

1 Consistent variations in personality traits and their potential for genetic improvement of
2 biocontrol agents: ~~in the biocontrol agent~~ *Trichogramma evanescens* as a case study

3
4 Silène Lartigue^{a,b,c}, Myriam Yalaoui^b, Jean Belliard^b, Claire Caravel^b, Louise Jeandroz^b,
5 Géraldine Groussier^b, Vincent Calcagno^b, Philippe Louâpre^c, François-Xavier Dechaume-
6 Moncharmont^d, Thibaut Malausa^b and Jérôme Moreau^{c, e}

7
8 ^a ENGREF AgroParisTech, Paris, France

9
10 ^b UMR Institut Sophia Agrobiotech, INRAE, UCA, CNRS, 06903 Sophia Antipolis, France

11
12 ^c UMR CNRS 6282 Biogéosciences, Université Bourgogne Franche-Comté, 6 Boulevard
13 Gabriel, 21000 Dijon, France

14
15 ^d Univ Lyon, Université Claude Bernard Lyon 1, CNRS, ENTPE, UMR 5023 LEHNA, 69622
16 Villeurbanne, France

17
18 ^e Centre d'Études Biologiques de Chizé, UMR 7372, CNRS & La Rochelle Université, 79360
19 Villiers-en-bois, France

20
21 * **Corresponding author:** S. Lartigue, UMR Institut Sophia Agrobiotech, INRAE, UCA,
22 CNRS, Sophia Antipolis, France. E-mail address: silene.lartigue@gmail.com

23 **Abstract**

24 Improvements in the biological control of agricultural pests require improvements in the
25 phenotyping methods used by practitioners to select efficient biological control agent (BCA)
26 populations in industrial rearing or field conditions. Consistent inter-individual variations in
27 behaviour (i.e. animal personality) probably affect BCA efficiency, but have never been taken
28 into account in the development of phenotyping methods, despite having characteristics useful
29 for phenotyping: repeatable (by definition), often heritable, etc. We developed a video-tracking
30 method targeting animal personality traits and evaluated the feasibility of its use for genetic
31 improvement in the BCA *Trichogramma evanescens*, by phenotyping 1,049 individuals from
32 24 isogenic lines. We found consistent individual variations in boldness, activity and
33 exploration. Personality differences between the 24 isogenic lines suggested a genetic origin of
34 the variations in activity and exploration (broad-sense heritability estimates of 0.06 to 0.11) and
35 revealed a trade-off between exploration and fecundity.

36 **Key words:** biocontrol, animal personality, *Trichogramma*, genetic improvement, intraspecific
37 variation, behavioural ecology, behavioural syndrome, genetic correlations, pace-of-life

38

39 **Introduction**

40 The demand for more sustainable agriculture is increasing worldwide (Godfray et al.,
41 2010; Willer & Lernoud, 2019). Various elements can be used in the development of sustainable
42 strategies, and biological control (BC) is one such element that is currently attracting
43 considerable attention (van Lenteren 2012). Most BC methods are based on the choice, rearing
44 and introduction of biological control agent (BCA) populations able to control the target pests
45 (Eilenberg, Hajek, & Lomer, 2001). Choosing the right BCA is key to the success of pest
46 regulation programmes and is based on (i) the ability of the BCA to control pest populations in
47 the field, (ii) its potential to adapt to the release environment, (iii) its expected impact on local
48 biodiversity, and (iv) the feasibility of mass-rearing and storing the BCA in industrial conditions
49 (Briese, 2000; Kruitwagen, Beukeboom, & Wertheim, 2018; Sforza, 2010). The identification
50 of BCA species or populations with as many of the desired features as possible is time-
51 consuming and complex, particularly given that the choice of non-indigenous species before
52 use as BCAs is constrained by increasingly strict regulations for the protection of biodiversity
53 (Lommen, Jong, & Pannebakker, 2017).

