Steering clear of the confusion effect: aerial predators target fixed points in dense prey aggregations

Caroline H. Brighton†1, Laura N. Kloeper†2,3, Lucy Larkman†1, Lillias Zusi†2, Kathryn McGowan†2, Graham K. Taylor†1

†1Department of Zoology, University of Oxford, Mansfield Road, Oxford, OX1 3SZ, UK
†2Department of Biological Sciences and Center for Acoustics Research and Education, Spaulding Hall, University of New Hampshire, Durham, NH 03824, USA
†3Department of Biology, Saint Mary’s College, 262 Science Hall, Notre Dame, IN 46556, USA

Summary

Collective behaviours are widely assumed to confuse predators, but empirical support for a confusion effect is often lacking, and its importance must depend on the predator’s targeting mechanism. Here we show that raptors steer attacks on swarming bats by turning towards a fixed point in the swarm, rather than by using closed-loop pursuit of an individual. Any prey that the predator is on a collision course with will appear on a constant bearing, so target selection emerges naturally from the geometry of a collision. Our results show how predators can simplify the demands on their sensory system by decoupling steering from targeting when grabbing prey from a dense swarm. We anticipate that the same tactic will be used against flocks and schools across a wide range of taxa, in which case a confusion effect is paradoxically more likely to occur in attacks on sparse groups, for which steering and targeting cannot be decoupled.

Main

Dense aggregations of flocking, schooling, and swarming animals are widely assumed to confuse predators, but evidence for a confusion effect is far from universal1. Other mechanisms reducing the predation risk of grouped prey depend closely on predator behaviour. Prey individuals will only benefit from attack abatement2-4 if a predator’s attack rate scales less than proportionally with group size, which depends on its search strategy and feeding efficiency. Theories of marginal predation5-7 and the selfish herd8-10 are likewise underpinned by the predator’s tendency to attack nearby individuals11, and even the shared benefits of group vigilance12 may depend on its mode of attack13. The occurrence and significance of any confusion effect must similarly depend on the predator’s attack behaviour1. For example, it is reasonable to expect confusion to arise in raptorial predators that single out individuals, rather than in engulfment predators that target the group, but the detail of how an attacker homes in on its target is presumably decisive. Here we analyse the visually guided flight behaviour of hawks attacking swarming bats emerging in massive column formation from their roost (Movie S1), using three-dimensional video reconstruction techniques and behavioural algorithms to model their behaviour.

Confusion is supposed to occur when the presence of many targets compromises a predator’s ability to target an individual14, particularly if those targets are similar in appearance or moving in a coordinated fashion. Yet, empirical support is often lacking – partly because of the difficulty of
designing experiments that eliminate other confounding factors. Bats emerging from their roosts often form massive and highly dynamic swarms that present many sensory challenges to their predators (Fig. 1). For instance, although the procession of the swarm may be quite coherent, individual movements are often highly erratic (Fig. 1A) producing a tangle of interwoven flight paths (Fig. 1B). Such protean behaviour makes individuals particularly difficult to track, and may be expected to amplify the confusion effect in a dense group. Nevertheless, a previous observational study of the model system that we consider here found that Swainson’s Hawks Buteo swainsoni hunting Mexican Free-tailed Bats Tadarida brasiliensis in broad daylight (Movie S1) were no less successful when attacking bats flying within the column rather than bats flying apart from it. Despite the lack of evidence for any confusion effect, the bats’ predation risk was still an order of magnitude lower when they were flying within the column, however, as a result either of overall attack abatement or the tendency of predators to attack marginal individuals. But how were the hawks able to avoid confusion, and what are the broader implications for our understanding of its effects in attacks on dense prey aggregations?

Attacks on solitary prey require accurate guidance. Open-loop interception may be possible against a predictably moving target, for example when bow-hunting small game, but closed-loop steering is essential when pursuing an evasive target. The quantitative mapping from the relative motion of the target to the steering response of an attacker is called a guidance law. Fitting guidance laws to empirical data has provided insight into the behavioural algorithms that different predators use to catch lone prey. For example, the terminal attack trajectories of Peregrines Falco peregrinus and Gyrfalcons F. rusticolus are well modelled by a guidance law called proportional navigation (PN)\(^{22,23}\). This commands the attacker’s turn rate \(\dot{\gamma}\) (i.e. the rate of change of the bearing \(\gamma\) of its velocity) in direct proportion to the target’s line-of-sight rate \(\dot{\lambda}\) (i.e. the rate of change of the bearing \(\lambda\) to its target). This PN guidance law, \(\dot{\gamma}(t) = N\dot{\lambda}(t - \tau)\) where \(N > 0\) is the navigation constant and \(\tau \geq 0\) is a time delay, is most effective against non-manoeuvring or smoothly manoeuvring targets. In contrast, the pursuit trajectories of Harris’ Hawks Parabuteo unicinctus are best modelled by augmenting PN with a proportional pursuit (PP) term that commands turning at a rate proportional to the deviation angle \(\delta\) between the attacker’s velocity vector and its target\(^{21}\). This mixed PN+PP guidance law, \(\dot{\gamma}(t) = N\dot{\lambda}(t - \tau) - K\delta(t - \tau)\) where \(K > 0\) is another guidance constant, outperforms PN against erratically manoeuvring targets. Although we have written these guidance laws in a scalar form applicable to planar pursuit for ease of interpretation, they are each readily generalized to three-dimensional (3D) motion by writing them in vector form and are analysed as such here (see Methods).

