

1 Learning modifies attention during bumblebee visual search
2 Théo Robert¹, Karolina Tarapata¹, Vivek Nityananda¹
3 Biosciences Institute, Newcastle University, Henry Wellcome Building, Framlington Place, Newcastle
4 Upon Tyne, NE2 4HH, UK

5 Corresponding authors:

6 Théo Robert
7 Email: theo.robert@newcastle.ac.uk
8 Vivek Nityananda
9 Email: vivek.nityananda@newcastle.ac.uk

10 Orcid ID:

11 Vivek Nityananda: 0000-0002-2878-2425

12 **Abstract**

13 The role of visual search during bee foraging is relatively understudied compared to the choices made
14 by bees. As bees learn about rewards, we predicted that visual search would be modified to prioritise
15 rewarding flowers. To test this, we ran an experiment testing how bee search differs in the initial and
16 later part of training as they learn about flowers with either higher- or lower-quality rewards. We then
17 ran an experiment to see how this prior training with reward influences their search on a subsequent
18 task with different flowers. We used the time spent inspecting flowers as a measure of attention and
19 found that learning increased attention to rewards and away from unrewarding flowers. Higher
20 quality rewards led to decreased attention to non-flower regions, but lower quality rewards did not.
21 Prior experience of lower rewards also led to more attention to higher rewards compared to
22 unrewarding flowers and non-flower regions. Our results suggest that flowers would elicit differences
23 in bee search behaviour depending on the sugar content of their nectar. They also demonstrate the

24 utility of studying visual search and have important implications for understanding the pollination
25 ecology of flowers with different qualities of reward.

26 **Keywords:** Reward, Saliency, Top-down attention, Insects, Bees, Cognitive Ecology

27 **Significance Statement**

28 Studies investigating how foraging bees learn about reward typically focus on the choices made by the
29 bees. How bees deploy attention and visual search during foraging is less well studied. We analysed
30 flight videos to characterise visual search as bees learn which flowers are rewarding. We found that
31 learning increases the focus of bees on flower regions. We also found that the quality of the reward a
32 flower offers influences how much bees search in non-flower areas. This means that a flower with
33 lower reward attracts less focussed foraging compared to one with a higher reward. Since flowers do
34 differ in floral reward, this has important implications for how focussed pollinators will be on different
35 flowers. Our approach of looking at search behaviour and attention thus advances our understanding
36 of the cognitive ecology of pollination.

37 **Introduction**

38 Foraging bees must learn which flowers are rewarding and which ones are not. Given this ecological
39 demand, they have evolved to be expert learners and are well-studied as models of visual cognition
40 (Avarguès-Weber et al. 2011; Giurfa 2012). Bees learn to choose rewarding flowers and avoid
41 differently coloured flowers without rewards (Lubbock 1881; Turner 1910; von Frisch 1914; Benard et
42 al. 2006; Avarguès-Weber and Giurfa 2014). They are also capable of discriminating between higher
43 and lower rewarding flowers (Baude et al. 2011; Riveros and Gronenberg 2012; Avarguès-Weber et al.
44 2018; Solvi et al. 2022). While a large body of research has demonstrated reward-based learning in
45 bees, most of the work has looked at how learning affects the choices made by bees. Much less
46 research has investigated the influence of rewards on visual search and attention in bees (Spaethe et
47 al. 2006; Morawetz and Spaethe 2012; Nityananda and Pattrick 2013; Nityananda and Chittka 2021).

48 Visuospatial attention has been defined as a spotlight focussing on one region compared to others
49 (Posner 1980) and is often measured by responses to targets in a region or the time spent looking at
50 specific regions or objects (Schütz et al. 2011; Henderson and Hayes 2018). Visual search experiments
51 look at how attention is deployed when searching for one target amongst others (Horowitz and Wolfe
52 2001). This approach has been used in several animals including jays, owls and fish (Dukas and Kamil
53 2000; Bond and Kamil 2002; Ben-Tov et al. 2015; Orlowski et al. 2015, 2018; Saban et al. 2017). Recent
54 work has begun to look at attention and visual search in insects (Nityananda 2016), especially in
55 bumblebees. Bees have been shown to flexibly switch between multiple rewarding targets
56 (Nityananda and Patrick 2013; Li et al. 2017). In experimental set-ups, floral rewards influence not
57 just their choices but their visual attention, as measured by the time spent around particular flowers
58 (Nityananda and Chittka 2021). In that study, bees spend more time around higher rewarding flowers
59 even when they are less salient than lower rewarding flowers. We still, however, know little about
60 how bee visual search changes over time as the bees learn about rewards.

61 Bee attention during learning could be influenced by multiple factors, including the reward value and
62 the saliency of the flowers. These factors have been shown to influence both bee choices and visual
63 search. Colour contrast against a background, one measure of saliency, influences their visual search
64 (Goulson 2000) – bee foraging efficiency reduced when the background was similar to the flowers.
65 Naïve bees also have an innate bias toward colours in the blue-green wavelength range and colours
66 that have spectral purity (Lunau 1990; Lunau et al. 1996). We would therefore expect that the visual
67 search of naïve bees would initially be directed towards more salient flowers and those for which they
68 have innate biases. Subsequently, as bees learn about the reward value of flowers, we should expect
69 them to pay more attention to rewarding flowers. We would also predict that there would be different
70 effects on bumblebee visual search if the rewarding flowers had lower rewards or higher rewards.
71 Given the effects of both reward value and stimulus saliency we would therefore expect learning to
72 increase attention to higher reward but lower saliency flowers compared to lower reward high
73 saliency flowers.

74 To test these ideas, we investigated the training bouts for bees trained as part of a previously
75 published study (Nityananda and Chittka 2021) that focussed on the behaviour of bees in tests after
76 the training. As part of that study, bees were trained on one of two flower types – either higher reward
77 lower saliency flowers or lower reward higher saliency flowers. Saliency was defined there too as
78 colour contrast against the background. In both training regimes, the rewarding flowers were
79 presented simultaneously with non-rewarding distractors. In the current study, we focus on the
80 training period prior to the tests, that have not been previously analysed. Our data are thus new data
81 not used in the previous study. We investigated the effect of learning on attention by comparing visual
82 search in the initial period of the training with visual search in the final stage of the training. Prior
83 expectations can change the perception of reward in social insects (Bitterman 1975; Gil et al. 2007;
84 Wendt et al. 2019). In a second experiment, we therefore also investigated how prior experience of
85 higher rewards and lower rewards influenced visual search when encountering new flowers that had
86 reward values that were lower or higher respectively.

87 We hypothesised that bees would increase their attention to rewarding flowers as they learnt about
88 the rewards and that this effect would be greater for flowers with higher reward. We therefore
89 predicted that in the first experiment, bees would attend more to rewarding flowers in the final phase
90 of their training compared to the initial phase of their training. We further predicted that this change
91 would be greater for the higher reward lower saliency flowers. Given the possibility of prior
92 expectations influencing behaviour, we hypothesised that bees that had experienced higher reward
93 should be less motivated by lower rewards. In the second experiment, we therefore predicted that
94 bees would spend more attention away from rewarding flowers if they encountered lower rewarding
95 flowers after having prior experience of higher rewards. We also predicted that bees would increase
96 attention to rewarding flowers if the bees encountered higher rewards after experiencing lower
97 rewards first.

