 2, and
together they performed another 12 flights. This process was then repeated in the next generation with a
new pair of birds composed of the former naïve bird and of a new naïve one. Data were gathered for a total
of 10 independent transmission chains, each lasting 5 generations (see Sasaki and Biro 2017 for details).
Birds flying at an average linear distance greater than 250 meters from each other were not considered as
pairs and were thus excluded from the analysis, leaving 343 flights with a mean ± SD flight duration of
8.65 ± 1.33 minutes. Additionally, in two control conditions, nine solo birds and six pairs (all initially naïve)
were released from the same site for a total of 60 releases (the equivalent of 5×12 releases for the
transmission chains).
438
Data collection and pre-processing. Flight trajectories of birds were sampled at a frequency of 5 Hz
using GPS loggers, converted from the geographic coordinate system to the metric system, and projected
over the 2-dimensional plane (see Sasaki and Biro 2017). Each trajectory consisted of a time-ordered
series of positions in space, $(x_i : (x^1, x^2)_i, i \geq 1)$, see Figure 1 (point 1). We encoded the pattern of rotations
of each flight using a binary symbolic representation where symbols 0 and 1 represent, respectively, a
clockwise and a *counter-clockwise* rotation (point 2). The direction of rotation was computed as the cross
product $\overrightarrow{x_{i-1}x_i} \times \overrightarrow{x_i x_{i+1}}$ between the motion vector at time i and that at time $i + 1$. The rotation is clockwise
when the product is negative and counter-clockwise when it is positive. We also measured the distance
$d_{EN}(i)$ of the experienced bird from the naïve one projected over the current direction of motion of the pair
(cf. Nagy et al. 2010 and Supplementary Material). Using this distance, we then determined the relative
position of birds over time: when $d_{EN} > 0$, the experienced bird was flying ahead of the naïve bird; when
$d_{EN} < 0$, it was flying behind. Previous tests using the same GPS loggers showed that these devices have
a sufficient level of accuracy with a small normally distributed spatial error (SD of 0.05 meters) affecting
the tracking accuracy of the direction of motion and a relatively larger error (median of 1.69 meters)
affecting that of the relative position (Pettit, Perna, et al. 2013). In our experiments, experienced and naïve
birds flew at an average distance from each other of 8.74 meters with a standard deviation of 38.54 meters.
455
Measuring information transfer. We quantified the amount of information transferred between birds
using information-theoretic measures (Cover and Thomas 2005; Pilkiewicz et al. 2020) estimated from the
series of rotations of the experienced, $E = (e_i, i \geq 1)$, and of the naïve, $N = (n_i, i \geq 1)$, birds (Figure 1,
point 3). We aimed to quantify causal interactions between birds in a Wiener–Granger sense by measuring
the extent to which the current behaviour of one bird allows us to predict the future behaviour of the other
(Bossomaier et al. 2016). Here we describe the process for predicting the naïve bird's behaviour from that

of the experienced one, but we also used the same method for the opposite direction. The average amount

of information necessary to fully predict the next rotation of the naïve bird is quantified by the marginal

entropy of its series of rotations $H(N^{i+1}) = -\sum_{n_i} p(n_i) \log_2 p(n_i)$ (Figure 1, Venn diagram, lower-left set).

This is equal to 1 bit if the flight of the naïve bird is maximally uncertain (i.e., clockwise and counter-

clockwise rotations are equally likely) and to 0 bits if the flight is fully deterministic (i.e., rotations are either

all clockwise or all counter clockwise). As a result of temporal autocorrelation (Mitchell et al. 2019), part of

this information might be contained in the recent history of rotations of the naïve bird, $n_i^{(k)} =$

$\{n_{i-k+1}, \ldots, n_{i-1}, n_i\}$ for history length k (Figure 1, point 4, dark green entries in the naïve time series). The

remaining predictive information, which is not explained by the past behaviour of the naïve bird, is

quantified by the marginal entropy,

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$$H(N^{i+1}|N^{(k)}) = -\sum_{n_i^{(k)}, n_{i+1}} p(n_i^{(k)}, n_{i+1}) \log_2 \frac{p(n_i^{(k)}, n_{i+1})}{p(n_i^{(k)})},$$

of its future rotations, N^{i+1}, conditioned on the outcome of the past k rotations, $N^{(k)}$ (Figure 1, Venn

diagram, overlap of the white and light purple areas).