Previous algorithmic studies of pursuit-evasion have only analysed attacks on solitary targets, and it remains unknown what guidance laws – if any – predators use when attacking aggregated prey. In principle, open-loop interception could be more reliable against dense prey aggregations, owing to the higher probability of fluking a hit when multiple targets are present. This could in turn have important implications for the occurrence and significance of any confusion effect, and hence for the functional mechanisms promoting collective behaviour. Here we test whether the behavioural algorithms that aerial predators use to attack dense swarms are the same as those used to attack lone prey. We go on to show how the different behavioural algorithms that are used when attacking dense swarms may avoid the confusion effect that has been widely assumed to be one of the key mechanisms by which individual prey benefit from engaging in collective behaviour.
Figure 1. Swarming behaviour of Mexican Free-tailed Bats during an attack by a Swainson’s Hawk. (A) Instantaneous flight direction (yellow arrows) of bats (black points) tracked during an attack by an incoming hawk (white circle), as seen from a fixed video camera. Not all bats are tracked, but the inset polar plot summarises the distribution of the projected flight directions from (A), with the black arrow indicating the circular mean. The bats’ motion is coherent in the sense that the swarm moves consistently from left to right, but displays significant variability reflecting protean behaviour within the swarm. (B) Flight trajectories of 20 bats (yellow points) leading up to the grab in the frame shown; flight trajectories are sampled at 0.02 s spacing, with the last 5 points highlighted in cyan. The trajectory of the targeted bat (red circle) grabbed by the hawk (white circle) is shown in magenta. Note that the multiple interwoven flight paths make it exceedingly difficult to track individuals – even with the benefit of being able to move back and forth through a sequence of video frames. Tracking is even more challenging in real time, so the bats’ protean behaviour could be expected to amplify the confusion effect arising from having many similar individuals visible simultaneously.
Results

We recorded attacks over 21 days, using three pairs of high-definition video cameras fixed in stereo configuration around the cave from which the bats emerged (Movie S1). Tracking the hawks and the individual bats that they grabbed (Fig. 1) allowed us to reconstruct their 3D attack trajectories. We reconstructed $n = 62$ terminal attack trajectories in this way, and another $n = 26$ long-range approaches in which we were able to track the attacking hawk but not the bat that it grabbed. We also reconstructed $n = 2$ long-range approaches recorded for a Peregrine or Prairie Falcon $F. mexicanus$ that attacked the swarm on two consecutive evenings. We analysed these flights by finding the values of the guidance constants $N$ and/or $K$ that minimized the mean absolute distance ($\eta$) between the observed and simulated data for each flight, in numerical simulations commanding steering according to the vector form of the mixed PN+PP guidance law $\dot{y}(t) = N\dot{\lambda}(t) - K\dot{\delta}(t - \tau)$, where we set $K = 0$ or $N = 0$ to fit pure PN or PP respectively (see Methods). Other things being equal, longer simulations have greater potential to diverge, and we therefore report the relative error $\varepsilon = \eta/l$, where $l$ denotes the total path length.

Terminal attack trajectories provide no evidence of closed-loop pursuit within the swarm

We began by fitting the three candidate guidance laws to the $n = 62$ terminal attack trajectories that we recorded from the hawks, under the assumption that they were tracking the bat they grabbed without delay, at $\tau = 0$ (Table S1). The median guidance parameter estimates for $N$ and $K$ were positive as expected under pure PN or PP guidance ($\bar{N} = 1.6$, CI: 1.1, 2.0; $\bar{K} = 0.7$, CI: 0.1, 1.7 $s^{-1}$; Fig. 2A,B), but the parameter estimates for $K$ were inconsistently signed under mixed PN+PP guidance ($\bar{N} = 1.8$, CI: 1.2, 2.6; $\bar{K} = 0.4$, CI: -0.3, 0.7 $s^{-1}$; Fig. 2C,D), which suggests that they were overfitted. Of the two pure guidance laws, PN fitted the data more closely than PP (two-tailed sign test: $p = 0.007$; $n = 62$; Fig. 2E,F). We obtained qualitatively similar results with delays $0 < \tau \leq 0.1$ s but found that the error was minimized at $\tau = 0$. We conclude that PN is the best supported of the three candidate guidance laws, conditional upon the assumption that the hawks steered after whichever bat they grabbed. We tested this assumption by asking whether the bat’s trajectory prior to the grab influenced the hawk’s steering but found no evidence that this was so. On the contrary, delay-free PN simulations treating the bat’s final position as the target of the hawk’s guidance ($\bar{N} = 1.7$, CI: 1.5, 2.1) fitted the data more closely (two-tailed sign test: $p < 0.001$; $n = 62$; Fig. 2E) and with lower median relative error ($\bar{\varepsilon} = 0.011$; CI: 0.009, 0.013) than those targeting its instantaneous position ($\bar{\varepsilon} = 0.021$; CI: 0.017, 0.027). We therefore find no evidence that the hawks engaged in closed-loop pursuit of the bat that they grabbed. As a check on the robustness of this conclusion, we applied an analogous model selection approach to published data on Peregrines and Gyrfalcons pursuing singleton targets,$^{22,23}$ which confirmed as expected that their attack trajectories were better modelled by delay-free PN targeting the instantaneous rather than final position of the target (Table S2). As this is the opposite of what we observed for hawks attacking swarming targets here, it is reasonable to test whether the hawks’ observed turns are instead consistent with the use of some form of open-loop steering behaviour.