98 **Materials and methods**

99 Bees

100 We obtained the bees from a commercial supplier (Syngenta Bioline, Weert, The Netherlands). We
101 then tagged them with Opalith number tags (Christian Graze KG, Weinstadt-Endersbach, Germany)
102 which allowed us to individually identify them. We transferred the bees under red light to one of two
103 chambers of a wooden nest box (length × width × height: 28 × 16 × 11 cm). The floor of the other
104 chamber was covered with cat litter to allow bees to discard refuse. This nest box was connected to
105 an arena with a 24.5 cm long transparent Perspex tunnel. The arena consisted of a wooden box (length
106 × width × height: 100 × 60 × 40 cm) covered with a UV-transparent Plexiglas lid and the arena floor
107 was covered with green card. The arena was lit from above with two twin lamps (TMS 24 F with HF-B
108 236 TLD (4.3 kHz) ballasts; Philips, The Netherlands) fitted with Activa daylight full spectrum
109 fluorescent tubes (Sylvania, New Haven, UK). Bees were allowed to forage for sucrose solution in the
110 arena and provided ~ 3g pollen directly in their colony every alternate evening.

111 Spectral reflectance of flowers

112 We used an Avantes AvaSpec 2048 spectrophotometer (Anglia Instruments Limited, Soham, UK) along
113 with a deuterium-halogen light source relative to a BaSO₄ white standard to measure the reflectance
114 spectra of the artificial flowers. We converted the spectra obtained into a bee-specific colour space
115 (Chittka 1992) using the spectral sensitivity of bumblebee photoreceptors (Skorupski et al. 2007), the
116 spectral distribution of the lights used and the spectral reflectance of the background. The colour
117 hexagon space has three vertices representing the points of maximum excitation of the blue, green
118 and ultraviolet (UV) photoreceptors of the bee (Figure 1A). The other three vertices correspond to the
119 response to mixtures of approximately equal excitation of each combination of two photoreceptors.
120 The Euclidean distance between the centre of the hexagon and each vertex is 1 and colour distance
121 greater than 0.1 can be distinguished by bees without special training procedures. After plotting the
122 reflectance values of our flowers in this space, we were able to measure the distance in perceptual
123 space between them. These data are provided in the previous paper (Nityananda and Chittka 2021).

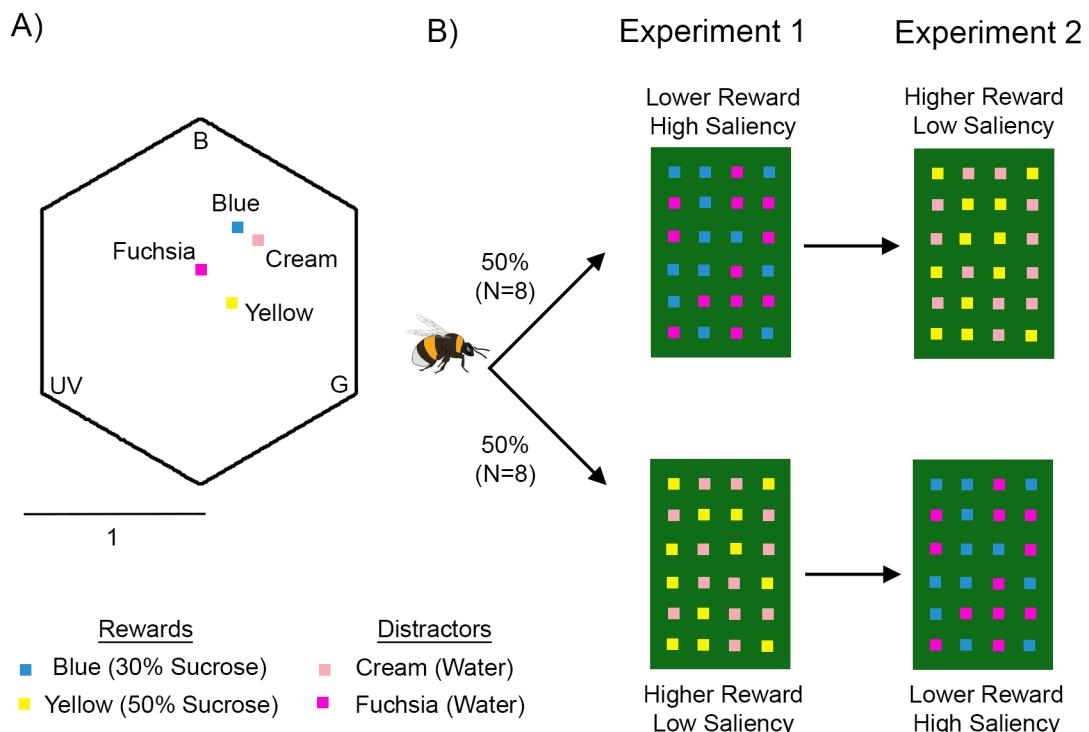
124 Pretraining

125 We trained bees with no experience of colour to forage for sucrose solution from transparent square
126 Perspex chips (side: 25 mm, thickness: 5 mm). These served as artificial flowers (henceforth “flowers”)
127 and the aim of the pretraining was to allow the bees to learn to forage from them. Each flower had a
128 central well that could be loaded with rewarding or unrewarding solutions. Once bees learned to
129 forage from these chips, we placed them on glass vials (1.5 cm diameter, 4 cm tall) and trained bees
130 to forage from them. We arranged 24 such vials in a 6 X 4 horizontal grid, placed 15 cm apart. Twelve
131 of the flowers had 12 μ l of 50% (v/v) sucrose solution in them and the others were empty. We
132 randomized the positions of rewarding and non-rewarding flowers. In the pretraining and in all
133 experiments these positions were randomized using the random number generator function RAND()
134 in Microsoft Excel®. We moved to the training phase once the bee had foraged on this grid for three
135 bouts.

136 Training

137 We trained 16 bees from three colonies on a visual discrimination task. Bees had to discriminate
138 between rewarding flowers (targets) of one colour and unrewarding flowers of another colour. These
139 flowers were coloured Perspex chips placed on glass vials in a grid as described above. In each
140 experiment, there were a total of 12 rewarding flowers and 12 unrewarding flowers. Rewarding
141 flowers contained 12 μ l of 50% (v/v) sucrose solution while the others contained 12 μ l of distilled
142 water. Within one foraging bout, flowers were not refilled but bees were allowed to revisit flowers
143 multiple times. Bees were allowed to forage over multiple bouts until they made 80% correct choices
144 of the rewarding flowers in their last 20 choices. Choices were recorded when the bees landed on a
145 flower and probed them for reward, including when bees revisited flowers. Between training bouts,
146 we cleaned the chips with 99% ethanol to remove scent markings, and then with water to remove
147 traces of ethanol. The positions of the rewarding flowers and unrewarding flowers was then
148 randomized again before the next bout.

149 We trained each bee in two consecutive experiments (Figure 1B). The first looked at how training
 150 affected the visual search of naive bees and the role of reward quality. The second experiment looked
 151 at how prior training (in the first experiment) affected subsequent visual search when different
 152 rewards were encountered.



153

154 Figure 1: Experimental protocol. A) Colour loci of the artificial flower colours used in the experiments
 155 in the colour hexagon (Chittka 1992). Three vertices correspond to maximum excitation of
 156 photoreceptors sensitive to blue (B), green (G) and ultraviolet (UV) light. The distance from the centre
 157 to any vertex is 1 (see scale) and represents how salient is a colour. The distance between points
 158 represents hue discriminability, with 0.1 being easily distinguishable. B) Training paradigm in the
 159 experiments. Half the bees followed the protocol in the top row and half the bees followed the one in
 160 the bottom row. High and low reward and saliency refers to the rewarding flowers. Rewarding (Blue
 161 and Yellow) and unrewarding (Fucshia and Cream) colours were the same in both but the order in
 162 which they were encountered (i.e. Experiment 1 or 2) was reversed. The rewarding flowers had 12 μ l

163 of 30% (lower reward) or 50% (higher reward) sucrose solution while unrewarding flowers had an
164 equal quantity of distilled water.