Of interest to us was how much of this remaining information (necessary to predict the future

direction of rotation of the naïve bird) can be obtained by the current behaviour of the experienced bird

(Figure 1, points 4–6). This is given by the transfer entropy, which estimates the time-delayed effects on

the naïve bird of its interaction with the experienced one: $T_{E \to N} = H(N^{i+1}|N^{(k)}) - H(N^{i+1}|N^{(k)}, E)$

(Schreiber 2000). Transfer entropy is time directional, from the present of one bird to the future of the other,

and considered for this reason a measure of information transfer (Lizier and Prokopenko 2010). It is defined

as

482
$$T_{E \to N} = \sum_{n_{i+1}, n_i^{(k)}, e_i} p(n_{i+1}, n_i^{(k)}, e_i) \log_2 \frac{p(n_{i+1} \mid n_i^{(k)}, e_i)}{p(n_{i+1} \mid n_i^{(k)})}$$

and formally measures the reduction of uncertainty of the future rotation N^{i+1} of the naïve bird given by

knowledge of the current rotation E of the experienced bird at the net of possible autocorrelations in the

naïve bird's past $N^{(k)}$ (Figure 1, Venn diagram, area with red border). The logarithmic part of the above

equation is known as local transfer entropy and measures over time whether the interaction at time i

(Figure 1, point 4, dark green and dark purple entries) was informative (positive) or misinformative

(negative) (Lizier, Prokopenko, and Zomaya 2008).

The above information-theoretic measures were computed in R 3.6.1 using the rinform-1.0.2

package (Moore et al. 2018) by estimating probabilities separately for each flight. To test whether causal

interactions were significant, we also evaluated a surrogate dataset artificially created by pairing

trajectories of birds that were not flown together: the trajectory of each experienced (or naïve) bird was

paired with that of the naïve (or experienced) one from every other pair of birds not containing the same

subjects.

Measuring exploration and exploitation. For each point in a focal trajectory, we determined the point-to-
point distance to a baseline trajectory as the minimum distance from the focal point to its closest point in
the baseline trajectory. Using this measure, pairwise analysis of successive routes from the last three flights
of the experienced bird during training (generation 1) showed that, once established, the bird largely
remained within a point-to-point distance of 300 meters from its idiosyncratic route (Figure S3a). Therefore,
we used 300 meters as a threshold to group points from a focal route into flight segments differentiated
between those exploring new solutions and those exploiting known ones. We then compared the
trajectories of consecutive flights for the same subjects to label each segment in the experimental and
control datasets. For each focal trajectory that we aimed to label, we considered the trajectory of the
previous release as a baseline trajectory for the comparison. In the case of paired birds (i.e., both
experimental and fixed-pairs control), we considered the trajectory of the pair defined by the mean position
of the two birds over time. For the first release of each generation in the experimental group, we used as
baseline trajectory the last release of the previous generation because in this case there was no previous
flight of the same pair to compare with. This approach to define exploration and exploitation led to a model
of exploitation (i.e., the baseline trajectory) that varied over successive releases. Because the introduction
of a naïve bird at each generation was likely to affect the baseline model of exploitation in a more
pronounced manner than that of solo and fixed pairs of birds, this model might have been susceptible to
differences between the experimental design of transmission chain experiments with respect to those of
the two controls. To control for this scenario, we also explored an alternative approach where exploitation
was defined on the basis of only information available to the experienced bird at the beginning of a new
generation. In this case, the last release of the previous generation was used as the baseline trajectory for
all releases within a given generation (see Supplementary Methods). In both cases, we also measured the
distance $d_{EN}(i)$ between experienced and naïve birds to determine which bird was flying at the front of the
pair for a given route segment.
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521 **References**
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Acknowledgements

704 GV, TPP, SIW, and SCP were supported by NSF grant No. PHY-1505048 awarded to SIW, TPP, and SCP.
705 GV was also supported by research funds from Arizona State University to Prof. Bert Hölldobler. DB was
706 supported by grant No. TWCF0316 from the Templeton World Charity Foundation's "Diverse Intelligences"
707 scheme. The authors would like to thank Dr. Albert B. Kao and Prof. Andrew Berdahl for helpful discussions.
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Author contributions statement

711 GV and TS designed the study with input from all authors. TS and DB provided data. GV analysed data
712 with feedback from all authors. All authors contributed to the writing of the manuscript.
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Additional information

716 **Accession codes:** Data and supplementary material that support the findings of this study will be available
717 in "*figshare*" with identifier "10.6084/m9.figshare.14043362" upon acceptance of the manuscript.
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719 **Competing interests:** The authors declare no competing interest.