Terminal attack trajectories approximate constant radius turning into the swarm

The simplest possible model of turning involves the hawk making a constant radius turn into the swarm. This behaviour can be simulated using the same algorithmic approach as before, because
Figure 2. Summary of guidance models fitted independently to \( n = 62 \) terminal attack trajectories of Swainson’s Hawks attacking Mexican Free-tailed Bats. Violin plots showing kernel density estimates with data points overlain, comparing (A-D) the fitted guidance parameters \( N \) and/or \( K \) and (E-F) relative error \( \varepsilon \) for the three alternative target definitions (full: instantaneous bat position; bp: final bat position; hp: final hawk position) and three candidate guidance laws: (A,E) PN, (B,F) PP, and (C,D) PN+PP. White circle denotes median; thick grey bar represents interquartile range; thin black line indicates outlying points falling >1.5 times the interquartile range below the first quartile or above the third quartile. Parameter estimates for \( N \) and \( K \) did not always converge on reasonable values for attacks involving little turning; panels (A-D) do not display some datapoints with higher values of \( N \) and \( K \), but all datapoints are used to compute the kernel density estimates and the median and interquartile range.
delay-free PN has the property of generating a constant radius turn towards a stationary target if the navigation constant is fixed at $N = 2$. Our finding that the hawks’ terminal attack trajectories were well modelled by PN targeting the bat’s final position at median $\hat{N} = 1.7$ (CI: 1.5, 2.1) could therefore be an artefact of constant-radius turning. We tested this interpretation by fitting delay-free PN targeting the final position of the hawk rather than the final position of the bat, which yielded navigation constants even closer to the theoretical value of $N = 2$ producing a constant radius turn ($\hat{N} = 1.9$, CI: 1.8, 2.0; Fig. 2A). As a final check, we fixed the navigation constant at $N = 2$ in delay-free PN treating the final position of the hawk as a virtual target (Fig. 3). The resulting simulation model has no free parameters, but still modelled the hawks’ terminal attack trajectories more closely (two-tailed sign test: $p < 0.001$; $n = 62$) and with lower median relative error ($\bar{e} = 0.003$; CI: 0.001, 0.008) than the best of the fitted guidance models targeting either the bat’s trajectory or its final position (Fig. 2E). It follows that there is no evidence for the use of closed-loop pursuit over even the simplest possible model of open-loop steering behaviour matching the hawk’s initial flight velocity and subsequent flight speed. In summary, our data provide no evidence for the use of closed-loop pursuit, and are more compatible with the hypothesis that the hawks turned on an approximately constant radius into the swarm, before grabbing at whichever bat they found themselves on a collision course with.

Grabbed bats remain on a near-constant bearing during the terminal attack trajectories

A striking feature of the hawks’ terminal attack trajectories is that their line-of-sight to the bat remains nearly parallel throughout (Fig. 3). This constant bearing geometry holds for any pair of objects on a collision course, so is inevitable during the final moments of an attack, but it could nevertheless simplify the problem of singling out a bat to grab (see Discussion). To verify whether this constant bearing geometry was particular to the grabbed bats, we therefore plotted the azimuth and elevation of their line-of-sight for all $n = 62$ terminal attack trajectories (Fig. 4A,B), together with those of 20 neighbouring bats tracked concurrently over each of four attacks on which the bats were scattered enough to be individually identified across camera views (Fig. 4B,C). These polar plots display a radial pattern demonstrating the near-constant azimuth and elevation of the line-of-sight for the grabbed bats (Fig. 4A,B), but show a much less radial pattern for other neighbouring bats (Fig. 4C,D). The circular standard deviation of the line-of-sight angle was significantly smaller in azimuth (Wilcoxon rank sum test: two-tailed $p = 0.008$) for the $n = 62$ grabbed bats (median: 0.19°; Q1, Q3: 0.11, 0.24°) than for the $n = 80$ neighbouring bats (median: 0.28°; Q1, Q3: 0.13, 0.37°). The circular standard deviation of the line-of-sight angle did not differ significantly between groups in elevation (Wilcoxon rank sum test: two-tailed $p = 0.57$), being small for both the grabbed individuals (median: 0.13°; Q1, Q3: 0.08, 0.18°) and the neighbouring bats (median: 0.15°; Q1, Q3: 0.06, 0.21°). This constant bearing geometry was therefore a reliable feature of the grabbed bats, which could potentially be used by the hawk to distinguish them from neighbouring bats during the terminal phase of the attacks (see Discussion).

Long-range approaches approximate constant radius turning into the swarm

The terminal attack trajectories that we modelled were necessarily quite short (median path length: $\bar{l} = 6.7$ m; Q1, Q3: 2.8, 11.8 m), as they were limited to the brief interval over which it was possible for us to track the bat that the hawk grabbed (median duration: 0.6 s; Q1, Q3: 0.3, 1.0 s). To test the range of distances over which the hawks’ steering behaviour could be well
Figure 3. Terminal attack trajectories of Swainson’s Hawks successfully attacking Mexican Freetailed Bats. (A-O) Each panel plots the reconstructed three-dimensional trajectory of a hawk (dark blue points) capturing a bat (dark grey points), connected by the instantaneous line-of-sight from the hawk to the bat (blue line), for all $n = 15$ attacks ending in a successful capture. The line-of-sight remains nearly parallel through time, because the hawk and bat are on a near-collision course (see Discussion; Figs. 4, 6). Orange lines plot simulations of the hawk’s flight trajectory generated under delay-free PN guidance at a fixed navigation constant of $N = 2$, treating the final position of the hawk as the target. This serves to generate a constant radius turn that satisfies the kinematic constraint of passing through the hawk’s initial and final positions whilst also matching its initial flight velocity. Flights involving steep dives or climbs are not usually well modelled as a constant radius turn (e.g. C, I, N), but other flights are modelled quite closely. Grid spacing: 1 m.
Figure 4. Constancy of line-of-sight angle during the terminal attack trajectories of Swainson’s Hawks attacking Mexican Free-tailed Bats. (A,B) Polar plots displaying the azimuth and elevation of the line-of-sight (LOS) from the hawk to the bat it grabbed for all \( n = 62 \) attacks; the radial coordinate represents the bat’s range. The plotted lines are nearly radial, indicating that the grabbed bat remains on a nearly constant bearing to the hawk. (C,D) Polar plots displaying LOS azimuth and elevation for 20 neighbouring bats tracked concurrently on each of four colour-coded attacks; darker lines display the LOS azimuth and elevation of the bat that was grabbed. These darker lines are more nearly radial than the lighter lines displaying the LOS azimuth and elevation of the neighbouring bats, confirming that the constant bearing geometry is specific to the grabbed bat. Only one of these four flights (purple) ended in a successful capture, and has the most nearly radial of the dark lines – this being the flight on which the collision conditions are most closely met.
approximated as a constant radius turn, we therefore simulated delay-free PN targeting the hawk’s final position at a fixed value of $N = 2$ for the 26 long-range approaches (Fig. S1). We ran these simulations beginning from different points in the flight, from 1.0 s up to a maximum of 20.0 s before the grab, in 0.2 s intervals. We identified the longest simulation for which the relative error was $\varepsilon \leq 0.012$, this being the same error tolerance used in previous studies. We found that we could model 20 of the $n = 26$ long-range approaches under delay-free PN at $N = 2$ to within this 1.2% error tolerance, with a median path length of 24.5 m (Q1, Q3: 11.9, 45.6 m) and a median duration of 2.1 s (Q1, Q3: 1.4, 3.9 s). For comparison, we also applied the same modelling approach to the two long-range approaches that we recorded from a Peregrine or Prairie Falcon. Simulating delay-free PN guidance at a fixed value of $N = 2$, we found that we could model both falcon trajectories at 1.2% error tolerance over distances of 48.4 and 70.6 m and durations of 4.0 and 4.4 s (Fig. S2). It follows that most of the long-range attacks that we observed could be well approximated as constant radius turns initiated well before the hawk had entered the swarm.