165 Experiment 1: The Effects of Colour-Naïve Training

166 Bees were divided into two groups of eight bees each (Figure 1B). The first group was trained on blue
167 rewarding flowers with a lower reward of 30% (v/v) sucrose solution and unrewarding fuchsia flowers.
168 The second group were trained on yellow rewarding flowers with a higher reward of 50% (v/v) sucrose
169 solution with unrewarding cream flowers. We trained the bees until they reached the success criterion
170 defined above.

171 Experiment 2: The Effects of Prior Training

172 In the second experiment (Figure 1B, right column), we continued to train the bees from Experiment
173 1 above on a novel task. For this task, we swapped the training regimes described in Experiment 1.
174 Bees that were trained on lower rewarding blue flowers in Experiment 1 were now trained on higher
175 rewarding yellow flowers with unrewarding cream flowers. Bees in the other group that were trained
176 in Experiment 1 on higher rewarding yellow flowers were now trained on lower rewarding blue flowers
177 with unrewarding fuchsia flowers. Higher rewards were always 50% (v/v) sucrose solution and lower
178 rewards were always 30% (v/v) sucrose solution. Unrewarding flowers always only held distilled water.

179 The choices made by the bees were noted to determine the success criterion and all bouts were
180 recorded using a Sony DCR-SR58E Handycam at 25 frames per second.

181 Video Coding

182 We analysed the videos using the open-source program Tracker (V5.15, ©2020 Douglas Brown,
183 physlets.org/tracker). We perspective corrected the videos and tracked the position of the bees in
184 each frame. Frames in which the bee was not clearly visible because of light reflections or because it
185 flew to a corner of the arena were marked as missing data and excluded from subsequent analysis.
186 We used the tracked position of the bees to obtain maps of bee search behaviour. To exclude time

187 spent visiting a flower we excluded frames on which the bee was within 1.77 cm from the centre of a
188 flower (for both unrewarding flowers and rewarding flowers). This distance corresponds to the
189 diagonal length from the centre to any corner of the artificial flowers we used. The visual search area
190 was thus the area of our arena, after excluding the flower areas. Within the visual search area, we
191 defined flower inspection regions on our maps as between 1.77 cm and 5 cm from the flowers' centre.
192 All other regions in the visual search area were defined as non-flower regions. We summed the total
193 number of frames that a bee spent in each region and converted this to a measure of inspection time
194 by dividing by the frame rate of the videos.

195 To investigate the effect of learning we compared the change in inspection time as our measure of
196 attention. With this measure, we tested how attention for rewarding flowers, unrewarding flowers,
197 and other regions differed between the first six choices the bees made and the last six choices. We
198 made the same comparison for both experiments.

199 Statistical Analysis

200 All analyses were run on R (version 4.2.1). We analysed the results using the `glm` and `glmmTMB`
201 function of the `glmmTMB` package (Brooks et al. 2017) to run general linear and generalized linear
202 mixed models. We assessed the fit of all our models using the `DHARMA` package (Hartig 2022).

203 To analyse the inspection time results, we calculated the proportion of time spent in each region
204 (rewarding, unrewarding or other) compared the total visual search time. To control for the differing
205 areas of each region we divided these proportions by the total area corresponding to each region. We
206 then log-transformed the weighted proportions and used this as the dependent variable in a general
207 linear model. We used the models to test the three-way interaction effect of learning stage (first or
208 last six), rewarding flower colour (blue or yellow) and region (rewarding, unrewarding or other), with
209 each of these predictors included as factors. We ran the same analysis separately for both the first
210 and second experiments. For the second experiment, we had to exclude two data points where the
211 proportion of frames for the unrewarding flower was 0 and therefore could not be log-transformed.

212 We also ran an analysis on the duration (total number of frames) over which bees made their first and
213 last choices. To do this we used the glmmTMB function from the glmmTMB package to run a
214 generalised linear mixed model with duration as the dependent variable and a negative binomial
215 family. The independent variables were learning stage and rewarding flower colour and bee identity
216 was included as a random variable.

217 Finally, we analyzed the choices of the bees using a generalized linear mixed model with the
218 proportion of choices of rewarding flowers as the dependent variable, bee identity as a random
219 variable and a binomial family distribution and a logit link function. In consecutive models, we included
220 rewarding flower colour, experiment and learning stage as independent variables and selected the
221 best model after comparing the models with anova() function in R. Details of the model selection
222 process are provided in the code in the supplementary material.

223

224 **Results**

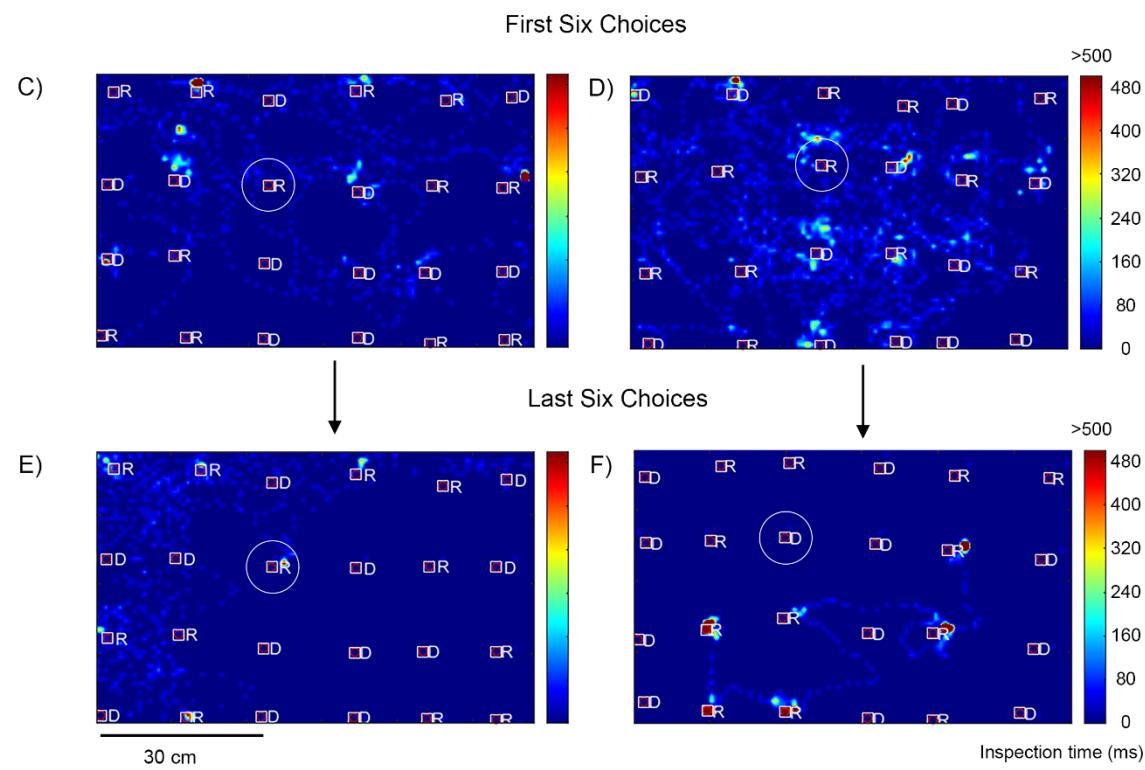
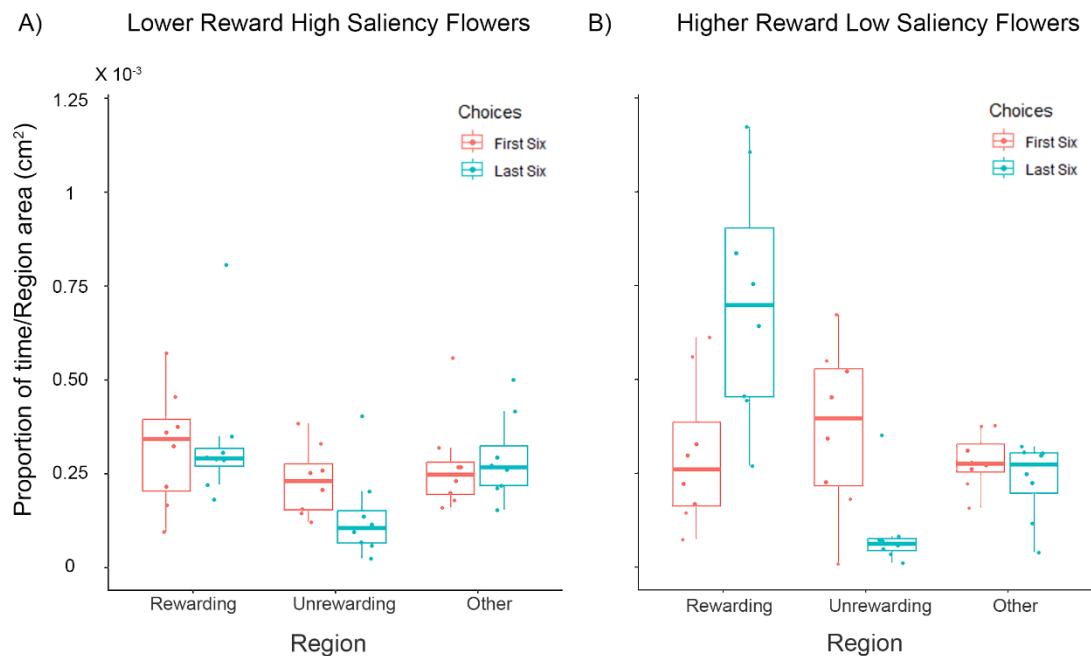
225 **Experiment 1:**