54 Phenotyping is key for (i) the efficient characterisation of traits related to the desirable
55 features of BCAs listed above, (ii) smart choices of BC taxa when screening the available
56 natural enemy diversity and (iii) the management of phenotypic evolution in industrial contexts
57 involving rearing procedures and quality control (Kruitwagen et al., 2018; Lommen et al.,
58 2017). However, the phenotyping methods currently used in the choice of BCAs or for quality
59 control are mostly low-throughput and based on single proxies of fitness, such as predation or
60 parasitism rate, size, sex ratio, longevity, or developmental rate (Hopper, Roush, & Powell,
61 1993; Prezotti, Parra, Vencovsky, Coelho, & Cruz, 2004; Roitberg, Boivin, & Vet, 2001; Smith,
62 1996). These proxies are intuitively correlated with fitness under laboratory conditions, but
63 their actual relevance for biocontrol, in industrial mass-rearing or field conditions, remains a

64 matter of debate (Lommen et al., 2017; Roitberg et al., 2001). This situation calls for drastic
65 improvements in the phenotyping capacities of the community involved in BC research and
66 innovation.

67 Behavioural traits are among the most promising of the traits to which more attention
68 could be paid in BCA phenotyping procedures. Most behavioural traits are likely to affect the
69 performance of BCA both during industrial mass rearing and in the field (Roitberg, 2007;
70 Wajnberg, 2009; Wajnberg, Roitberg, & Boivin, 2016). Indeed, studies of BCA behavioural
71 traits have suggested that these traits could (i) facilitate the selection of BCAs that are specific
72 to the targeted pest, (ii) improve release strategies (through studies of the BCA response to pre-
73 release handling or BCA mating behaviour, for example), and (iii) predict the efficiency of
74 target pest suppression by the BCA (Mills & Kean, 2010). However, there have been few
75 studies of BCA behavioural traits for the development of phenotyping methods, and behaviour
76 has been largely neglected by those using BC (Wajnberg, Bernstein, & Alphen, 2008).

77 As a consequence, the current state-of-the-art for insect behavioural studies displays
78 several key limitations. The first limitation is the lack of diversity of possible target traits for
79 phenotyping. Indeed, although many studies have focused on traits relating to foraging
80 behaviour (Lirakis & Magalhães, 2019; Mills & Wajnberg, 2008), tools for measuring other
81 aspects of behaviour remain scarce. A second limitation is the insufficient focus on the
82 intraspecific variation of traits. Such variation has been comprehensively investigated for only
83 a limited number of BCA species and a limited number of traits (Kruitwagen et al. 2018; Lirakis
84 and Magalhães 2019), but see however (Dumont, Aubry, & Lucas, 2018; Dumont, Réale, &
85 Lucas, 2017; Nachappa, Margolies, Nechols, & Campbell, 2011; Nachappa, Margolies,
86 Nechols, & Morgan, 2010). This situation is detrimental because the investigation of only a
87 fraction of the available intraspecific variability makes it difficult to identify the populations
88 displaying the highest performance for biocontrol, and prevents the development of efficient

89 genetic improvement programmes based on selective breeding and controlled evolution
90 (Wajnberg 2004; Bolnick et al. 2011; Lommen et al. 2017; Kruitwagen et al. 2018, Lirakis and
91 Magalhães 2019). A third limitation is the reliance of most choices in BC exclusively on
92 comparisons between average trait values for species or populations (Lommen et al., 2017).
93 Published studies have suggested that individual variation can affect the characteristics of the
94 population thought to be important for BC (Biro & Stamps, 2008; Michalko, Pekár, & Entling,
95 2019; Réale, Reader, Sol, McDougall, & Dingemane, 2007; Wolf & Weissing, 2012).