Proportional navigation is a plausible mechanism for targeting a fixed point in the swarm

The foregoing simulations of PN guidance against a virtual target at $N = 2$ are not necessarily intended to provide a literal model of the hawks’ underlying guidance mechanism. Rather, they provide an algorithm generating the unique constant radius turn that satisfies the kinematic constraint of passing through the hawk’s initial and final positions whilst also matching its initial flight velocity. It can be shown theoretically that the resulting flight path is a curve of radius $(2 \sin \delta_0)/r_0$ where $\delta_0$ is the initial deviation angle and $r_0$ the initial distance to the target. It follows that whilst it is possible in principle for an attacker to calculate the radius of turn required to reach a given point in space, doing so requires a range estimate that may be lacking here given the lack of evidence for stereopsis in diurnal raptors. We are therefore left with two alternative hypotheses: first, that the birds enter a turn of approximately constant radius, without knowing where they will intercept the swarm; second, that the birds use PN guidance at $N \approx 2$ to target a fixed point in space, without knowing what radius of turn they will follow. The first hypothesis describes a form of open-loop steering behaviour, whereas the second describes a form of closed-loop guidance.

Distinguishing between these alternatives is challenging, but as a check on the validity of the second hypothesis, we tried modelling the long-range approach flights using delay-free PN at the best-fitting value of $N$ for each flight, treating the bird’s final position as a virtual target (Fig. 5). This allowed us to model 24 of the $n = 26$ hawk approaches to within the same 1.2% relative error tolerance as before, with a median path length of 28.8 m (Q1, Q3: 14.3, 43.8 m), and a median duration of 2.5 s (Q1, Q3: 2.0, 4.6 s). The median guidance parameter estimate in these models ($\tilde{N} = 1.9$, CI: 1.6, 2.4) was statistically indistinguishable from the theoretical value of $N = 2$ corresponding to a constant radius turn. Hence, although there is a risk of over-fitting associated with estimating the best-fitting value of $N$ independently for each flight (Fig. 5), allowing this to vary between flights extends the total distance of flight modelled at 1.2% error tolerance by just over one fifth for the hawks, and by more than a half for the falcons (Fig. S2). In summary, the long-range approaches that we observed are consistent with the hypothesis that the birds steered open-loop into the swarm on a turn of approximately constant radius, but are also consistent with the hypothesis that they used PN guidance to steer closed-loop towards a fixed point in the swarm at a navigation constant of $N \approx 2$.  


Figure 5. Long-range approaches of Swainson’s Hawks attacking Mexican Free-tailed Bats modelled under delay-free PN targeting a fixed point in the swarm. (A–O) Each panel plots the reconstructed three-dimensional attack trajectory of an incoming hawk (dark blue points); dark blue lines are dropped vertically from each point to accurately convey the three-dimensional shape of the trajectory. It was not possible to track the bats that the hawks grabbed at this range, but orange lines plot simulations of the hawk’s flight trajectory generated under delay-free PN guidance at the best-fitting value of the navigation constant $N$, assuming flight at the same speed as measured and treating the final position of the hawk as the target. Trajectories are plotted for the longest section of flight for which the relative error remains below the threshold value of $\varepsilon \leq 0.012$, displaying the subset of 15 flights with the longest simulations meeting this criterion. Grid spacing: 10 m.
Hawks display no evidence of confusion and no evidence of preferentially targeting lone bats