226 Our model shows a significant main effect of training stage on bee inspection time (GLM, Estimate =
227 0.957, S.E.= 0.350, P = 0.008; Figure 2A and B), showing that as bees learnt about the rewards they
228 were more likely to spend time inspecting rewarding flowers. In the first training stage, there were no
229 significant main effect of flowers of different reward value (GLM, Estimate = 0.132, S.E.= 0.350, P =
230 0.708, Figure 2A and B, pink plots). In this stage, bees were not significantly more likely to inspect
231 rewarding flowers compared to unrewarding flowers for both values of reward (Yellow flowers: GLM,
232 Estimate = -0.022, S.E. = 0.350, P= 0.951; Blue flowers: GLM, Estimate = -0.264, S.E. = 0.350, P = 0.453,
233 Figure 2B, pink plots). They were also not significantly more likely to inspect rewarding flowers
234 compared to non-flower regions for either reward value (Yellow flowers: GLM, Estimate = 0.102, S.E.
235 = 0.350, P = 0.771; Blue flowers: GLM, Estimate = -0.105, S.E. = 0.350, P= 0.765, Figure 2A, pink plots).

236 For the higher rewarding yellow flowers, there were interaction effects between the training stage
237 and the region type, showing that bees spent significantly more time inspecting these rewarding
238 flowers compared to unrewarding flowers and other regions in the later stage of training
239 (Unrewarding flowers: GLM, Estimate = -2.335, S.E.= 0.495, P < 0.001; Non-flower regions: GLM,
240 Estimate = -1.282, S.E.= 0.495, P = 0.011; Figure 2A and B, blue plots). This was not true for the lower
241 rewarding blue flowers (Unrewarding flowers: GLM, Estimate = -0.843, S.E. = 0.495, P = 0.092; Non-
242 flower regions: GLM, Estimate = -0.019, S.E. = 0.495, P = 0.970; Figure 2A and B, blue plots)

243 In the later training stage, there was a main effect of reward value on the inspection time of reward,
244 indicating that bees were more likely to inspect the higher rewarding yellow flowers compared to
245 lower rewarding blue flowers (GLM, Estimate = -0.730, S.E. = 0.350, P = 0.040). At this training stage,
246 bees inspected both types of flowers significantly more than unrewarding flowers (Yellow flowers:
247 GLM, Estimate = -2.356, S.E. = -6.738, P < 0.001; Blue flowers: Estimate = -1.107, S.E. = 0.350, P =
248 0.002). Bees trained on the yellow flowers, also inspected these higher rewarding flowers significantly
249 more than non-flower regions (GLM, Estimate = -1.180, S.E. = 0.350, P = 0.001). Crucially, this was not
250 true for bees trained on the lower rewarding blue flowers (GLM, Estimate = -0.124, S.E. = 0.350, P =
251 0.724).

252 These results suggest that bee attention to rewards is increased as they learn about the flowers. They
253 also show that higher rewarding and lower rewarding flowers have slightly different effects. When
254 flowers have lower rewards, bumblebees continue their searching behaviour rather than focussing on
255 the rewarding flowers.



256

257 Figure 2. Bee visual search in Experiment 1. Top row: Proportion of time spent by the bees in different
 258 regions weighted by the area of each region. Bees were presented with A) lower reward (30% sucrose)
 259 high saliency blue flowers or B) higher reward (50% sucrose) lower saliency yellow flowers. Pink plots
 260 depict data from the first six choices and blue plots depict data from the last six choices of the training.
 261 Box plots depict the median and the first and third quartiles, the whiskers depict the largest and

262 smallest values that are within 1.5 times the interquartile range from the edge of the boxes. Dots
263 represent data from individual bees. C-F) Example visual search maps for two bees depicted as a top
264 view of the flight arena with rewarding and unrewarding flowers. Colours depict the inspection times
265 up to a maximum of 500 ms. Squares depict flower zones and the inner bound of the defined
266 inspection zones, white circles illustrate the outer bound of the inspection zones. Only one circle is
267 depicted in each figure here for ease of illustration. *R* = Rewarding flowers; *D* = Unrewarding flowers.
268 C) and D) depict examples for the first six choices of training for bees trained on lower reward and
269 higher reward flowers respectively. E) and F) depict the visual search during the last six choices of the
270 same two bees as in C) and D) respectively.

271 The number of correct choices made by the bees was best explained by a model that included the
272 training stage and the flower (and thus reward) type as predictors but not the experiment (first or
273 second). Bees were significantly less likely to make correct choices for blue flowers compared to
274 yellow flowers (GLMM, Estimate = -1.525, S.E. = 0.464, P = 0.001). This likely reflects innate biases of
275 the bees to blue but the bees were generally highly accurate. In the initial training stage of Experiment
276 1, bees made 79% correct choices to yellow flowers and 88% correct choices to blue flowers. The
277 accuracy of the very first choices was however 50% (4 out of 8 bees) for yellow flowers and 62.5% (5
278 out of 8 bees) for blue flowers. These proportions were not different from chance (binomial tests,
279 Yellow: P = 0.273, Blue: P = 0.219). In the later training stage, all bees were 100% accurate. This result
280 also highlights the differences found when analysing choices and visual search and how both analyses
281 complement each other.

