Automated video tracking and flight analysis show how bumblebees solve a pattern discrimination task using active vision

3 Running title: *Active scanning strategy in bumblebees*

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14 Summary statement

Automated video tracking and flight analysis is proposed as the next milestone in understanding
 mechanisms underpinning active vision and cognitive visual abilities of bees.

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20 Abstract

21 Active vision, the ability of the visual system to actively sample and select relevant information out of a visual 22 scene through eye and head movements, has been explored in a variety of animal species. Small-brained 23 animals such as insects might rely even more on sequential acquisition of pattern features since there might 24 be less parallel processing capacity in their brains than in vertebrates. To investigate how active vision 25 strategies enable bees to solve visual tasks, here, we employed a simple visual discrimination task in which 26 individual bees were presented with a multiplication symbol and a 45° rotated version of the same pattern 27 ("plus sign"). High-speed videography of unrewarded tests and analysis of the bees' flight paths shows that 28 only a small region of the pattern is inspected before successfully accepting a target or rejecting a distractor. 29 The bees' scanning behaviour of the stimuli differed for plus signs and multiplication signs, but for each of 30 these, the flight behaviour was consistent irrespective of whether the pattern was rewarding or unrewarding. 31 Bees typically oriented themselves at ~±30° to the patterns such that only one eye had an unobscured view of 32 stimuli. There was a significant preference for initially scanning the left side of the stimuli. Our results suggest 33 that the bees' movement may be an integral part of a strategy to efficiently analyse and encode their 34 environment.

35

36 Keywords

37 active vision, Bombus terrestris, cognitive strategy, flight analysis, scanning behaviour, visual recognition

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Active scanning strategy in bumblebees

38 Introduction

39 Bees are capable of memorising and discriminating a wide variety of visual patterns, including complex ones 40 that, for example, include different stripe orientations in each of four quadrants (Benard et al., 2006; 41 Srinivasan, 1994; Srinivasan, 2010; Stach et al., 2004; Turner, 1911; Von Frisch, 1914; Wehner, 1967). On the 42 other hand, there is a long history of observations that bees are incapable of discriminating some relatively 43 simple patterns (Avarguès-Weber et al., 2012; Hertz, 1929; Hertz, 1935; Horridge, 1996; Srinivasan, 1994; Von 44 Frisch, 1914). As one example, it was reported that honeybees (Apis mellifera) were not able to distinguish a 45 "plus pattern", made up of a vertical and horizontal bar, from the same pattern rotated through 45° i.e. 46 multiplication symbol (Horridge, 1996; Srinivasan, 1994) in a Y-maze setup where the patterns were displayed 47 at a fixed distance from the bees' decision point. Given the otherwise impressive capabilities of bees in 48 recognising complex visual patterns (Avarguès-Weber et al., 2011; Dyer et al., 2005; Srinivasan, 1994), the 49 difficulty in solving the plus versus multiplication sign discrimination task by bees is surprising. There is 50 evidence that the successes and failures of bees in discriminating visual patterns are not strictly related to 51 pattern complexity, but to the visual scanning procedures that bees use when examining and memorising the 52 patterns (Lehrer and Srinivasan, 1994).

53 In vertebrates, the repertoire of such active vision strategies is already well researched (Land, 1999; Land 54 and Nilsson, 2012; Yarbus, 2013). To scan visual targets, there can be large scale movement by the body or 55 head, or smaller scale movements of the eyes (saccades) (Juusola et al., 2017; Najemnik and Geisler, 2005; 56 Yang and Chiao, 2016). Such active vision is essential to obtain an accurate three dimensional representation 57 of the material world (Kagan, 2012; Martinez-Conde and Macknik, 2008; Martinez-Conde et al., 2013; Werner 58 et al., 2016). In some vertebrates, eye movements are also used as a sampling strategy, generating fine spatial 59 information and improving the encoding of high spatial frequency of natural stimuli (Anderson et al., 2020; 60 Kuang et al., 2012; Rucci and Victor, 2015). Some animals adopt a characteristic route during a visual task to 61 facilitate target recognition (Chittka and Skorupski, 2017; Dawkins and Woodington, 2000). For instance, 62 pigeons took stereotyped approach paths when learning to discriminate visual patterns (Dawkins and 63 Woodington, 2000; Theunissen et al., 2017). Interestingly, they failed at these tasks when they were prevented 64 from using their developed route. Also, characteristic head movements were observed in pigeons when 65 stabilizing the image for forward locomotion (Theunissen and Troje, 2017).

In insects with their miniature brains, and thus possibly more limited parallel processing, there might be an even stronger need to acquire spatial information by sequential scanning than in large-brained animals (Chittka and Niven, 2009; Chittka and Skorupski, 2011; MaBouDi et al., 2020; Spaethe et al., 2006). Indeed in bumblebees, there is evidence that complex patterns cannot be discriminated when they are only briefly flashed on a screen, preventing bees from sampling in a continuous scan (Nityananda et al., 2014).

Furthermore, bees exhibit defined sequences of movements in response to particular visual stimuli (Collett et al., 1993; Guiraud et al., 2018; Lehrer and Srinivasan, 1994; MaBouDi et al., 2020; Werner et al., 2016).

Here we return to one of the pattern discrimination tasks that reportedly are challenging or impossible for bees (Srinivasan, 1994) the plus versus multiplication sign discrimination task. We examine whether, and more importantly, how bumblebees can solve it. By recording the bees' flight trajectories, and analysing their scanning movements, we aimed to determine the strategies employed in solving this visual task, specifically to investigate whether they are able to develop an active sampling strategy to solve the task if they are allowed to fly as close to the patterns as they desired.

79 Materials and Methods

80 Animals and Experimental Setup

81 Twenty bees from three colonies of bumblebees (Bombus terrestris audax, purchased from Agralan Ltd., 82 Swindon, UK) were used during this study. Colonies were housed in wooden nest boxes (28 x 16 x 11 cm) 83 connected to a wooden flight arena (60 x 60 x 40 cm) via an acrylic tunnel (25 x 3.5 x 3.5 cm). The arena was 84 covered with a UV-transparent Plexiglas ceiling (Fig. 1A). Illumination was provided via high frequency 85 fluorescent lighting (TMS 24F lamps with HF-B 236 TLD ballasts, Phillips, Netherland; fitted with Activa daylight 86 fluorescent tubes, Osram, Germany); the flicker frequency of the lights was ~42kHz, which is well above the 87 flicker fusion frequency for bees (Skorupski and Chittka, 2010; Srinivasan and Lehrer, 1984). The walls of the 88 arena were covered with a Gaussian white and pink pattern (MATLAB generated); this provided good contrast 89 between the colour of the bees and the background, required for the video analysis. Sugar water was provided 90 at night through a mass gravity feeder and removed during the day when bees were performing experiments 91 to ensure motivation. Pollen was provided every two days into the colonies.