Of the $n = 62$ terminal attack trajectories that we recorded from Swainson’s Hawks, 8 involved attacks on lone bats flying outside of the column. This proportion (13%; CI: 6, 24%) is consistent with the results of an earlier study of the same model system using other data\textsuperscript{19}, which found that lone bats were attacked at disproportionately high frequency (10%; CI: 7, 15%) compared to their overall representation in the population (~ 0.2%). Moreover, as in the previous study\textsuperscript{19}, we found no significant difference in the success rates of attacks on lone bats versus attacks on the swarm (Fisher’s exact test: odds ratio: 2.1; CI: 0.4, 10.1; $p = 0.39$; $n = 62$), at observed success rates of 38% and 22% respectively (Table S3). This earlier study left unresolved whether the higher attack rates observed against lone bats resulted from the hawks having an individual preference for targeting lone bats, or from a statistical tendency for the hawks to encounter marginal individuals first\textsuperscript{19}. Our finding that the hawks showed no evidence of steering after the bats they grabbed strongly suggests that the hawks were not targeting individual bats at all. Indeed, even for the subset of attacks made on lone bats, the hawks’ trajectories were more closely fitted by PN targeting the hawk’s final position at a fixed value of $N = 2$ than the bat’s instantaneous position (sign test: $p = 0.008$; $n = 8$). Hence, although we cannot exclude the possibility that the hawks computed an open-loop collision course to enable them to intercept these lone bats, it is more parsimonious to suppose that they encountered them at random, and that the higher predation risk experienced by lone bats simply reflects the fact that marginal individuals are likely to be encountered first when turning into the swarm.

Discussion

A bat is an agile target that cannot be picked out easily within a swarm. However, whereas the question of how predators guide their attacks on singleton targets has been the subject of several algorithmic studies\textsuperscript{21-23,26}, the question of how they guide their attacks on agile swarming targets has not been addressed previously. Applying the same algorithmic approach to natural behaviour\textsuperscript{20}, we find no evidence that hawks attacking swarming bats use closed-loop pursuit of an individual bat, in contrast to raptors attacking singleton targets. Instead, the hawks appear to turn into the swarm, then to grab whichever bat they find themselves on a collision course with as they close range. This could be explained either as an open-loop steering behaviour that involves making an approximately constant radius turn (Figs. 3, S2), or as a closed-loop steering behaviour that involves targeting a spatially fixed point within the swarm (Fig. 5). We cannot distinguish reliably between these hypotheses – not least, because steering under closed-loop PN guidance at $N = 2$ will also generate a constant radius turn (Fig. 3, S2). Nevertheless, the general hypothesis that the hawks used closed-loop steering towards a fixed point in the swarm explains more of the data (Fig. 5), and assumes only that whatever guidance mechanism is used to pursue individual targets can also be applied to a virtual target defined as a spatially fixed point within the swarm. This, in effect, is an Eulerian rather than Lagrangian approach to targeting, and it is an approach that works only in the context of a suitably dense swarm.

This may in turn explain why neither our present study nor our previous study of the same model system\textsuperscript{19} found any evidence to suggest that attacks on the swarm were less successful than attacks on lone bats. This runs counter to the usual assumption that predator hunting efficiency declines
Figure 6. Mechanism by which the constant bearing geometry of an incidental collision course can avoid a confusion effect in a dense swarm. (A) Composite image showing the motion of a hawk and the bat that it grabs from the swarm, as seen from a fixed video camera. The trajectories of the hawk (blue points) and the bat that it ultimately grabs (black points) are sampled at 0.02 s intervals and are shown connected by the instantaneous line-of-sight from the hawk to the bat (white lines). The composite images of the hawk are sampled from the video at 0.2 s intervals and overlain on the frame corresponding to the grab; the red starburst shows the actual point of capture. Note that the line-of-sight remains approximately parallel as the hawk plunges into the swarm, which is because the hawk is on a collision course with the bat that it grabs. (B) Projected flight velocity (yellow arrows) of the bats tracked in Fig. 1A (black points), as seen by a stationary observer. The projected flight velocity of the highlighted bat (red arrow) is used as a reference in the next panel. (C) Apparent velocity (cyan arrows) of the same bats, as seen by a moving observer whose translational motion makes the apparent velocity of the highlighted bat (red circle) zero, such that it appears stationary against the distant background. Provided that its range is decreasing simultaneously, this constant bearing geometry means that the observer must be on a collision course with the highlighted bat. Hence, whereas (B) represents the optic flow seen by a stationary observer, (C) represents the optic flow seen by a moving observer that happens to be on a collision course with the highlighted bat. The constant bearing geometry of an incidental collision course thereby simplifies the sensory challenge of identifying candidate targets in attacks on dense swarms.
as group density increases, owing to the heightened sensory challenge of targeting individual prey\(^7\). On the contrary, it is this very density that makes it probable that any hawk swooping into the swarm will find itself on a collision course with a bat. This is an appealing interpretation from a sensory perspective, because candidate targets will appear to remain on a constant bearing to an incoming attacker (Fig. 4A, B). The geometric reason for this can be seen by treating the line-of-sight as one side of a curved triangle whose two other sides are formed by the tracks of the predator and prey. If the two individuals are on a direct collision course, then the collision triangle remains geometrically similar through time, such that the line-of-sight remains parallel (Fig. 3). This constant bearing geometry (Fig. 4) is often taken as evidence of PN guidance\(^{27,28}\) or motion camouflage\(^{29-31}\) in dyad interactions, so it is important to emphasize that it holds for any pair of objects that happen to be on a collision course. For instance, the same constant bearing geometry would arise fortuitously for a bat hit by an arrow shot at random into the swarm. Conversely, objects that are not on course for a collision with each other will usually appear to change their bearing, so drivers learn intuitively to adjust their speed when merging onto a busy freeway, so that the vehicles in the lane they are entering appear on an advancing or receding bearing.