282 Bees also made their last six choices faster than they made their first six choices (comparison with null
283 model: $\chi^2 = 26.216$, df = 1, P < 0.001: First six VS last six: Estimate = -1.004, S.E. = 0.154, Z = -6.51, P <
284 0.001). Including flower colour as a factor did not improve the model (comparison with model having
285 only training stage: $\chi^2 = 0.753$, df = 2, P = 0.686), suggesting that the effect was comparable in both

286 higher and lower reward flowers. Learning thus increases the speed of choices regardless of flower
287 reward level.

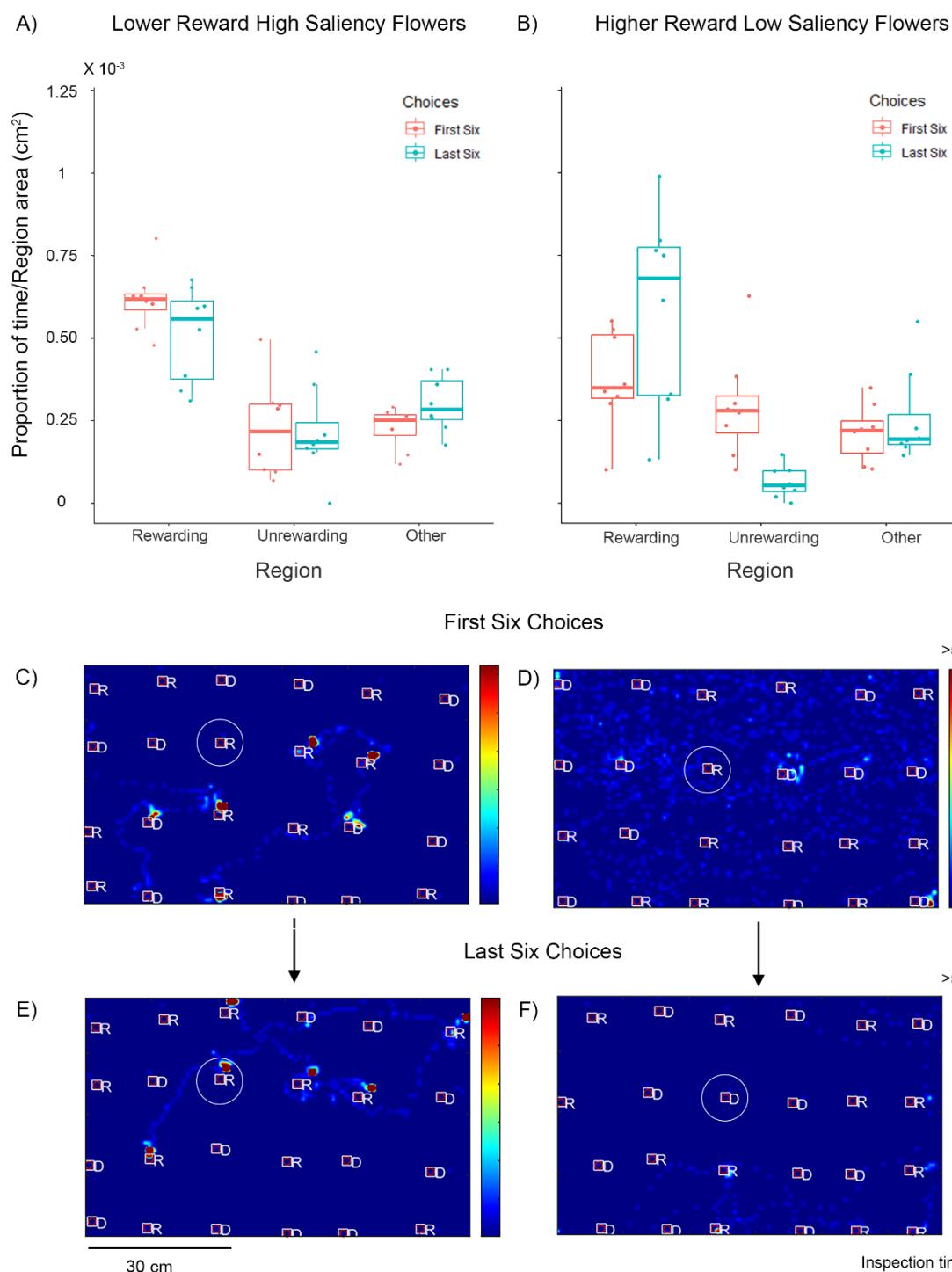
288 Experiment 2:

289 Unlike in Experiment 1, we did not find a main effect of training stage in this Experiment indicating
290 that the training did not here increase bee inspection of rewarding flowers (Yellow flowers: GLM,
291 Estimate = 0.384, S.E. = 0.244, P = 0.119, Blue flowers: GLM, Estimate = -0.218, S.E. = 0.244, P = 0.374,
292 Figure 3A and B). In the first training stage, the group of bees that had switched from lower to higher
293 rewarding flowers between experiments were more likely to inspect rewarding flowers compared to
294 the other group (GLM, Estimate = 0.586, S.E. = 0.244, P = 0.019, Figure 3A and B, pink plots). This
295 demonstrates the effect of the prior experience.

296 In this training stage, bees in both groups were significantly more likely to inspect rewarding flowers
297 compared to non-flower regions (Yellow flowers: GLM, Estimate = -0.547, S.E. = 0.244, P = 0.028, Blue
298 Flowers: GLM, Estimate = -1.02, S.E. = 0.244, P < 0.001, Figure 3A and B, pink plots). Bees with prior
299 experience of high rewards also were significantly more likely to attend to the now lower rewarding
300 flowers compared to unrewarding flowers (GLM, Estimate = -1.197, S.E. = 0.244, P < 0.001). For bees
301 in the other group this effect was not significant (GLM, Estimate = -0.272, S.E. = 0.244, P = 0.267).

302 There was a significant interaction effect showing that training increased attention to rewarding
303 flowers compared to unrewarding flowers when bees encountered higher rewarding yellow flowers
304 (GLM, Estimate = -1.819, S.E. = 0.351, P < 0.001). This was not true for the lower rewarding blue flowers
305 (GLM, Estimate = 0.423, S.E. = 0.351, P = 0.231). In the later training stage, there was no main effect
306 of reward value (GLM, Estimate = -0.002, S.E. = 0.317, P = 0.996). In this stage, bees inspected
307 rewarding flowers significantly more than unrewarding flowers (Yellow flowers: GLM, Estimate = -
308 2.063, S.E. = -0.325, P < 0.001, Blue flowers: GLM, Estimate = -0.802, S.E. = 0.327, P = 0.017). They also
309 inspected rewarding flowers significantly more than non-flower regions (Yellow flowers: GLM,
310 Estimate = -0.854, S.E. = 0.250, P = 0.001; Blue flowers: Estimate = -0.583, S.E. = 0.255, P = 0.025).

311 Training thus boosted attention to rewarding flowers for the bees that had previously encountered
 312 low rewarding flowers in Experiment 1 but now encountered higher rewarding flowers. This confirms
 313 our predictions of the effect of prior experience on visual search to subsequent rewards. However,
 314 our prediction was not true for bees that switched from higher to lower rewards. Here bees continued
 315 attending to the rewarding flowers even though they now encountered lower rewards.