92

93 Stimuli

94 The stimuli patterns were printed on laminated white discs (10 cm in diameter) to allow for cleaning (using 95 70% ethanol solution) in between training bouts and also tests. The training patterns consisted of two black 96 bars (1 x 10 cm) presented in two configurations: 1) Plus pattern: one vertical and one horizontal bar aligned 97 at their centre (\oplus) . 2) Multiplication pattern: same as the plus pattern but rotated by 45° (\otimes). Additional 98 patterns were constructed for a transfer test; these only presented the top half of the training stimuli (Fig. 1B). 99 All patterns had 2 mm black margins around the outer circumference of the pattern. The centre of each disk 100 was attached to the back wall of the arena via the feeder made out of a small 0.5 ml Eppendorf tube without 101 the cap (5 mm in diameter), which contained 10 μ l of either 50% sucrose solution (w/w), saturated quinine 102 solution (0.12%), or sterilised water.

103 Training and test protocol

104 Prior to the experiments, bumblebees could freely fly between the colony and a gravity feeder providing 30% 105 sucrose solution (w/w) placed in the centre of the flight arena. Successful foragers were individually marked 106 on the thorax with number labels (Opalithplättchen, Warnholz & Bienenvoigt, Germany) for identification 107 during the experiment. Marked bees were randomly selected and pre-trained to receive 50% sucrose solution 108 from eight white discs presented on the rear wall of the area. These pre-training stimuli were 10 cm in 109 diameter with 2 mm wide black margins at the edges. After several bouts of pre-training, a forager that learned 110 to take the sucrose from the feeder at the centre of the white pattern was selected for the individual 111 experiment. During training, only the selected bee was allowed to enter the flight arena.

To improve the accuracy and the speed of learning, a differential conditioning protocol was used. Four multiplication and four plus pattern stimuli were randomly affixed to set positions on the rear wall of the arena. Each stimulus was 3-6 cm horizontally, and 5 cm vertically separated from the next stimulus, or arena wall/floor/ceiling (Fig. 1B). One group of bees (n=10) was trained to receive 10 µl 50% sucrose solution (w/w) from the feeding tubes at the centre of the plus pattern stimuli, and to avoid the multiplication patterns that contained 10 µl saturated quinine solution. The second group (n=10) was trained on the reciprocal arrangement, *i.e.* associate the multiplication pattern with a reward and avoided the plus pattern.

Bees were allowed to freely choose and feed from multiple stimuli, until they were satiated and returned to their hive; empty tubes were refilled with 10 µl of sucrose solution after the bee had left the correct stimulus and made its next choice. A bout of training was completed once the bee returned to the hive. After each bout, all feeding tubes were cleaned with soap and 70% ethanol and then rinsed with water. The patterns were separately washed with 70% ethanol. Both tubes and patterns were air-dried in the lab before reuse. The position of stimuli on the wall were randomly varied for each bout to prevent bees from using the location of the reward when solving the task.

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127 After five bouts of training the bees were subjected to two tests, to evaluate if and how bees could recognize 128 and select the correct pattern. In the first test, the learning test, bees were presented with the same 129 multiplication and plus patterns used during training; this was to verify that bees had learned to associate the 130 correct pattern with the reward, and to control for any possible olfactory cues the bees may have used during 131 training. In the second test (transfer test), the bees were exposed to novel stimuli that only presented the top 132 half of the multiplication and plus patterns (see Fig. 1B). This was to determine if the bees could still recognize 133 the 'correct' pattern based only on the top half of the patterns. As during training, both tests provided four 134 correct and four incorrect stimuli, randomly positioned on the rear arena wall. All stimuli feeding tubes were 135 filled with 10 µl of sterilized water (*i.e.* no reward or punishment). One to two refreshment bouts of training

(with reward and punishment) were conducted between tests to maintain the bees' motivation. The sequenceof the two tests was randomly chosen for each bee.

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139 Video Analysis

The arena was equipped with two cameras to record all activity of bees during tests. An iPhone 5 (Apple, Cupertino, USA) with 1280x720 pixels and 240 fps (frames per second) was positioned above the arena entrance tunnel viewing the rear stimuli wall, filming the bee's flight in front of the stimuli wall and patterns. The second camera, a Yi sport camera (Xiaomi Inc. China) with 1280x720 pixels at 120 fps, was placed on the top of the rear wall orientated downward to view the stimuli. The first 120 seconds of each test were recorded and analysed.

146 To analyse bees' scanning behaviours in front of the stimuli, prior to their choices, a MATLAB algorithm was 147 developed that detected the bees automatically and then tracked the centroid of the bee bodies within each 148 frame as they flew through the arena. For each frame, the algorithm subtracted a background mask image to 149 find new candidate positions of the bee using MATLAB's blob detection function. The parameters of this 150 function were set to detect the blob with the same approximate size of a bee. In addition, an elliptic filter was 151 used in the frames from the top camera to extract the bees' body orientations. We utilised the MATLAB 152 smoothing function ('filter') to exclude any erroneous data points and correct trajectories. Examples of the 153 annotated flight paths and corresponding video recordings are shown in Figure 2 and Video S1.

154 Using the first frame of each video recording, we manually specified the x, y pixel position of each of the 155 eight pattern centres (i.e. entrances to the feeding tubes). After calculating the speed of each bee at each 156 point of the trajectory, a threshold rule was applied to the trajectories close to the feeding tube positions to 157 identify if the bees had landed, labelling the decision as either a correct/incorrect accept or rejection. This 158 'landing' threshold was determined by K-means clustering (MaBouDi et al., 2020a) of all bee speeds within the 159 specified region of the feeding tubes. For further analysis of flight speeds, distances from wall, orientation, 160 inspection times, areas of interest, and heat maps, we extracted the bees' trajectory data (using the above 161 procedure) from a cylindrical region in front of each stimulus, with a diameter 12 cm around the pattern centre 162 and 2 cm out from the stimuli wall. Bespoke MATLAB algorithms were developed to calculate and plot the 163 required datasets for each of these individual stimuli analyses (see examples: Figs. 2,3). Unfortunately one of 164 the learning test videos from Group 1 (trained to plus) was accidently recorded at just 30 fps; we therefore 165 excluded it from the above flight analysis. This video was sufficient, however, for the behavioural results of 166 the choices and rejections of the bee to be extracted.