96 One way to overcome each of these three limitations would be to apply approaches used
97 in the field of animal personality to BC. Indeed, these approaches provide a framework offering
98 (i) sets of behavioural traits rarely studied in BC and displaying features (repeatability,
99 heritability) that make them good candidates for use in genetic improvement for BC, and (ii)
100 phenotyping methods suitable for analyses of intraspecific variation, including inter-individual
101 variation. Animal personality research focuses on inter-individual differences in behaviour that
102 are consistent over time and context (Dingemane, Kazem, Reale, & Wright, 2009; Denis Réale
103 et al., 2007). Interest in animal personality has increased over the last few decades, and studies
104 have been performed on diverse taxa, including insects (Amat, Desouhant, Gomes, Moreau, &
105 Monceau, 2018; Bell, Hankison, & Laskowski, 2009; Dingemane et al., 2009; Gosling, 2001;
106 Kralj-fiser & Schuett, 2014; Mazué, Dechaume-Moncharmont, & Godin, 2015; Monceau et al.,
107 2017; Denis Réale et al., 2007; Sih, Bell, & Johnson, 2004; van Ooers & Sinn, 2011) and, more
108 specifically, insects used as BC agents (Gomes, Desouhant, & Amat, 2019; Michalko et al.,
109 2019). Réale et al. (2007) described five main categories of personality traits: boldness,
110 exploration, activity, aggressiveness and sociability. Boldness represents an individual's
111 reaction to a risky but not new situation. Exploration is defined as an individual's reaction to a
112 new situation. Activity reflects the general level of activity of an individual. Finally, in a social
113 context, aggressiveness corresponds to an individual's agonistic reaction to his conspecifics,

114 and sociability provides information on an individual's reaction to the presence or absence of
115 con-specifics. Personality traits have been shown to be correlated with traits relevant for pest
116 control, such as foraging capacity, fecundity, growth, survival (Biro & Stamps, 2008), dispersal
117 ability (Cote, Fogarty, Weinersmith, Brodin, & Sih, 2010) and insecticide resistance (Royauté,
118 Buddle, & Vincent, 2014). These traits are probably, therefore, of interest in the context of BC.
119 Moreover, personality traits are repeatable, by definition, and can be heritable (Dochtermann,
120 Schwab, & Sih, 2014; Denis Réale et al., 2007; Stirling, Reale, & Roff, 2002), making them
121 suitable tools for genetic improvement. From a methodological point of view, animal
122 personality provides valuable information for the design of phenotyping and genetic
123 improvement strategies in BC. Indeed, animal personality studies are based on standardised
124 methods designed to measure inter-individual variation and to investigate correlations between
125 traits (e.g. by looking for behavioural syndromes) (Denis Réale et al., 2007; Sih et al., 2004).
126 This is particularly relevant to the objective of selecting several combined BC traits rather than
127 a single trait, as recently recommended by Lommen et al. (2017) and Kruitwagen et al. (2018).
128 The investigation of correlations between traits is also important, to detect trade-offs that may
129 constrain genetic improvement programmes or affect BC traits if mass-rearing causes
130 uncontrolled trait selection (Mackauer, 1976).

131 In this study, we assessed the potential for BCA phenotyping based on concepts and
132 methods used in the field of animal personality. We used the egg parasitoid *Trichogramma*
133 *evanescens* Westwood, 1833 (Hymenoptera: Trichogrammatidae) as a test species.
134 *Trichogramma* micro-wasps are used worldwide in augmentative BC against lepidopteran pests
135 (Hassan, 1993; van Lenteren, 2012). Their economic importance (Robin & Marchand, 2020;
136 Thibierge, 2015) justifies investments in research and development aiming to improve their
137 genetic potential. Our aims were (i) to determine whether behavioural traits meeting the criteria
138 of personality traits could be measured in these micro-wasps of approximately 0.5 mm in

139 length; (ii) to investigate the relationships between personality traits and traits classically
140 measured on BCAs in the laboratory, and (iii) to determine whether personality traits could be
141 used in genetic improvement strategies for BCAs. We developed a method based on the video-
142 tracking and measuring, at individual level, of multidimensional behavioural traits relating to
143 boldness, activity and exploration. We investigated the relationship between these behavioural
144 traits and further tested whether these traits were related to individual fitness traits relevant to
145 mass rearing (offspring number, longevity, tibia length). We then compared the traits between
146 24 near-isogenic strains, to obtain a first insight into the broad-sense heritability of these traits.
147 We looked for genetic correlations potentially constraining the use of these traits for genetic
148 improvement.