316

317 Figure 3. Bee visual search in Experiment 2. The data here represent bees initially trained on one
318 reward type (higher or lower) in Experiment 1 (Figure 2) and subsequently trained on the opposite
319 reward type (lower or higher) in Experiment 2. Top row: Proportion of time spent by the bees in
320 different regions weighted by the area of each region. Bees were presented with A) lower reward (30%
321 sucrose) high saliency blue flowers or B) higher reward (50% sucrose) lower saliency yellow flowers.
322 Pink plots depict data from the first six choices and blue plots depict data from the last six choices of
323 the training. Box plots depict the median and the first and third quartiles, the whiskers depict the
324 largest and smallest values that are within 1.5 times the interquartile range from the edge of the
325 boxes. Dots represent data from individual bees. C-F) Example visual search maps for two bees
326 depicted as a top view of the flight arena with rewarding and unrewarding flowers. Colours depict the
327 inspection times up to a maximum of 500 ms. Squares depict flower zones and the inner bound of the
328 defined inspection zones, white circles illustrate the outer bound of the inspection zones. Only one
329 circle is depicted in each figure here for ease of illustration. *R* = Rewarding flowers; *D* = Unrewarding
330 flowers. C) and D) depict examples for the first six choices of training for bees trained on lower reward
331 and higher reward flowers respectively. E) and F) depict the visual search during the last six choices of
332 the same two bees as in C) and D) respectively. The example in C) and E) here is the same bee trained
333 in D) and F) in Figure 2 and the example in D) and F) here is the same bee trained in C) and F) in Figure
334 2.

335 The choices of the bees in Experiment 2 were again highly accurate. In the initial training stage of
336 Experiment 2, bees made 65% correct choices to yellow flowers and 92% correct choices to blue
337 flowers. In the later stage, bees from both groups chose the rewarding flower with 100% accuracy.
338 Contrary to what we observed in the first experiment, including flower colour made for a better
339 model, suggesting that the effect of learning on the time taken to perform six choices differed with
340 the reward quality (or colour) (Comparison with model including only training stage: $\chi^2 = 7.544$, df =
341 2, $P = 0.023$). There was no effect of training for bees that switched from higher rewarding flowers in
342 Experiment 1 to lower rewarding flowers in Experiment 2 (First six blue vs last six blue: Estimate = -

343 0.212, S.E. = 0.156, Z = -1.36, P = 0.175). However, when the bees changed from lower rewarding
344 flowers to higher rewarding flowers, their first six choices were faster than these of the other flower
345 colour group (First six blue VS First six yellow: Estimate = 0.439, S.E. = 0.196, Z = 2.24, P = 0.025) and
346 training further increased the speed at which these bees made their last six choices (First six yellow
347 VS last six yellow: Estimate = -0.630, S.E. = 0.222, Z = -2.84, P = 0.005).

348

349 **Discussion**

350 We tested whether learning modified visual search in two related experiments. We found that
351 learning resulted in an increase in the proportion of time spent by the bees around rewarding flowers
352 compared to unrewarding flowers and notably to non-flower regions, but this depended on reward
353 value and prior experience of rewards. Lower rewarding flowers led to greater visual search in areas
354 away from both rewarding and unrewarding flowers. This suggests that attention is more widely
355 distributed for lower rewarding flowers compared to higher rewarding flowers. It's also important to
356 note that since the unrewarding flowers were not rewarding, they would also provide the bees with
357 negative reinforcement. This would also partially explain the clear difference in attention between
358 rewards and unrewarding flowers. Bees with prior experience of higher reward were also more likely
359 to persist in attending to rewarding flowers even when they later encountered lower rewards. In our
360 experiments, we cannot disentangle the effect of reward quality and colour or saliency. However, the
361 fact that the high saliency of low reward flowers did not increase inspection of these flowers in
362 Experiment 1 indicates that our results are likely to reflect reward quality of the flowers rather than
363 their saliency.

364 In primates, eye movements are often used as proxies for overt attention (Schütz et al. 2011). The
365 duration spent looking at aspects of a scene have also been used to compute attentional maps
366 (Henderson and Hayes 2018). Our maps of inspection time perhaps best parallel these attentional
367 maps. Results from these studies of eye movements have shown that attention can be influenced by

368 several factors including saliency, reward value and the structure of a scene (Navalpakkam et al. 2010;
369 Schütz et al. 2011; Henderson and Hayes 2018). In non-primates, attentional limitations have most
370 often been studied in predators (Dukas and Kamil 2000; Dukas et al. 2002; Dukas 2004). There, findings
371 show that attentional resources are more focussed when searching for cryptic prey. Conversely,
372 hunger leads to praying mantises widening their search for possible prey (Bertsch et al. 2019; Pickard
373 et al. 2021). Our findings further argue that learning about reward value also influences attention in
374 bees, even when the rewarding flowers are not cryptic.

375 The median nectar sugar concentrations for flowers in Europe and globally is around 40% (Pamminger
376 et al. 2019). Our reward values therefore correspond to a higher than median reward (50%) and lower
377 than median reward (30%). The latter corresponds to the reward value of flowers in the 25th
378 percentile. However, bees in our experiment were pretrained on 50% sucrose solution. This could
379 have had some influence our results. Prior experience of a particular reward can influence future
380 behaviour of bees and ants in response to higher or lower rewards – a phenomenon called incentive
381 contrast (Bitterman 1975; Wendt et al. 2019). However, in our first experiment, we do not see a
382 difference between visual search to the two rewards in the first training stage, suggesting that the
383 experience of higher rewards during pre-training did not have an immediate effect. We do, however,
384 see an increase in visual search to the rewards compared to other areas, but only in the later training
385 stage, suggesting that as bees encounter higher (but not lower) rewards, they spend less time
386 searching and more time inspecting rewarding flowers. We find some evidence that prior experience
387 can influence visual search behaviour in line with ideas about incentive contrast – bees that experience
388 lower rewards in Experiment 1 had increased attention to higher rewards during Experiment 2 and
389 made faster choices. However, we did not find the converse for bees that switched from higher
390 rewards to lower rewards. One confounding factor here could be that the lower rewarding blue
391 flowers in Experiment 2 were more attractive due to the innate biases of bees. Alternatively,
392 continuous experience of high rewards from the pre-training and Experiment 1 might have boosted
393 the motivation of bees to a high level and persisted for a longer time.

394

395 Our results demonstrate the value of investigating search behaviour rather than focussing on flower
396 choice alone. Previous work (Nityananda and Chittka 2021) has looked at bee visual search when faced
397 with a choice between multiple rewarding flowers of different reward and saliency values. Research
398 there found that reward value biased inspection time at flowers, even for lower saliency flowers.
399 Other work has also shown that bees fly shorter distances after encountering rewarding flowers,
400 compared to non-rewarding flowers (Dukas and Real 1993). Our results further show how reward
401 value modifies bee visual search during learning. This difference in bee behaviour might specifically
402 reflect foraging behaviour in bumblebees. Bumblebees are less flower constant than honeybees
403 (Wells and Wells 1983; Waser 1986; Hill et al. 1997), sampling other flowers even when specializing
404 on a specific flower type – behaviours that have been called ‘minoring’ and ‘majoring’ respectively
405 (Heinrich 1976, 1979). Given that honeybees are more constant to flowers, it is possible that we might
406 find different results from honeybees with more focussed attention even for lower rewarding flowers.
407 Previous work on bee visual search has already shown differences between honeybees and
408 bumblebees. Honeybees show serial visual search, while bumblebees are capable of parallel visual
409 search – their visual search for a target is independent of the number of distractors (Spaethe et al.
410 2006; Morawetz and Spaethe 2012). Running similar experiments to ours with honeybees and other
411 bees might bring up further interesting differences in visual search and attention.