168 Statistical analysis

169 A generalised linear mixed model (GLMM) with Binomial distribution and link function "logit" was applied to 170 the bees' choices recorded during the training phase to evaluate the effect of colony and group on bees' 171 performance and compare the learning rate between two groups of bees. To assess the bees' performances 172 in the tests, we analysed the proportion of correct choices for each individual bee. The proportion of correct 173 choices was calculated by the number of correct choices divided by the bee's total choices during the first 120s 174 of the test. A choice was defined as when a bee touched a microcentrifuge feeding tube with her antennae or 175 when she landed on a feeding tube. We then applied the Wilcoxon signed rank or Wilcoxon rank-sum tests to 176 compare the bees' responses to the learning test and the transfer test.

- 177 Results
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179 Bumblebees performance in a visual recognition task

We first confirmed that bumblebees, when allowed to fly as close to the patterns as they desired, could perform the simple visual discrimination task of identifying and associating either a plus or a multiplication symbol with a reward (sucrose solution) and the other with quinine solution (penalty, bitter taste).

- 183 Bees in Group 1 recognised the plus patterns as rewarding above chance after just 20 choices (Wilcoxon 184 signed rank test; z=2.04, n=10, p=0.04; mean=59%). Conversely, Group 2 achieved the same performance on 185 the multiplication patterns after 30 choices (Wilcoxon signed rank test; z=2.58 n=10, p=9.7e-3). Nevertheless, 186 there was no notable difference in the learning rate between the two groups after 30 choices (p=0.72), and 187 the bees' performance was not affected by colony (p=0.17) (see Table S1). The bees continued to increase in 188 performance during the 70 choices of the training (per block of 10 choices, see Fig. 1C); whereupon all bees 189 achieved \geq 92% (± 7.8 s.d.) correct choice performance. The results of a generalised linear mixed model 190 (GLMM) analysis confirmed that both groups of bees had learned to select the rewarding patterns significantly 191 above chance (>50%) after training (Fig. 1C, p=3.84e-10). Additionally, the bees' performance in the learning 192 tests indicated that both groups of trained bees successfully learned to discriminate the plus from the 193 multiplication symbol, and vice versa (Fig. 1D) (Wilcoxon signed rank test; z=2.82, n=10, p=4.8e-3 for Group 1; 194 z=2.84, n=10, p=4.5e-3 for Group 2); again there was no significant difference between the performance of 195 two groups in the learning test (Wilcoxon rank sum test; z=0.23, n=20, p=0.81). The bees' performance in the 196 leaning test was similar to that seen during the last block of 10 choices of the training phase (Wilcoxon signed 197 rank test z=1.32, n=20, p=0.18).
- 198

During the learning tests, 6 out of the 10 bees in Group 1 (trained to plus) initially inspected (flight within 12 cm diameter of centre of pattern and 10 cm out from rear stimulus wall) the correct plus pattern. However,

an equal number of bees in Group 2 also inspected the plus first, in their case the incorrect pattern. Therefore, as a whole, the bees' pattern selection from a distance (*i.e.* from arena entrance to stimuli wall) was no different to chance (50% correct initial pattern inspections; χ^2 test, Chi-square statistics=0.8, df=18; p=0.37) (Fig. 1E). In addition, during all the correct initial inspections, the bees still scanned the pattern before flying to the feeder tube (see Fig. S1).

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207 Bumblebee flight speeds and dynamics during the learning tests

To explore how bees choose the correct patterns and reject the incorrect ones, we analysed the bees' inspection behaviours, employing a custom algorithm to track the bee locations and body orientations within each frame of the videos (Fig. 2A, B and Video S1; See Video analysis in Method section).

211 The bees' initial flight speed upon entering the flight arena and approaching the first inspected stimuli was 212 on median 0.20 (± 0.13 s.d.) ms⁻¹ (Figs. 2C, S1). The speed reduced to a median of 0.11 (± 0.10 s.d.) ms⁻¹ whilst 213 in front of stimuli; the highest proportion of flight speeds was less than 0.1 ms⁻¹ (Fig. 2D). Bees' speed increased 214 to a median of 0.20 (±0.24 ms⁻¹) whilst traversing between the presented patterns. Bees typically scanned the 215 patterns from a distance of 10 mm to 50 mm from the stimuli (Fig. 2E). The bees spent approximately 1.5 (± 216 0.5 s.d.) seconds in front of a stimulus, irrespective of whether this was a plus or multiplication, or the correct 217 or incorrect pattern (Fig. 3H). The flight speed when rejecting a pattern was on average three times that of 218 when the bee accepted a pattern and flew to the feeder. However, analysis of the flight trajectories (Fig. 219 3A,C,E) shows this was due to the bee accelerating away from the current pattern to the next. Interestingly, 220 the bees showed an overall tendency to scan the patterns with their bodies oriented at ~±30° relative to the 221 rear stimuli wall, keeping one or other eye predominantly aligned to the stimuli during the scans (Fig. 3F). 222 Conversely, when flying between the patterns, they mostly looked forward in the direction of their motion 223 with a much wider range of flight directions relative to the rear wall (see Discussion).

224

225 Bumblebees scanned specific regions of the patterns prior to making a decision

226 As the bees did not appear to be making pattern selections from a distance (Fig. S1), we further analysed the 227 movements of the bees whilst directly in front of the patterns. In most instances (Group 1 trained to \oplus : 89.2%, 228 Group 2 trained to \otimes : 87%), the bees first traversed to, and then scanned, the lower part of the patterns 229 regardless of whether the target was rewarding or aversive (Fig. 2F). Each scan led to either a landing on the 230 feeding tube (an accept) or the bee flying to another stimulus without landing (a rejection). In Figure 2F, the 231 proportion of bees selecting each region of the patterns prior to a decision (accept or rejection) are plotted 232 for each group of trained bees. The highest proportion of interest was the bottom centre of the pattern with 233 correct choices of 54.5% within Group 1, and 39.7% of correct rejections in Group 2. However, this was similar 234 to the accumulated instances of lower left corner, lower right corner and both lower corners (summed totals

for correct choices Group 2: 47.5%, correct rejections Group 1: 35%). It should be noted that the bees showed a consistent preference for the lower left corner, described further below. These preferences can be clearly seen on the heat map representation of the accumulated bee positions during scanning (Fig. 3B,D).