149 **Methods**

150 ***Laboratory rearing of *T. evanescens****

151 We used 24 near-isogenic lines (hereafter referred to as “lines”) of *Trichogramma*
152 *evanescens*. Lines were created from inbred crosses in populations established from individuals
153 sampled in different parts of France (geographic origins detailed in Table 7 in the appendix),
154 from 2010 to 2016, and reared in the laboratory at 18 ± 1 °C, $70 \pm 10\%$ RH and 16:8 h L:D
155 (details of the protocol followed to create the lines are provided in the appendix). Genetic
156 diversity within lines was below 1.1 alleles per locus at 19 microsatellite loci (unpublished
157 data), and individuals within lines were considered genetically identical. We created two
158 sublines for each line (Lynch & Walsh, 1998), to disentangle the confounding effects of rearing
159 tubes and lines (which may be caused by maternal effects). We considered variation between
160 lines to be of genetic origin, and variation within lines to be of environmental origin. We reared
161 *Trichogramma evanescens* individuals on sterilised *Ephestia kuehniella* Zeller 1879
162 (Lepidoptera: Pyralidae) eggs, renewed every 10 days, at 25.5 ± 1 °C, $70 \pm 10\%$ RH and 16:8

163 h L:D (Schöller & Hassan, 2001). We kept populations in glass tubes (height: 73 mm, diameter:
164 11 mm), and fed adults with honey *ad libitum*.

165 ***Measurement of variables***

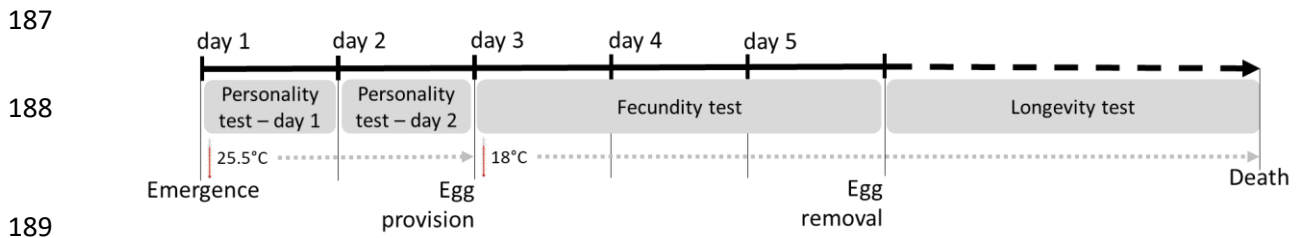
166 *General experimental design*

167 The following experimental design was used to measure phenotypic traits in
168 *Trichogramma* females (Figure 1). We used mated *T. evanescens* females that had mated
169 emerged within the last 24 hours, randomly chosen from each line. We checked the physical
170 integrity of these females, which were isolated in glass tubes before the beginning of the
171 experiment (height: 73 mm, diameter: 11 mm) and fed with honey, *ad libitum*. On the first two
172 days, we assessed the behavioural traits of the females. We estimated the number of offspring
173 on days 3 to 5, and longevity from day 6. The experiment lasted from May to July 2019 (about
174 six generations of *T. evanescens*), and was split into 17 experimental sessions, in each of which,
175 we used three females per line. The physiological, developmental and behavioural traits of
176 *Trichogramma* wasps, and of *T. evanescens* in particular, are dependent on temperature (Ayvaz,
177 A., Karasu, E., Karabörklü, S., Tunçbilek, 2008; Schöller & Hassan, 2001). Moreover, as
178 Suverkropp et al. (2001) showed that *T. brassicae* individuals have similar levels of activity
179 throughout the day at temperatures of about 25 °C or higher, we assumed that our *T. evanescens*
180 individuals had similar responses to temperature throughout the day. Therefore, we performed
181 the behavioural experiments at 25.5 ± 1 °C, $70 \pm 10\%$ RH. We then measured female longevity
182 and offspring number at 18 ± 1 °C, $70 \pm 10\%$ RH, to ensure that the females would live long
183 enough for the final stages of the experiment (Cônsoi & Parra, 1995; Schöller & Hassan, 2001).