412 In addition, our results suggest that flowers that have higher concentration of reward are more likely
413 to have focussed attention from bumblebees. Bees that encounter flowers with lower rewards would
414 be expected to keep searching even as they visit the flowers. We might therefore expect flowers with
415 lower rewards to compensate for this loss of attention. One possibility is that these flowers might be
416 more salient than flowers with higher rewards. However, blue flowers which are salient in temperate
417 zones are actually the ones that are the most rewarding to bees (Giurfa et al. 1995). Another study in
418 Australia found no significant correlation between reward values and chromatic contrast (Shrestha et

419 al. 2020). Our results also show that merely having high saliency does not lead to focussed attention
420 as the high saliency lower reward flowers did not attract greater bee visual search. Flowers with lower
421 rewards might therefore be under selective pressure to either invest in multimodal cues (Kulahci et
422 al. 2008) or include secondary compounds in their nectar that might affect pollinator memory and
423 attention. Caffeine and nicotine have both been shown to have effects on bee learning (Wright et al.
424 2013; Couvillon et al. 2015; Baracchi et al. 2017; Arnold et al. 2021) and we would predict that they
425 should be more likely to be present in the nectar of flowers with lower concentrations of reward.

426 This study demonstrates the importance of investigating bee behaviour beyond flower choices.
427 Understanding how visual search and attention is influenced by a variety of factors could further
428 enhance our understanding of pollination ecology and bee cognition.

429 **Data Availability:** All code and data relevant to this paper are included as supplementary material to
430 this paper.

431 **Acknowledgements:** VN and TR are supported by a BBSRC David Phillips fellowship BB/S009760/1 to
432 VN. This work was also partly supported by a Marie Curie Incoming International Fellowship (PIIF-GA-
433 2009–253593) to VN.

434 **Author Contributions:** VN conducted the experiment. KT and VN analysed the videos. VN and TR ran
435 the statistical analyses and wrote the paper.

436 **Competing Interests:** The authors have no competing interests to declare that are relevant to the
437 content of this article.

438

439 **References**

440 Arnold SEJ, Dudenhöffer J-H, Fountain MT, et al (2021) Bumble bees show an induced preference for
441 flowers when primed with caffeinated nectar and a target floral odor. *Curr Biol* 4127–4131.

442 <https://doi.org/10.1016/j.cub.2021.06.068>

443 Avarguès-Weber A, Deisig N, Giurfa M, et al (2011) Visual cognition in social insects. *Annu Rev Entomol* 56:423–43. <https://doi.org/10.1146/annurev-ento-120709-144855>

445 Avarguès-Weber A, Giurfa M (2014) Cognitive components of color vision in honey bees: How
446 conditioning variables modulate color learning and discrimination. *J Comp Physiol A* 200:449–
447 461. <https://doi.org/10.1007/s00359-014-0909-z>

448 Avarguès-Weber A, Lachlan R, Chittka L (2018) Bumblebee social learning can lead to suboptimal
449 foraging choices. *Anim Behav* 135:209–214. <https://doi.org/10.1016/j.anbehav.2017.11.022>

450 Baracchi D, Marples A, Jenkins AJ, et al (2017) Nicotine in floral nectar pharmacologically influences
451 bumblebee learning of floral features. *Sci Rep* 7:1–8. <https://doi.org/10.1038/s41598-017-01980-1>

453 Baude M, Danchin É, Mugabo M, Dajoz I (2011) Conspecifics as informers and competitors: An
454 experimental study in foraging bumble-bees. *Proc R Soc B Biol Sci* 278:2806–2813.
455 <https://doi.org/10.1098/rspb.2010.2659>

456 Benard J, Stach S, Giurfa M (2006) Categorization of visual stimuli in the honeybee *Apis mellifera*.
457 *Anim Cogn* 9:257–70. <https://doi.org/10.1007/s10071-006-0032-9>

458 Bertsch DJ, Martin JP, Svenson GJ, Ritzmann RE (2019) Predatory behavior changes with satiety or
459 increased insulin levels in the praying mantis (*Tenodera sinensis*). *J Exp Biol* 222:jeb197673.
460 <https://doi.org/10.1242/jeb.197673>

461 Chittka L (1992) The colour hexagon: a chromaticity diagram based on photoreceptor excitations as a
462 generalized representation of colour opponency. *J Comp Physiol A* 170:533–543

463 Couvillon MJ, Al Toufailia H, Butterfield TM, et al (2015) Caffeinated forage tricks honeybees into
464 increasing foraging and recruitment behaviors. *Curr Biol* 25:2815–2818.
465 <https://doi.org/10.1016/j.cub.2015.08.052>

466 Dukas R (2004) Causes and consequences of limited attention. *Brain Behav Evol* 63:197–210.

467 <https://doi.org/10.1159/000076781>

468 Dukas R, Kamil AC (2000) The cost of limited attention in blue jays. *Behav Ecol* 11:502–506.

469 <https://doi.org/10.1093/beheco/11.5.502>

470 Dukas R, Real LA (1993) Effects of recent experience on foraging decisions by bumble bees.

471 *Oecologia* 94:244–246

472 Dukas R, Trans Soc Lond B Reuven Dukas PR, Dukas R (2002) Behavioural and ecological

473 consequences of limited attention. *Philos Trans R Soc Ser B Biol Sci* 357:1539–1547.

474 <https://doi.org/10.1098/rstb.2002.1063>

475 Giurfa M (2012) Visual Cognition in Honey Bees: From Elemental Visual Learning to Non-elemental

476 Problem Solving. In: *Honeybee Neurobiology and Behavior*. pp 471–484

477 Giurfa M, Nunez J, Chittka L, Menzel R (1995) Colour preferences of flower-naive honeybees. *J Comp*

478 *Physiol A* 177:247–259

479 Goulson D (2000) Are insects flower constant because they use search images to find flowers? *Oikos*

480 88:547–552. <https://doi.org/10.1034/j.1600-0706.2000.880311.x>

481 Heinrich B (1979) “Majoring” and “Minoring” by Foraging Bumblebees, *Bombus Vagans*: An

482 Experimental Analysis. *Ecology* 60:245–255. <https://doi.org/10.2307/1937652>

483 Heinrich B (1976) The Foraging Specializations of Individual Bumblebees. *Ecol Monogr* 46:105–128.

484 <https://doi.org/10.2307/1942246>

485 Henderson JM, Hayes TR (2018) Meaning guides attention in real-world scene images: Evidence from

486 eye movements and meaning maps. *J Vis* 18:1–18. <https://doi.org/10.1167/18.6.10>

487 Hill PSMM, Wells PH, Wells H (1997) Spontaneous flower constancy and learning in honey bees as a

488 function of colour. *Anim Behav* 54:615–27. <https://doi.org/10.1006/anbe.1996.0467>

489 Kulahci IG, Dornhaus A, Papaj DR (2008) Multimodal signals enhance decision making in foraging
490 bumble-bees. *Proc R Soc B Biol Sci* 275:797–802. <https://doi.org/10.1098/rspb.2007.1176>

491 Li L, MaBouDi H, Egertová M, et al (2017) A possible structural correlate of learning performance on
492 a colour discrimination task in the brain of the bumblebee. *Proc R Soc B Biol Sci* 284:20171323.
493 <https://doi.org/10.1098/rspb.2017.1323>

494 Lubbock SJ (1881) Observations on ants, bees, and wasps. IX. Color of flowers as an attraction to
495 bees: Experiments and considerations thereon. *Zool J Linn Soc* 16:110–121

496 Lunau K (1990) Colour saturation triggers innate reactions to flower signals: Flower dummy
497 experiments with bumblebees. *J Comp Physiol A* 166:827–834