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239 Bees trained on the protocol with the plus pattern as rewarding (Group 1) would typically approach the lower 240 half of the stimulus (89% of inspections). If a plus was observed they would scan the lower centre of the 241 pattern (containing the vertical bar) and then fly directly to the pattern centre to access the feeding tube (see 242 Video S2). However, if the bees observed a multiplication they would usually scan the lower left corner of the 243 pattern, containing the oriented bar of the multiplication. Of these trails, over half consisted of a single corner 244 scan before the bees rejected the patterns. A scan of the whole pattern was clearly not required: the 245 inspection of a single diagonal pattern element was sufficient to ascertain that the pattern was not a plus sign. 246 In the remaining cases the bees would traverse to the opposite lower corner, then scan the remaining oriented 247 bar before rejection (Fig. 2F). On average, only 4.5% of such inspections did the bees only scan the right corner. 248 Bees trained on the multiplication pattern (Group 2) showed a slightly different behaviour. If the bees were 249 inspecting a multiplication stimulus, they would first approach the left or right lower section of the pattern 250 (see Videos S3, S4). We still observed the same preference for the left side inspections, double that of the 251 lower right side scans. However, there were far fewer instances of bees inspecting both corners before flying 252 to the feeding tube (Fig. 2F). When Group 2 bees (trained to the multiplication pattern) encountered a plus 253 pattern they would again scan the lower centre at the base of the vertical bar (see Video S3). In contrast to 254 the Group 1 bees accepting the multiplication symbol, these bees would also, on occasion, scan the lower left 255 corner where no oriented bar was present (Fig. 3C).

256

To control for the possibility that the upper region of patterns may have also influenced the bees' decisions, we carried out a transfer test (see Method section), in which bees were confronted with only the top halves of the patterns. None of the bees were able to recognize the correct pattern element, choosing equally both types of novel patterns (Fig. 2G,E) (Wilcoxon signed rank test; z=0.17, n=10, p=0.85 for Group 1; z=-0.05, n=10, p=0.90 for Group 2). We therefore concluded that bumblebees learned to only use the information of the lower sections of these patterns for recognition (similar to what is seen in honeybees (Giurfa et al., 1999), see Discussion).

These results demonstrate that bumblebees are able to learn specific features within a pattern to both accept and reject stimuli. In addition, for the specific paradigm used in this experiment, bees did not make their stimulus selection from a distance, only choosing to visit a feeder after close inspection of the presented patterns.

269 **Discussion**

270

271 In this study, we aimed to explore the flight characteristics and active vision underpinning a simple visual 272 recognition task in bees. Such a task was failed by honeybees (Apis mellifera) when they were prevented from 273 viewing the stimuli up close (Srinivasan, 1994). Our results show that bumblebees (Bombus terrestris audax) 274 can discriminate these symbols within our flight arena design. However, they chose to inspect both the 275 rewarding and aversive stimuli from a distance of just 1 to 5 cm before making their decisions. There was no 276 indication that the bees chose their initial stimulus from a distance when entering the arena (with 50% of 277 initial inspections at the incorrect patterns (Fig. 1E)); even with the correctly chosen stimuli, the bees always 278 performed a scan of the pattern elements before landing on the feeder (Fig. S1). Our experimental paradigm 279 cannot confirm with certainty that bumblebees are unable to discriminate these simple patterns from a 280 distance; for that we would need to control for distance as done with the honeybee experiments (Horridge, 281 1996; Srinivasan, 1994). However, our experiment allowed us to carefully analyse the bees' scanning 282 behaviour of visual features and to extract useful insights into the active vision of bees.

283

In brief, our bumblebees had no difficulty in learning to identify and associate either the plus or multiplication signs with reward, with all bees achieving over 90% accuracy after 70 trials (Fig. 1C). This performance was preserved during the unrewarded learning tests (Fig. 1D). Our bespoke video analysis toolkit allowed us to track the bee positions and body yaw orientations for every frame of each learning test. The most notable, and consistent, characteristics observed were:

289 Partial pattern inspection. The bees primarily flew to, and scanned, the lower half of the patterns (Fig. 2F). 290 This suggests that the lower half was all the bees learned. Indeed, when exposed to a transfer test with only 291 the top half of the pattern available, bees failed to identify the correct halves of the training patterns (Fig. 2E). 292 A previous study showed that honeybees (Apis mellifera) trained in a Y-Maze using absolute conditioning 293 (where only the positive pattern and a secondary blank stimulus is provided) assigned more importance to the 294 lower half of the pattern to that of the top half (Giurfa et al., 1999). During tests with only the top half of the 295 training pattern and a novel pattern they failed to select the correct pattern half. Conversely, if bees were 296 presented with the lower half of the training pattern and again a novel pattern they could identify the correct 297 stimulus. In contrast, when trained using differential conditioning (using both rewarded and unrewarded 298 patterns), the honeybees learned the whole pattern; correctly identifying both bottom and top half patterns 299 during tests. However, in this instance, unlike in our study, the bees' choice was recorded from a distance (for 300 apparatus details, see (Horridge, 1996)) and bees' flights were not analysed systematically.

301 In a more recent study, in which the flight path of bees was also analysed, (Guiraud et al., 2018) 302 showed how honeybees (*Apis mellifera*) can solve a conceptual learning task of 'above and below' by scanning

303 the lower of two pattern elements presented on the stimuli; this provided sufficient information for the bees 304 to make a decision without needing to understand, or inspect, the relationship between the top and bottom 305 pattern elements.

306 Initial side preference. The bees had a significant preference for initially scanning the left side of the 307 multiplication pattern (Fig. 2F). This left side preference for visual objects, known as pseudoneglect, is also 308 seen in humans (Jewell and McCourt, 2000), and birds (Diekamp et al., 2005; Rugani et al., 2015). This 309 preference may allow an individual to always start its inspection of a stimulus at the same location, allowing 310 for consistent learning and recognition of natural stimuli; but it remains a curiosity as to why the left 311 preference was so prevalent amongst the bees tested (Fig. 2F). In humans and birds this lateralisation of spatial 312 attention may have evolved once in a common ancestor (Diekamp et al., 2005). However, since the visual 313 system of insects evolved largely independently from that of vertebrates, the left-side bias must have emerged 314 by convergent evolution. Its computational neural advantages in bees or vertebrates (if any), is not known.

315 *Common body orientation during scans.* The yaw orientation of the bees' bodies was most often at ~±30° 316 to the stimuli during pattern inspections. In this manner, one or other of the bees' eyes would face the pattern, 317 with only a small proportion of the opposite eye having visual access to the pattern. There was no overall 318 preference for the left or right eye (with median orientations at ~-33° and ~+27° respectively) during scans 319 (Fig. 3F). In our previous modelling work (Roper et al., 2017), we showed that lateral connections from both 320 the left and right lobula to the bee mushroom bodies allowed for better pattern recognition during partial 321 occlusion of stimuli. However, this came at the expense of fine detail recognition. Therefore, and 322 counterintuitively, having one eye mostly obscured from the pattern may provide the mushroom bodies 323 (learning centres of the bee brain) with more distinct neural inputs. It may also allow bees to learn both the 324 pattern and location cues simultaneously whilst scanning a resource. Future work will be needed to see if this 325 behaviour is particular to the patterns used in this experiment, or a stereotypical behaviour.