A similar heuristic could simplify the sensory challenge of selecting a target from within a swarm (Fig. 6). For example, if the hawks were to stabilize their gaze inertially against the distant background, then the retinal position of any candidate target would remain approximately constant, whereas other bats would appear to move against the background. The visual confusion that we experience as stationary observers when viewing a swarm is therefore at least partly resolved for a mobile predator by the constant-bearing geometry of a collision (Fig. 6A). Specifically, whereas all the members of a swarm may appear to move coherently from the viewpoint of a stationary observer (Fig. 6B), any member that is on a collision course with a mobile observer will appear stationary against the residual movement of the swarm (Fig. 6C). It follows that the motion of a swarm will look fundamentally different to a moving observer (Fig. 6C) than to a stationary one (Fig. 6B), with these differences aiding in target selection and capture. In conclusion, our results show that the confusion effect may be less important to predators attacking dense swarms than has previously been assumed\(^1\), and may also explain why some visually-guided predators, including raptors attacking bats\(^32\), preferentially target larger or denser groups of prey. This is because plunging into a dense swarm makes it highly probable that an attacker will find itself on a collision course with a prey individual, and the geometry of a collision means that candidate targets will fall on a constant bearing, making them comparatively straightforward to identify and grab. Collective behaviours that appear confusing from a human standpoint – from swarming bats\(^19\) to murmuring birds\(^33,34\), schooling fish\(^35,36\), and herding ungulates\(^37\) need not be so confusing after all to predators taking the plunge.

### Methods

#### Study system

We conducted our research at a remote cave site on private land in the Chihuahuan Desert, Sierra County, New Mexico, USA from 8 to 29 June 2018. The surrounding area is an elevated volcanic plateau at approximately 1500 m altitude, with the remains of collapsed lava tubes forming a deep canyon with cave and arch features. The site was chosen because of the presence of a population
of Swainson’s Hawks (*Buteo swainsoni*) that predate the population of Mexican Free-tailed Bats (*Tadarida brasiliensis*) that emerge from the caves *en masse* daily throughout the summer. The bats migrate to the site during their breeding season from May to September, and use the caves as a day roost before flying to their feeding grounds at dusk. The population consists of a maternal colony of approximately 700,000 to 900,000 bats which inhabit two connected caves named North and South. The largest and most reliable emergence was from the South cave, occurring every evening without exception. Emergence from the North cave was less reliable, with no bats emerging at all on some nights during the first week of observations. The numbers of bats were topped up in the second week by new arrivals, and emergence from the North cave was reliable thereafter. Emergence began at a variable time between approximately 18:30 and 20:00 MDT and lasted from 10 to 25 minutes depending on the number of bats emerging. Sunset was between 20:16 and 20:21 MDT, so the bats usually emerged in broad daylight. During the third week of observations, a substantial second emergence usually occurred at each cave, beginning around 0.5 h after the end of the first emergence, when fewer hawks were present. No ethical issues were identified by the Local Ethical Review Committee of the University of Oxford’s Department of Zoology. We attended only as observers, and never entered the caves, so the risk of causing disturbance as the bats emerged was low.

**Video observations**

We recorded video of the hawks attacking the bats every evening from 8 to 29 June 2018, except for one evening that had to be missed due to bad weather. We used three pairs of high-definition video cameras (Lumix DMC-FZ1000/2500, Panasonic Corporation, Osaka, Japan) to enable reconstruction of the three-dimensional flight trajectories of the hawks and bats, setting the camera lens to its widest zoom setting. We recorded 25 Hz video at 3840×2160 pixels for the first three days and 50 Hz video at 1920×1080 pixels for the remainder of the study (Movie S1). This higher frame rate proved necessary to facilitate tracking of the bats’ erratic movements but was traded off against lower spatial resolution. Each camera pair was set in widely spaced stereo configuration to enable three-dimensional reconstruction of the attacks, with a baseline distance of 16 to 27 m. The cameras were mounted on tripods which were adjusted to the same height using an optical level kit (GOL20D/BT160/GR500, Robert Bosch GmbH, Gerlingen, Germany). We used the same optical level kit to measure the baseline distance between the cameras.

We set up two camera pairs facing approximately north and south across the South cave for the duration of the study. As the swarm’s overall flight direction was variable and influenced by the wind, we positioned the north- and south-facing camera pairs to allow them to be panned from northeast to northwest and from southeast to southwest, respectively. This enabled us to cover most flight directions, except due east (where the bats rarely flew) and due west (which was subject to glare). We set up a third camera pair to view the emergence that occurred from the North cave from the second week onward. When leaving the North cave, the bats usually flew along the lava tube and beneath a rock arch before climbing out of the canyon. We therefore positioned the cameras close to where the swarm began climbing out above the canyon rim, aiming to capture attacks as the hawks swooped low over the canyon.

The hawks consistently appeared within a few minutes of the start of emergence, which enabled us to observe the general direction in which the bat swarm was emerging, and to reorient the
cameras to view the swarm before the attacks began. As soon as the bats began emerging, the cameras were turned on and left to record. To begin with, all fieldworkers retreated into make-shift hides, but these were gradually phased out for reasons of practicality. The birds quickly became habituated to our presence, venturing close to the cave even when fieldworkers were present. Each attack began with the hawk approaching the swarm in level flight or stooping in from above. This was followed by fast flight through the stream of bats, with one or more attempts made to grab a bat using a pitch-up, pitch-down, or rolling grab manoeuvre with the legs and talons extended (Movie S1). If the first attack was unsuccessful, then the hawks would usually perform further short-range swoops through the stream until they made a catch. Once a bat was caught, the hawk would drift away from the swarm, to consume its prey on the wing.

**Videogrammetry**

We synchronised the videos using the DLTdv5 video tracking toolbox in MATLAB (MathWorks Inc., Natick, MA). To do so, we matched the complex motions involved in the hawks’ attack manoeuvres visually between videos, and applied the relevant frame offset to synchronise them to the nearest frame (i.e. to within ± 0.01 s at the primary 50 Hz frame rate). We then used the DLTdv5 toolbox to identify the pixel coordinates of the hawk in both videos within a pair, manually tracking the visual centre of the subject’s body from the point at which it appeared in both cameras up to the point of interception. We used the same method to track the bat that the hawk grabbed or attempted to grab during the terminal attack sequences that we recorded at close range. The bats were too distant to be tracked individually in recordings of the hawks’ long-range approaches, but the point of actual or attempted capture was nevertheless obvious from the hawks’ flight behaviour.