498 Lunau K, Wacht S, Chittka L (1996) Colour choices of naive bumble bees and their implications for
499 colour perception. *J Comp Physiol A* 178:477–489. <https://doi.org/10.1007/BF00190178>

500 Morawetz L, Spaethe J (2012) Visual attention in a complex search task differs between honeybees
501 and bumblebees. *J Exp Biol* 215:2515–2523. <https://doi.org/10.1242/jeb.066399>

502 Navalpakkam V, Koch C, Rangel A, Perona P (2010) Optimal reward harvesting in complex perceptual
503 environments. *Proc Natl Acad Sci U S A* 107:5232–5237.
504 <https://doi.org/10.1073/pnas.0911972107>

505 Nityananda V, Chittka L (2021) Different effects of reward value and saliency during bumblebee
506 visual search for multiple rewarding targets. *Anim Cogn* 24:803–814.
507 <https://doi.org/10.1007/s10071-021-01479-3>

508 Nityananda V, Pattrick JG (2013) Bumblebee visual search for multiple learned target types. *J Exp
509 Biol* 216:4154–60. <https://doi.org/10.1242/jeb.085456>

510 Pickard SC, Bertsch DJ, Le Garrec Z, et al (2021) Internal state effects on behavioral shifts in freely
511 behaving praying mantises (*Tenodera sinensis*). *PLOS Comput Biol* 17:e1009618.

512 <https://doi.org/10.1371/journal.pcbi.1009618>

513 Ben-Tov M, Donchin O, Ben-Shahar O, Segev R (2015) Pop-out in visual search of moving targets in
514 the archer fish. *Nat Commun* 6:6476. <https://doi.org/10.1038/ncomms7476>

515 Bitterman M (1975) Incentive Contrast in Honey Bees. *Science* (80-) 192:380–382

516 Bond AB, Kamil AC (2002) Visual predators select for crypticity and polymorphism in virtual prey.
517 *Nature* 415:609–13. <https://doi.org/10.1038/415609a>

518 Brooks M, Kristensen K, van Benthem K, et al (2017) glmmTMB Balances Speed and Flexibility Among
519 Packages for Zero-inflated Generalized Linear Mixed Modeling. *R J* 9:378–400.
520 <https://doi.org/doi:10.32614/RJ-2017-066>.

521 Dukas R, Kamil AC (2000) The cost of limited attention in blue jays. *Behav Ecol* 11:502–506.
522 <https://doi.org/10.1093/beheco/11.5.502>

523 Gil M, De Marco RJ, Menzel R (2007) Learning reward expectations in honeybees. *Learn Mem*
524 14:491–496. <https://doi.org/10.1101/lm.618907>

525 Hartig F (2022) DHARMa: Residual Diagnostics for Hierarchical (Multi-Level / Mixed) Regression
526 Models. R package version 0.4.6. <http://florianhartig.github.io/DHARMa/>

527 Horowitz TS, Wolfe JM (2001) Search for multiple targets: remember the targets, forget the search.
528 *Percept Psychophys* 63:272–85

529 Nityananda V (2016) Attention-like processes in insects. *Proc R Soc B Biol Sci* 283:20161986.
530 <https://doi.org/10.1098/rspb.2016.1986>

531 Nityananda V, Chittka L (2021) Different effects of reward value and saliency during bumblebee
532 visual search for multiple rewarding targets. *Anim Cogn* 24:803–814.
533 <https://doi.org/10.1007/s10071-021-01479-3>

534 Orlowski J, Beissel C, Rohn F, et al (2015) Visual pop-out in barn owls: Human-like behavior in the

535 avian brain. *J Vis* 15:1–13. <https://doi.org/10.1167/15.14.4.doi>

536 Orlowski J, Ben-Shahar O, Wagner H, et al (2018) Visual search in barn owls : Task difficulty and
537 saccadic behavior. *J Vis* 18:1–13. <https://doi.org/10.1167/18.1.4>.

538 Pamminger T, Becker R, Himmelreich S, et al (2019) The nectar report: quantitative review of nectar
539 sugar concentrations offered by bee visited flowers in agricultural and non-agricultural
540 landscapes. *PeerJ* 7:e6329. <https://doi.org/10.7717/peerj.6329>

541 Posner MI (1980) Orienting of attention. *Q J Exp Psychol* 32:3–25

542 Saban W, Sekely L, Klein RM, Gabay S (2017) Endogenous orienting in the archer fish. *Proc Natl Acad
543 Sci* 114:7577–7581. <https://doi.org/10.1073/pnas.1700574114>

544 Wendt S, Strunk KS, Heinze J, et al (2019) Positive and negative incentive contrasts lead to relative
545 value perception in ants. *eLife* 8:1–22. <https://doi.org/10.7554/eLife.45450>

546 Riveros AJ, Gronenberg W (2012) Decision-making and associative color learning in harnessed
547 bumblebees (*Bombus impatiens*). *Anim Cogn* 15:1183–93. [012-0542-6](https://doi.org/10.1007/s10071-
548 012-0542-6)

549 Schütz AC, Braun DI, Gegenfurtner KR (2011) Eye movements and perception: A selective review. *J
550 Vis* 11:1–30. <https://doi.org/10.1167/11.5.1>

551 Shrestha M, Garcia JE, Burd M, Dyer AG (2020) Australian native flower colours: Does nectar reward
552 drive bee pollinator flower preferences? *PLoS One* 15:11–15.
553 <https://doi.org/10.1371/journal.pone.0226469>

554 Skorupski P, Döring TF, Chittka L (2007) Photoreceptor spectral sensitivity in island and mainland
555 populations of the bumblebee, *Bombus terrestris*. *J Comp Physiol A Neuroethol Sensory,
556 Neural, Behav Physiol* 193:485–494. <https://doi.org/10.1007/s00359-006-0206-6>

557 Solvi C, Zhou Y, Feng Y, et al (2022) Bumblebees retrieve only the ordinal ranking of foraging options

558 when comparing memories obtained in distinct settings. *eLife* 11:1–12.

559 <https://doi.org/10.7554/eLife.78525>

560 Spaethe J, Tautz J, Chittka L (2006) Do honeybees detect colour targets using serial or parallel visual

561 search? *J Exp Biol* 209:987–993. <https://doi.org/10.1242/jeb.02124>

562 Turner CH (1910) Experiments on color-vision of the honey bee. *Biol Bull* 18:257–279.

563 <https://doi.org/10.1086/bblv18n5p213>

564 von Frisch K (1914) Der Farbensinn und Formensinn der Biene. *Zool Jahrb Abt Allg Zool Physiol Tiere*

565 37:1–238

566 Waser NM (1986) Flower Constancy: Definition, Cause, and Measurement. *Am Nat* 127:593–603

567 Wells H, Wells PH (1983) Honey Bee Foraging Ecology : Optimal Diet , Minimal Uncertainty or

568 Individual Constancy ? *J Anim Ecol* 52:829–836

569 Wright GA, Baker DD, Palmer MJ, et al (2013) Caffeine in floral nectar enhances a pollinator's

570 memory of reward. *Science* (80-) 339:1202–1204. <https://doi.org/10.1126/science.1228806>

571 **Supplementary Information**

572 ESM_1: CSV file with all data supporting this paper.

573 ESM_2: R code for generating the boxplots included in this paper.

574 ESM_3: R code for the analysis of choice durations

575 ESM_4: R code for the analysis of visual search in Experiment 1

576 ESM_5: R code for the analysis of visual search in Experiment 2

577 ESM_6: R code for the analysis of bee choices

578