326 *Commonality in scan strategies is based on stimuli, not protocol.* It may seem sensible, from the bees' 327 perspective, if trained on plus, only to inspect the lower centre of the pattern for the vertical bar. However, 328 both groups of trained bees initially approached and scanned the plus and multiplication in the same manner, 329 typically checking the lower left corner of the multiplication sign and the vertical bar of the plus. This might 330 suggest that the bees did not learn the relative position of the cues and simply searched for the first visual 331 item at the lower left of the pattern. However, the flight tracking analysis conflicts with this hypothesis, with 332 the inspection of the multiplication occasionally consisting of a scan of the adjacent bar of the multiplication, 333 and with the group trained on multiplication, after scanning the vertical bar of the plus they occasionally flew 334 to the lower left corner to presumably check for the multiplication signs oriented bar. We therefore assume 335 that the stimulus is directing the scanning behaviour of the bee, and in turn the bee is learning both rewarding

and aversive pattern features during training. Other experiments will be required to ascertain the particular
 rules which dictate the bees' flight manoeuvres based on the 2D and 3D stimuli provided.

338 In the pioneering works of Karl Von Frisch, free-flying bees were trained to find sugar reward on certain black 339 or coloured patterns placed horizontally on a white background (Von Frisch, 1914). Later studies showed that 340 bees only used local cues corresponding to their approach direction when the stimuli were presented to them 341 horizontally (Wehner, 1967). Since bees were not able to capture global shapes, this might be the reason bees 342 could only recognise some simple patterns in the early studies. However, vertical presentation of stimuli was 343 developed to examine what diversity of visual features bees may use, such as orientation (Van Hateren et al., 344 1990), radial or, bilateral symmetry (Giurfa et al., 1996; Horridge, 1996), or spatial frequency and ring-like 345 structures (Horridge and Zhang, 1995). To control the decision distance and understand which cues were 346 utilised by bees to recognise the target pattern, the Y-maze was introduced (Srinivasan and Lehrer, 1988). 347 Previous research has shown that honeybees and bumblebees can solve visual tasks by extracting the localised 348 or elemental features within the pattern (Giurfa et al., 1999; Guiraud et al., 2018). Bees may use different 349 parts of a stimulus to discriminate between correct and incorrect patterns, depending on the training protocol 350 employed or the specific patterns presented (Giurfa et al., 1999; Stach and Giurfa, 2005). Although the Y-maze 351 enabled researchers to control the cues that bees could see when making decisions about visual patterns from 352 a distance, it is a less useful paradigm to inspect the scanning strategies used by bees. Therefore, despite 353 several decades of research in bee vision, it is still debated why, and how, bees fail to recognise some simple 354 patterns while they show excellent recognition in other complex patterns (Avarguès-Weber et al., 2011; Dyer 355 et al., 2005; Srinivasan, 2010). We therefore used an experimental setup in which the bees' flight and scanning 356 behaviour could be examined while they were close to the targets that were to be discriminated.

357 In this study we showcase a new suite of tools for automatic video tracking of bees in free flight and during 358 their scanning manoeuvres, as well the algorithms needed to analyse and visualise the large amount of 359 positional and orientation data this tracking produces. In our previous work on 'above and below' conceptual 360 learning (Guiraud et al., 2018) we had to manually view and annotate 368 hours of video footage (46 hours of 361 video footage taken at 120 fps, watched at 1/8th speed). In contrast, here the only manual process was 362 providing a mask frame (without the bee present) per test, and marking the feeder positions within that frame. 363 With only a small number of test videos to process this was not an issue, but even here, recent advances in 364 making convoluted neural networks for pattern recognition accessible to non-programmers 365 (playground.tensorflow.org, runwayml.com), as well as the more research programmer-centric DeepLabCut 366 (Nath et al., 2019), allows researchers to provide a few dozen labelled mask frames and have these systems 367 process thousands of mask images for all the other videos (Egnor and Branson, 2016). Similarly, the ability to 368 visualise either individual flight paths (Fig. 3A) or combined heat maps of positional data (Fig. 3B) allowed us

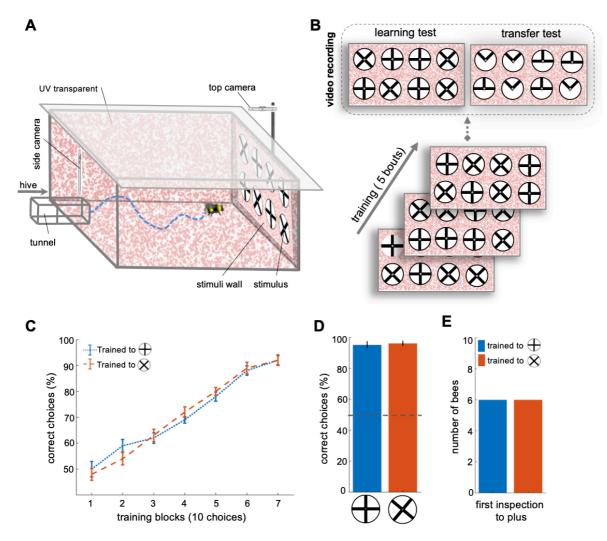
369	to quickly identify behavioural aspects of interest. Histograms of velocity, distance and orientation can b
370	quickly generated, but more importantly the parameters defining the areas of interest can be modified and
371	processed in a matter of minutes. Previous studies have relied upon binary fixed decision lines (Avarguès-
372	Weber et al., 2012; Horridge, 1996; Horridge and Zhang, 1995; Srinivasan and Lehrer, 1988), with
373	experimenters manually recording these limited behavioural data. Our in-depth analysis on such a
374	straightforward pattern recognition task highlighted key behavioural characteristics, which can now influence
375	future work on active vision, this simply would not have been viable without these automated tools.
376	
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379	
380	Competing Interests
381	All authors declare no conflict of interest.
382	
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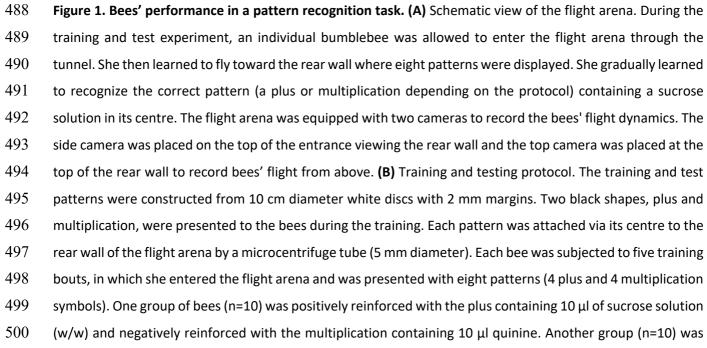
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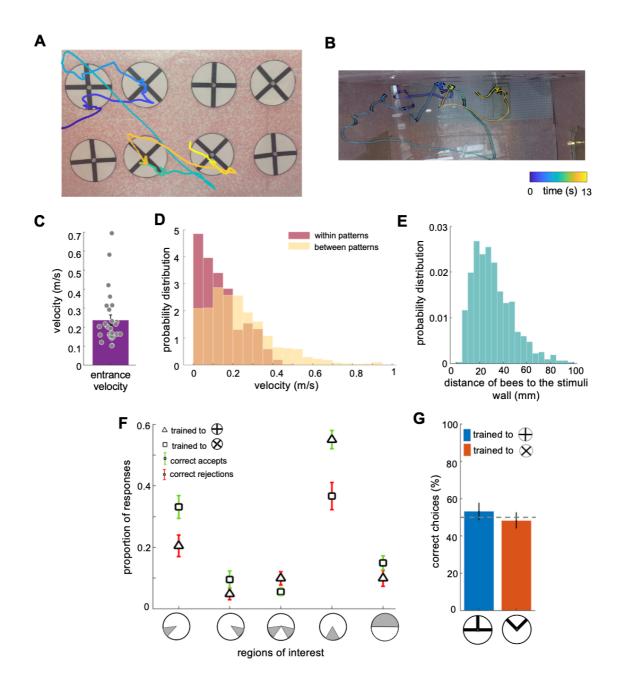
486 Figure legends