We calibrated the cameras by matching 15 points across both frames, including background features and points on the hawk, which we selected with the objective of covering as much of the capture volume as possible. The image coordinates of these calibration points were exported from the DLTdv5 toolbox into custom-written software in MATLAB, which solved the camera collinearity equations using a nonlinear least squares bundle adjustment. The bundle adjustment routine identifies jointly optimal estimates of the camera calibration parameters and unknown spatial coordinates of the calibration points, by minimizing the sum of the squared reprojection error of the associated image points. The reprojection error of an image point matched across camera views is simply the difference between its measured image coordinates and those expected under the camera calibration model given its estimated spatial coordinates. This nonlinear approach enabled us to self-calibrate the cameras using identified features of the environment, whilst also incorporating prior knowledge of the intrinsic and extrinsic camera parameters. This in turn avoided the need to move a known calibration object through the very large imaging volume.

We set the calibrated baseline distance between the cameras equal to the measurement that we made of this in the field using the optical level. We fixed the focal length of each camera at 1468.9 pixels for the 1920×1080 recordings and at 3918.5 pixels for the 3840×2160 recordings. These values were estimated using the Camera Calibrator toolbox in MATLAB, from a set of 20 calibration images of a checkerboard pattern held in front of the camera. Lens distortions were found to be minimal, and we therefore assumed a central perspective projection in which we assumed no lens distortion and no principal point offset with respect to the camera sensor. The
resulting stereo camera calibration was used to solve for the spatial coordinates of the tracked hawk and bat in MATLAB. This is a least squares solution, in the sense that it minimizes the sum of the squared reprojection error for each image point matched across stereo video frames. We therefore report the root mean square (RMS) reprojection error as a check on the accuracy of the calibrations and reconstructions.

For the terminal attack trajectories filmed at close range, the mean RMS reprojection error of the 16 calibrations was $0.73 \pm 0.35$ pixels, whilst for the reconstructed flight trajectories it was $1.22 \pm 1.18$ pixels for the hawks and $1.87 \pm 2.39$ pixels for the bats over all $n = 62$ flights (mean ± SD). For the long-range approaches filmed at a distance, the RMS reprojection error of the 18 calibrations was $0.53 \pm 0.61$ pixels, whilst for the reconstructed flight trajectories it was $1.08 \pm 1.07$ pixels for the hawks over all $n = 28$ flights (mean ± SD). The sub-pixel reprojection error that we achieved in the calibrations is appropriate to the method. The higher reprojection error of the reconstructions is also to be expected, because whereas the bundle adjustment optimizes the camera calibration parameters jointly with the estimated spatial coordinates of the calibration points, the calibration is held fixed in the reconstructions. In addition, any spatiotemporal error in the matching of points across camera frames will manifest itself as reprojection error in the reconstructions.

The foregoing calibration reconstructs the spatial coordinates of the matched image points in a Cartesian coordinate system aligned with the sensor axes of one of the cameras. To aid visualization and interpretation of the flight trajectories, we therefore transformed the spatial coordinates of the hawks and bats into an Earth axis system in which the $z$-axis was vertical. To do so, we filmed and reconstructed the ballistic trajectory of a small rock thrown high into the air through the volume of stereo overlap. We identified the image coordinates of the peak of its parabolic path, together with the image coordinates of two flanking points located ± 20 or 25 frames to either side. We took the line dropped from the peak of the parabola perpendicular to the line connecting the two flanking points to define the direction of gravitational acceleration. We then used this to identify the rotation needed to transform the spatial coordinates of the hawks and bats into Earth axes with the $z$-axis as vertical. Finally, we made use of the fact that the two cameras in each pair were fixed at the same height to verify the transformation to Earth axes. For the 16 calibrations used to reconstruct the terminal attack trajectories, the inclination of the baseline between the cameras in Earth axes had a median absolute value of just $1.2^\circ$ (1st, 3rd quartiles: $0.8^\circ$, $2.2^\circ$), providing independent validation of the calibration method that we used.

**Trajectory analysis**

All trajectory analysis was done using custom-written software in MATLAB. We used piecewise cubic Hermite interpolation of the reconstructed trajectories to estimate the spatial coordinates of the hawk or bat for any occasional frames in which this was obscured. We then smoothed the trajectories using quintic spline fitting. For the long-range approaches, we used a spline tolerance designed to remove an RMS spatial position error of 0.5 m, corresponding approximately to the wing length of a hawk. For the terminal attack trajectories, we used a tolerance designed to remove an RMS position error of 0.12 m, corresponding approximately to the wing length of a bat. These values were chosen as representative estimates of the accuracy with which it was possible to match points across frames at long and close range, respectively. Finally, we differentiated and evaluated
the splines analytically to estimate the velocity and acceleration of the bird and bat at an up-sampled frequency of 2 kHz. This ensured a suitably small integration step size for the subsequent numerical simulations. On average, the hawks flew faster than the bats (Fig. S3A), so were tracked over longer distances (Fig. S3B), but with considerable overlap in their respective distributions.