184

185

186



187
188
189
190 **Figure 1.** Overview of the experimental design, for one session.

191 *Behavioural trait measurement*

192 We observed individuals in an arena composed of two sheets of glass (24 cm x 18 cm),
193 one for the floor and one for the ceiling. The 2 mm space between them was maintained by
194 walls made of a black rubber sheet. We placed this arena on an LCD screen (Samsung© 28”
195 LED, 3840*2160 pixels, 60 Hz) ~~displaying which was used to display~~ a white circle with a
196 diameter of 5.5 cm on a dark background (Figure 2.a). The LCD screen was turned on one hour
197 before the beginning of the experiment, to ensure that a stable temperature of 25.5 ± 1 °C was
198 achieved in the area. The conditions in the growth chamber in which the experimental design
199 was set up were as follows: 22.5 ± 1 °C and $70 \pm 10\%$ RH. We used a fine paintbrush to
200 introduce a randomly chosen female into the centre of the arena while the screen was showing
201 a white background. The glass ceiling was replaced, and we then switched to a background with
202 a white circle on a dark background, with the female positioned in the middle of the white circle.
203 We observed the behaviour of the female for 90 seconds, with video recording at 25 frames per
204 second (with a resolution of 1080 p), with a Nikon© D750 camera (Figure 2.a).

205 We then analysed the videos files, determining the 2D spatial position (*x-y* coordinates)
206 and body orientation (in radians) of the female on each frame, with C-trax software (Branson,
207 Robie, Bender, Perona, & Dickinson, 2009). We independently determined the exact position
208 of the border between the white circle and the black background with ImageJ software
209 (Abràmoff, Magalhães, & Ram, 2004). We thus defined regions of interest of 0.5 cm on either
210 side of the border, for investigation of the behaviour of the insect near the border (Figure 2.b).

211 We imported the C-trax and ImageJ output files into R software v.3.6.1 (R Core Team 2019)
212 and cleaned our data to remove tracking artefacts. We used the “trajr” package (Mclean &
213 Volponi, 2018) to calculate speed and sinuosity, in each region of interest. We calculated seven
214 variables we considered to be linked to three personality traits — boldness, exploration and
215 activity — as defined by Réale et al. (2007). As we measured all the variables linked to the
216 three personality traits in the same arena (for feasibility reasons, considering the lifespan of
217 individuals in particular), we decided to measure each variable set linked to each personality
218 trait on a different area of the arena to increase their extent of independence.

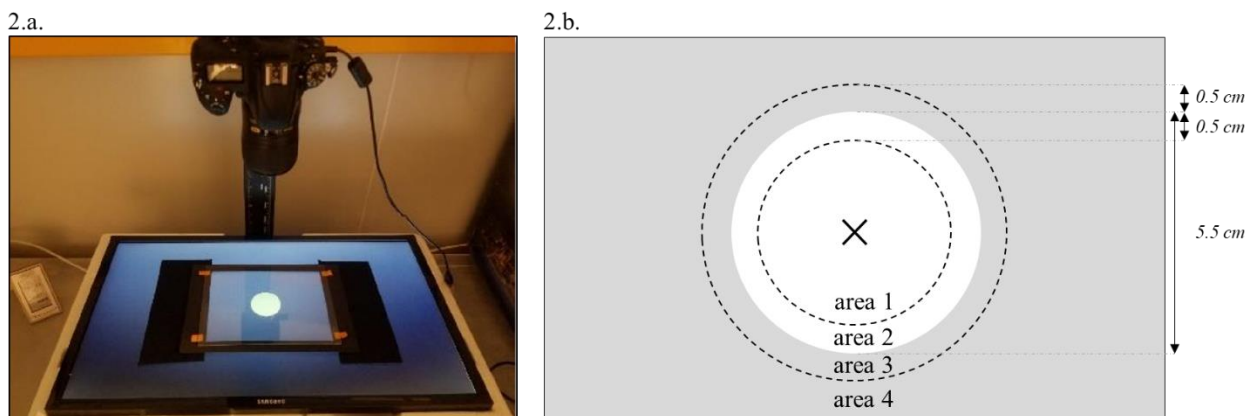
219 Boldness is the reaction of the individual to a risky situation (Réale et al., 2007). We
220 estimated boldness by measuring three variables. The first was the time until the female first
221 entered the dark area (area 3 in Figure 2.b). Higher values indicated that the female took longer
222 to cross the border, which we interpreted as meaning that the female was less bold. The second
223 and third variables were the absolute difference in speed between areas 2 and 1 (Figure 2.b) and
224 the absolute difference in sinuosity between areas 2 and 1 (Figure 2.b). We considered high
225 values for these two variables to indicate a larger change in behaviour at the border, which we
226 interpreted as meaning that the female was more affected by the border and was, therefore, less
227 bold.