501 trained on the reciprocal protocol. The bees were free to sample the rewarding and unrewarding patterns 502 multiple times (refilled after departing), and return to the nest box when satiated. The place of patterns was 503 randomly changed before each bout of the training phase. Following training, each bee's performance was 504 examined with two tests; here the positive and negative reinforcement were replaced with sterile distilled 505 water. In the learning test, the bee was presented with the same patterns as during training. In the second 506 test, the bee was confronted with the novel patterns that only displayed the upper half of the plus and 507 multiplication symbols to the bees. One or two refreshment training bouts were used between tests to keep 508 the bee motivation high. Bees' flight paths were recorded for the initial 120 seconds via the two cameras. (C) 509 The learning curves of two groups, blue: Group 1 (trained to plus rewarding), orange: Group 2 (trained to 510 multiplication rewarding). Both groups of bees similarly learned to distinguish between patterns. (D) The 511 performance of bees during the unrewarded learning test shows that all bees had successfully learned to 512 distinguish between patterns (p<4.8e-3). (E) The number of first pattern inspections, upon entering the flight 513 arena, that were of the plus symbol. The bees did not fly to the correct pattern from a distance (> 5 cm) more 514 than chance (accumulated results: 10/20 correct initial visits). Blue: Group 1 (trained to plus rewarding), 515 orange: Group 2 (trained to multiplication rewarding).

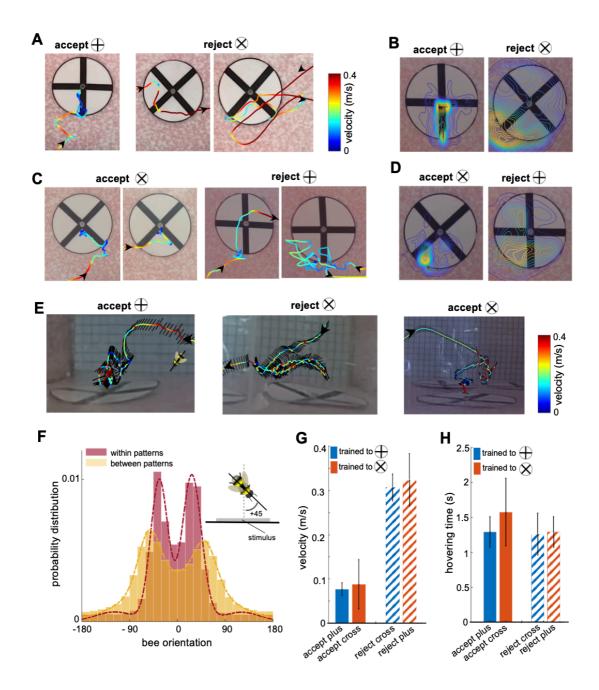


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519 Figure 2. Bees' flight analysis demonstrates an efficient strategy of scanning in bumblebees (A) An example 520 of a flight path showing the activity of a bee during part of the learning test; presented bee trained to select 521 the plus and avoid the multiplication. Each point on the flight path corresponds to a single video frame with 522 an interval of 4 ms between frames which was recorded from the front camera (left panel) and top camera 523 (right panel). The bee sequentially inspected each pattern, correctly landed on multiplication and avoided the 524 plus. The colour map changes from blue to the yellow with increasing time (See Video S1). The black lines in 525 the left panel exhibit bee's body yaw orientations. (B) Distribution of average entrance flight speed toward 526 the wall in the learning tests (See Fig. S1). Filled dots: speed of each individual bee. (D) Probability distribution 527 of the bees' speed in two conditions; when they were inspecting patterns (Red), and when they were flying 528 between patterns (Yellow). This indicates that they scanned patterns slower when they accepted them than

- when they flew to another pattern. **(E)** Probability distribution of the bees' distance from the stimuli wall whilst inspecting patterns. **(F)** From video analysis, the proportion of scanned regions (mean ± s.e.m.) of bees' inspections before the correct accept or correct rejection (x-axis: regions of interest are highlighted in grey). Triangles: inspection proportion (mean ±SEM) of Group 1 bees (trained to plus rewarding), squares: Group 2 bees (trained to multiplication), green error bars: correct accepts; red error bars: correct rejections. **(G)** The performance of bees during the novel test; bees equally chose both upper half-patterns (p>0.85), demonstrating that they did not learn the upper half of the patterns during training. Grey dashed lines=chance
- 536 level (50%). Bar: mean performance (error bars: ± s.e.m.).