We simulated the hawk’s attack trajectory in the Earth axes using a guidance law of the form:

\[
a(t) = N\omega(t - \tau) \times v(t) - K\delta(t - \tau) \times v(t)
\]

where \(a\) is the hawk’s commanded centripetal acceleration, \(v\) its velocity, \(\omega\) the angular velocity of the line-of-sight \(r\) from the hawk to its target, and \(\delta\) the deviation angle between \(r\) and \(v\), written in vector form with \(\delta\) mutually perpendicular to \(r\) and \(v\). Here, \(t\) is time, \(\tau\) is a fixed time delay, and \(N\) and \(K\) are guidance constants. With \(K = 0\), Eq. 1 describes proportional navigation (PN), whereas with \(N = 0\), Eq. 1 describes pure proportional pursuit (PP). In the case that \(K \neq 0\) and \(N \neq 0\), Eq. 1 describes mixed PN+PP guidance. Dividing through by the hawk’s speed \(v = |v|\) converts the commanded centripetal acceleration to the commanded angular velocity. It can therefore be seen that Eq. 1 generalizes, in vector form, the PN+PP guidance law that is written as \(\dot{\gamma}(t) = N\dot{\lambda}(t - \tau) - K\dot{\delta}(t - \tau)\) in the main text, with scalar turn rate \(\dot{\gamma} = |a|/v\), line-of-sight rate \(\dot{\lambda} = |\omega|\), and deviation angle \(\dot{\delta} = |\delta|\).

Our simulations make use of the kinematic equations:

\[
r = \hat{x}_T - x
\]

\[
\omega = \frac{r \times (\hat{v}_T - v)}{|r|^2}
\]

\[
\delta = \left(\cos^{-1}\frac{r \cdot v}{|r||v|}\right)\left(\frac{r \times v}{|r \times v|}\right)
\]

where \(x\) is the simulated position of the hawk, and where \(\hat{x}_T\) and \(\hat{v}_T\) are the measured position and velocity of the target with respect to the Earth axes. Our simulations are implemented in discrete time by coupling the guidance law (Eq. 1) with the kinematics equations (Eq. 2a-c) using the difference equations:

\[
x_{n+1} = x_n + \Delta t \; v_n
\]

\[
v_{n+1} = \hat{v}_{n+1} \; \frac{v_n + \Delta t \; a_n}{|v_n + \Delta t \; a_n|}
\]

where the subscript notation indicates the values of the variables at successive time steps, such that \(t_{n+1} = t_n + \Delta t\), and where \(\hat{v}\) is the hawk’s measured groundspeed. The simulations were initiated
given the hawk’s measured initial position $\mathbf{x}_0 = \mathbf{x}_0$ and velocity $\mathbf{v}_0 = \mathbf{v}_0$, and were used to predict the trajectory that it would follow under the guidance law (Eq. 1) parameterized by the guidance constants $N$ and $K$, and time delay $\tau$. Note that Eq. 3b matches the hawk’s simulated groundspeed $\nu = |\mathbf{v}|$ to its measured groundspeed $\nu_0$ at all times, such that the guidance law is only used to command turning. We verified that the step size of our simulations ($\Delta t = 5 \times 10^{-4}$ s) was small enough to guarantee the numerical accuracy of the fitted guidance parameters and prediction error to the level of precision at which they are reported in the Results.

We defined the prediction error $\eta$ of each simulation as the mean absolute distance between the measured and simulated flight trajectories:

$$\eta = \frac{1}{k} \sum_{n=1}^{k} |\mathbf{x}_n - \hat{\mathbf{x}}_n|$$

(4)

where $\hat{\mathbf{x}}$ is the hawk’s simulated position, and $k$ is the number of time steps in the simulation. We fitted the guidance constants $K$ and/or $N$ under the various combinations of guidance law (i.e. PN, PP or PN+PP) and target definition (i.e. measured bat position, final bat position, final hawk position) for delays of $0 \leq \tau \leq 0.1$ s at 0.02 s spacing corresponding to the inter-frame interval. In each case, we used a Nelder-Mead simplex algorithm in MATLAB to find the value of $K$ and/or $N$ that minimised the prediction error $\eta$ for each flight at the given time delay $\tau$. To ensure that we fitted the same section of flight for all time delays $0 \leq \tau \leq 0.1$ s, we began each simulation from 0.1 s after the first point on the trajectory, and ended the simulation at the time of intercept or near-miss. However, as we found the best-fitting delay to be $\tau = 0$, we subsequently re-fitted the simulations with no delay to begin from the first point on the trajectory and report these simulations in the Results. For the terminal attack trajectories, we took the first point on the trajectory to be the earliest point from which it was possible to track the grabbed bat, and took the time of intercept or near-miss to be the time at which the measured distance between the hawk and bat was minimal. For the long-range approaches, we tested a range of alternative start points from 1.0 s up to a maximum of 20.0 s before the observed grab, in 0.2 s intervals, to accommodate the fact that the hawk could sometimes be tracked for longer than it appeared to be engaged in directed attack behaviour.

Statistical analysis

All statistics were computed using MATLAB. Because the distributions of the model parameters and errors are skewed (Fig. 2), we report their median, denoted using tilde notation, together with a bias-corrected and accelerated bootstrap 95% confidence interval (CI) computed using 100,000 resamples. For robustness, we use two-tailed sign tests to compare their distributions between different guidance models and target definitions. We state sample proportions together with a 95% confidence interval (CI) computed using the Clopper-Pearson method. We used a two-tailed Fisher’s exact test to compare the odds of success in attacks on lone bats versus attacks on the swarm. Following our previous observational study, bats classified as lone bats were judged to be flying > 5 body lengths from their nearest neighbours and/or appeared to be flying in a different direction to the coordinated members of the swarm (Table S3).
Data availability Trajectory reconstructions and analysis code have been uploaded to figshare https://doi.org/10.6084/m9.figshare.19196966. Video data are stored locally and will be made available upon reasonable request.

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**Competing interests** We declare no competing interests.

**Materials and Correspondence** Correspondence and requests for materials should be addressed to graham.taylor@zoo.ox.ac.uk.