228 Exploration represents the individual's reaction to a new environment (Réale et al.,
229 2007). Exploration was estimated in area 1 as (1) the total area explored per unit time, and (2)
230 the mean sinuosity of the pathway (Figure 2.b). For this variable, we hypothesised that the
231 females with the most winding pathways remained closer to their release point, indicating a
232 lower level of exploration.

233 Finally, we measured activity in area 4, ~~so exploration and activity were measured in~~
234 ~~different areas of the experimental arena.~~ Activity was estimated as (i) the proportion of time
235 the female was considered to be active (with a speed of more than 0.01 centimetres per second),

236 referred to hereafter as “activity rate”, and (ii) mean speed (Figure 2.b), considering faster
237 movement to be indicative of a higher level of activity.

238 We estimated the repeatability of measurements, by conducting two observations per
239 female, with 24 hours between the two measurements, a time interval corresponding to 20% of
240 the mean lifespan of this species. Females were tested in a random order on day 1, and then in
241 the same order on day 2, to ensure that all individuals had exactly the same time interval
242 between two measurements. Between behavioural experiments, each female was placed back
243 in its glass tube and fed with honey, *ad libitum*, in a growth chamber at 25.5 ± 1 °C, $70 \pm 10\%$
244 RH and 16:8 h L:D. Behavioural trait measurements were obtained for 776 to 996 females in
245 total from the 24 lines.



246
247 **Figure 2.** Experimental set-up of the behavioural experiment. Fig. 2.a. shows a photograph of the experimental
248 setup: the LCD screen displaying the white circle on a dark background, the arena and the Nikon® D750 camera
249 above. Fig. 2.b. represents the defined areas of the arena. The grey shading corresponds to the dark background,
250 the white part indicates the white circle, and the dark cross is the site at which the female was placed at the start
251 of the experiment. The dotted lines represent the virtual borders defined between areas 1 and 2, and between areas
252 3 and 4. The three variables we measured to estimate boldness were (i) the first time until the female first entered
253 the dark area (area 3), (ii) the absolute difference in speed between areas 2 and 1, and (iii) the absolute difference
254 in sinuosity between areas 2 and 1. Both variables we used to estimate exploration (the total area explored per unit
255 time and the mean sinuosity of the pathway) were measured in area 1. Finally, both variables we used to estimate

256 activity (the proportion of time the female was considered to be active and the mean speed) were measured in area
257 4, so exploration and activity were measured in different areas of the experimental arena.