538 Figure 3. Bee scanning strategy in a pattern recognition task. (A) The flight paths of one example of 539 acceptance and two examples of rejection behaviours of a bee trained to plus; the bee accepted the plus 540 pattern after scanning the lower half of the vertical bar, while she rejected the multiplication pattern after 541 scanning one or both diagonals bars. Line colour: flight speed 0.0 - 0.4 ms⁻¹ (See Videos S2, S3). (B) Group 1 542 (trained to plus) probability maps (heat-maps) of bees' locations per frame in front of plus and multiplication 543 type stimuli during all learning tests. The yellow colours show most visited regions. (C & D) same analysis as 544 A, B for Group 2 bees trained to discriminate multiplication from plus. This indicates that bees typically 545 scanned the lower half of the pattern with a lower speed to Group 1 bees, prior to their decisions (See Videos 546 S4, S5). (E) Three examples of bees' flight paths shown from the top camera; black lines show bees' body 547 orientation during the flight, and arrows designate the start and ending time of scanning. (F) probability

548 distribution of the bees' body yaw orientation perpendicular to the rear stimuli wall in two conditions: when 549 they were inspecting patterns (red) and when they were flying between patterns (yellow). Inset figure exhibits 550 one example of bee's orientation with +45°. This suggests the bees viewed the patterns at a median ~±30° 551 whilst scanning, with one or other eye having a predominant view. On the other hand, when they flew to 552 another pattern the body orientation was more parallel to the flight direction with a wider distribution of 553 orientations relative to the stimuli wall, resulting in a median of ~±50° perpendicular to the stimuli wall. The 554 dashed lines show the Gaussian mixture distribution models were fitted to each distribution (flights within 555 patterns: $\mu_1 = +27$, $\mu_2 = -33$; flights between patterns: $\mu_1 = +51$, $\mu_2 = -55$). (G) mean flight speed (± 556 s.e.m.) of scanning flight prior to decisions (accept and rejection) for both groups of bees. Blue: Group1 557 (trained to plus); orange: Group 2 (trained to multiplication). (H) inspection time (i.e. the time spent hovering 558 in front of a pattern) for each symbol type for both groups of bees; inspection times of bees in front of both 559 pattern types were equal regardless of their decision or training protocol.

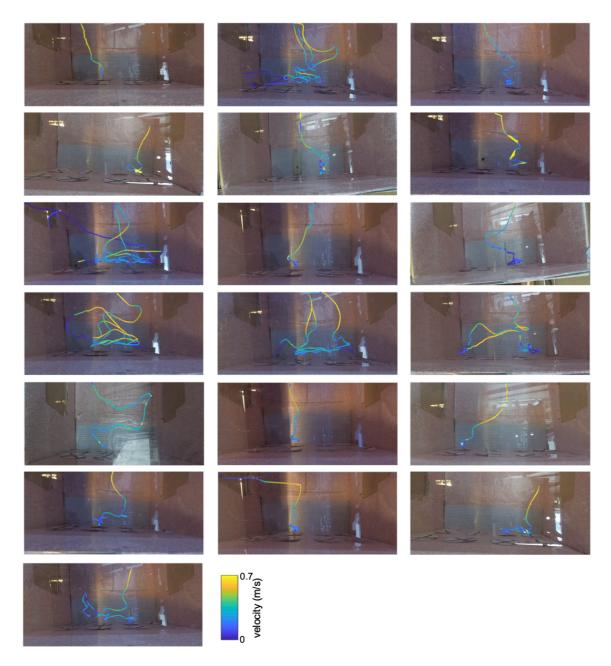
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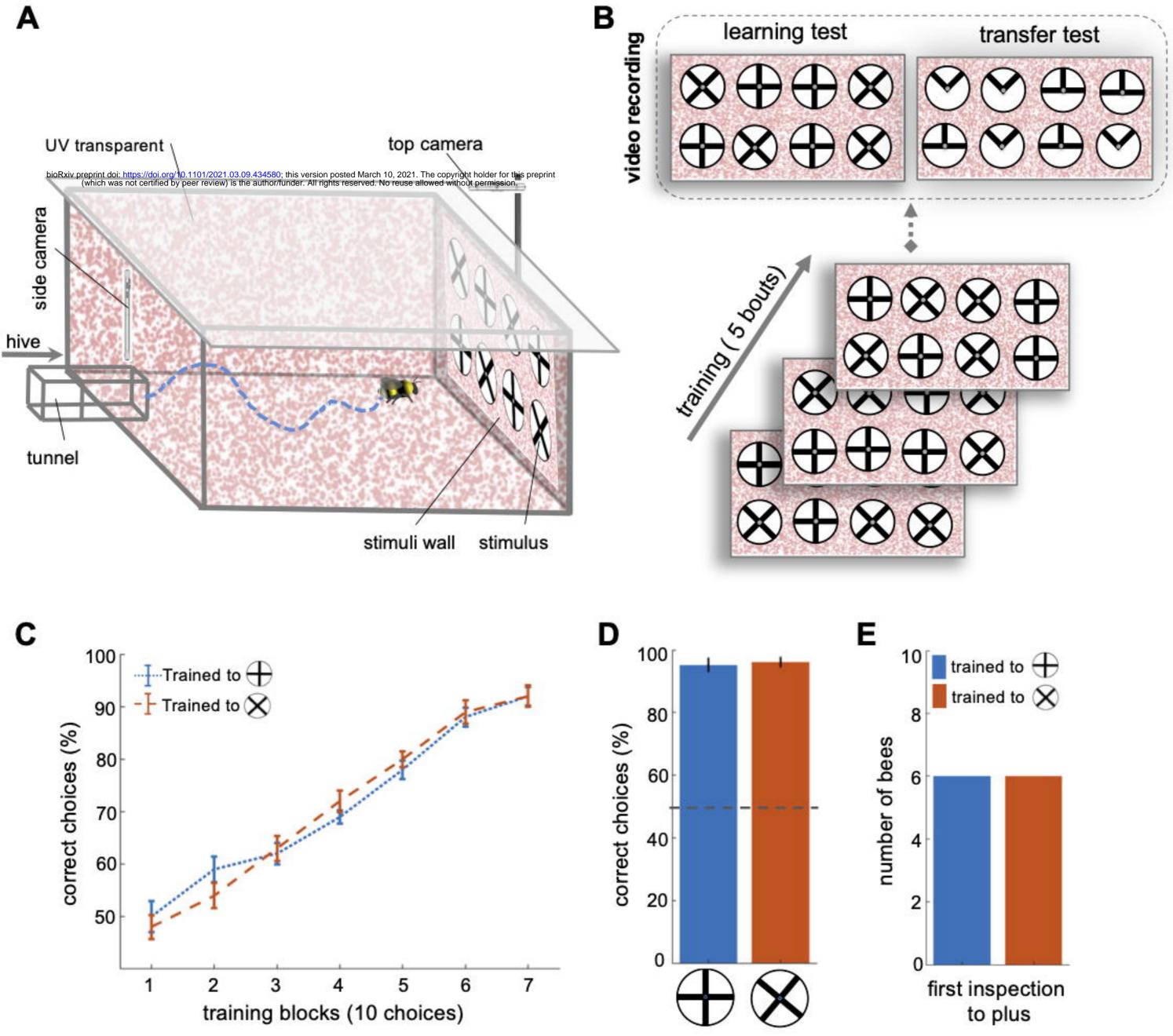
564 Appendices

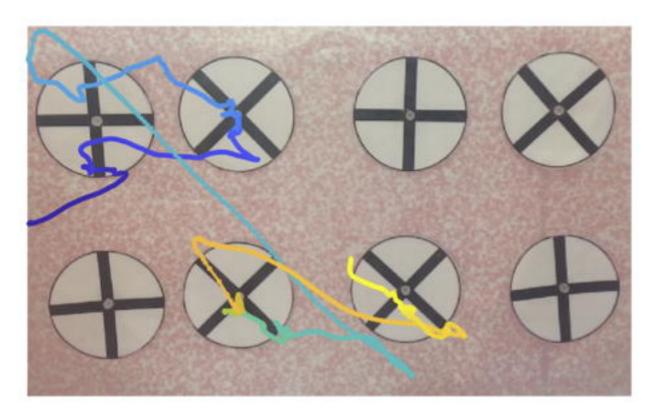
565 Supplementary Figures



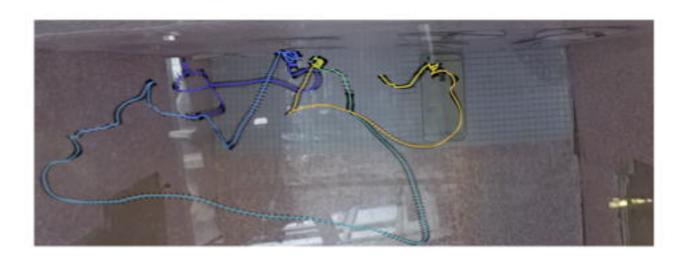
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Figure S1. Bees' flight paths upon first entering the arena during learning tests. In ten instances the bees initially inspected the correct stimulus, scanned the pattern and visited the feeder. In the remaining nine flights the bees initially inspected the incorrect pattern, then rejected the pattern and flew to another, usually adjacent pattern. One video is missing where the footage was only recorded at 30 fps; this bee initially inspected the incorrect pattern, and again rejected the stimulus. The bees' first inspection appears to be random with 50/50 correct pattern selections from the arena entrance; this suggests bees have to scan the stimuli before making decisions. Line colour from blue to yellow: flight speed 0.0 - 0.7 ms⁻¹.

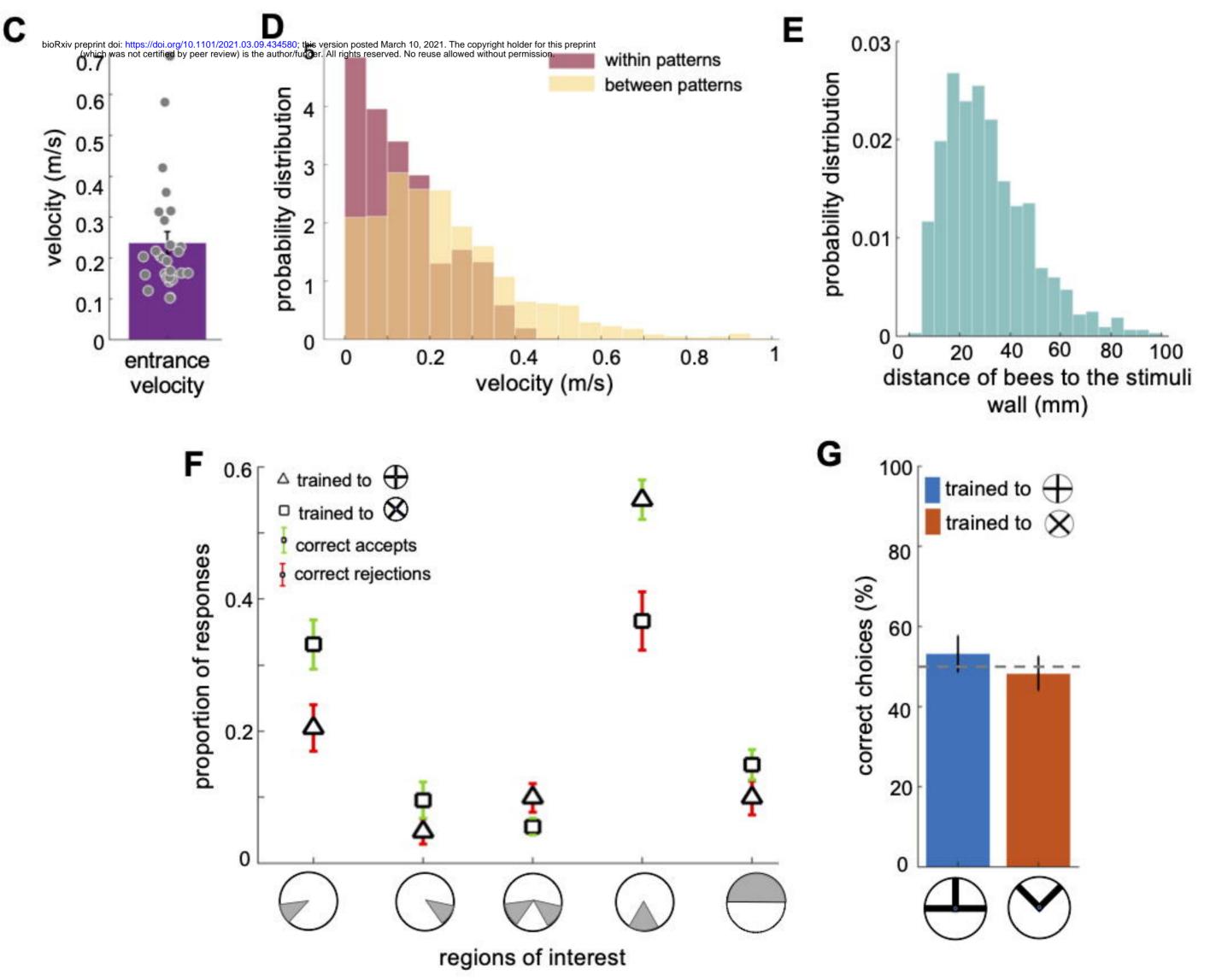




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