258 *Offspring number, longevity and tibia length measurement*

259 After the second day of behavioural observation, females were kept in their glass tubes
260 at 18 ± 1 °C, $70 \pm 10\%$ RH and 16:8 h L:D and fed with honey, *ad libidum*. We provided each
261 female with a piece of paper 4.50 cm x 0.85 cm in size, covered with *E. kuehniella* eggs, *ad*
262 *libidum*. *E. kuehniella* eggs were removed 72 hours later and placed in conditions of 25.5 ± 1
263 °C, $70 \pm 10\%$ RH and 16:8 h L:D. Three days later, we counted the number of parasitised eggs
264 (considered as black eggs), to estimate the size of the progeny of each female over a period of
265 72 hours, providing a proxy for female fitness. From day 6, we measured female longevity (the
266 females were still kept in the same individual tubes with honey, but with no *E. kuehniella* eggs,
267 at 18 ± 1 °C, $70 \pm 10\%$ RH and 16:8 h L:D). Tubes were checked every day at 5 p.m., until the
268 death of the female. Dead females were conserved in ethanol, for subsequent measurement of
269 tibia length on a micrograph (obtained with an Axioplan II microscope), with ImageJ software
270 (Abràmoff et al., 2004). Images were acquired at the Microscopy Platform of Sophia
271 Agrobiotech Institute, INRA, UNS, CNRS, UMR 1355-7254, INRA PACA, Sophia Antipolis.
272 Not all individuals lived long enough for all the phenotypic measurements to be made. We
273 therefore collected progeny measurements for 929 females, longevity measurements for 655
274 females and tibia size measurements for 959 females, from all 24 lines.

275 *Data analysis*

276 We used the R software v.3.6.1 for all statistical analyses. For each variable, we first
277 fitted a linear mixed model with the lme4 package (Bates, Maechler, Bolker, & Walker, 2015),
278 with individual, line, subline and session as random effects. For each variable, data
279 transformations were chosen after graphical inspection of the distribution of model residuals,
280 estimated with the “simulateResiduals” function of the DHARMA package (Hartig, 2019). We

281 performed logarithmic transformations for all behavioural variables except for the area explored
282 within area 1. We addressed several questions regarding the data, and the data analysis for each
283 of these questions is presented below.

284 *Are the measured behavioural traits repeatable?*

285 We first estimated the repeatability of the behavioural traits measured with generalised
286 linear mixed models, using the rptR package (Stoffel, Nakagawa, & Schielzeth, 2017). The
287 “rptGaussian” function of the rptR package was used to provide repeatability estimates. As
288 repeatability can be defined as the proportion of variation explained by between-individual
289 variation (Nakagawa & Schielzeth, 2010), we included only two random effects in these
290 models: individual (assuming that the effects of line and subline on variation were included in
291 the individual effect) and session, with individual as a grouping factor. In subsequent analyses,
292 we considered only traits that were significantly repeatable.

293 *Do the measured traits identify individual behavioural strategies?*

294 Based on methods generally used in animal personality studies, we first investigated
295 correlations between behavioural traits and then summarized the data by principal component
296 analysis (PCA). We first obtained a single value for each trait for each individual, by extracting,
297 from the linear mixed model described above, linear predictors for each individual, with the
298 “ranef” function of the lme4 package. We used these values to measure the phenotypic
299 correlation between traits, by calculating Spearman’s rank correlation coefficients, to determine
300 whether individuals adopted different strategies, or whether it was possible to describe
301 behavioural syndromes. We estimated bootstrapped 95% confidence intervals from 1000
302 bootstraps, to assess the significance of the Spearman's rank correlation coefficients obtained
303 (Nakagawa & Cuthill, 2007), using the “spearman.ci” function of the RVAideMemoire package
304 (Hervé, 2020). *P*-values were adjusted by the false discovery rate method (Benjamini &

305 Hochberg, 1995). We then performed PCA with the “PCA” function of the FactoMineR
306 package (Le, Josse, & Husson, 2008), using both values obtained for each individual (days 1
307 and 2, when available). We estimated two synthetic personality scores based on the first two
308 axes of the PCA. We used the “fviz_pca_biplot” function of the factoextra package
309 (Kassambara & Mundt, 2019) to obtain a graphical representation of the correlation between
310 repeatable behavioural traits and the distribution of individual values along the two first axes
311 of the PCA.

312 *Are the measured traits correlated with fitness-related traits?*

313 We studied the correlation between behavioural and fitness-related traits, using the same
314 linear mixed model as described in the introduction to this section. We extracted linear
315 predictors (using the “ranef” function of the lme4 package (Bates et al., 2015)) for each
316 individual and each personality score from this model. We assessed the correlation between the
317 linear predictors of these personality traits and scores, and offspring number, body size and
318 longevity, by calculating Spearman’s rank correlation coefficients. We estimated bootstrapped
319 95% confidence intervals to assess significance of the Spearman's rank correlation coefficients
320 obtained, with the same R function and method as described above. *P*-values were adjusted by
321 the false discovery rate method.

322 *Are the measured traits heritable?*

323 We sought to establish a first estimate of broad-sense heritability for each trait. To this
324 end, we followed the simple design proposed by Lynch and Walsh (1998) for clonal
325 populations, and approximated the proportion of the variance explained by genetic factors with
326 an estimate of the proportion of variance explained by the line effect in our generalised linear
327 mixed models. This estimate was obtained with the “rptGaussian” function of the rptR package
328 (Stoffel et al., 2017), with models including line, subline, individual and session as random
329 effects, and line as a grouping factor.

330 *Do personality traits differentiate the isogenic lines?*

331 We compared the personality scores of the 24 lines, taking into account variation due to
332 individual, subline and session effects. With the values of each personality score extracted from
333 the PCA (see above), we first fitted a linear mixed-effects model with the “lmer” function of
334 the lme4 package (Bates et al., 2015), with line as a fixed effect and individual, subline and
335 session effects as random effects. We performed a Tukey all-pairs comparison on lines with the
336 “glht” function of the multcomp package (Hothorn, Bretz, & Westfall, 2008). We graphically
337 represented the distribution of each line along the two personality scores, for the same PCA as
338 described above, estimated from individual values. We then used the “plot.PCA” function of
339 the FactoMineR package to represent only mean point values for each line on the graph.

340 *Are personality traits genetically correlated with fitness-related traits?*

341 We investigated the genetic correlation between genetic traits, using the same linear
342 mixed model as described in the introduction to this section. We first extracted linear predictors
343 for each line and trait, with the “ranef” function of the lme4 package. We then used these values
344 to calculate Spearman’s rank correlation coefficients. We estimated bootstrapped 95%
345 confidence intervals, to assess significance of the Spearman's rank correlation coefficients, and
346 adjusted the p -values as described above.

347 **Results**

348 *Are the measured behavioural traits repeatable?*

349 Repeatability estimates for the seven behavioural traits ranged from 0.04 to 0.35 (Table
350 1). The repeatability estimates had confidence intervals excluding zero for all traits except for
351 “time to first crossing of the border between the white and black areas” (Table 1). Only
352 repeatable traits were considered in the subsequent analysis.

353 **Table 1.** Estimated repeatability (R) and 95% confidence intervals (between square brackets) for behavioural traits.
 354 Repeatable traits (R-value in bold type) were used to estimate personality scores.

Personality trait category	Variable assessed	R [95% CI]
Activity	Mean speed in area 4	0.35 [0.29; 0.40]
	Activity rate in area 4	0.08 [0.01; 0.14]
Boldness	Change of speed in the border area (area 2)	0.10 [0.04; 0.17]
	Change of sinuosity in the border area (area 2)	0.12 [0.04; 0.19]
	Time to first crossing of the white/black border	0.04 [0.00; 0.11]
Exploration	Sinuosity in area 1	0.24 [0.17; 0.30]
	Area explored in area 1	0.18 [0.12; 0.24]

355

356 *Do the measured traits identify individual behavioural syndromes?*

357 All repeatable variables were correlated with at least one other variable (Table 2),
 358 indicating the existence of a behavioural syndrome. We combined these six variables into two
 359 personality scores based on the first two axes of a PCA, which accounted for 56.8% of the
 360 variance (Table 3). The first axis (personality score 1, PC1) was positively correlated with the
 361 “area explored in area 1” and inversely correlated with “sinuosity in area 1” and with the
 362 “change of sinuosity in the border area 2” (Table 3). Highly positive values of PC1
 363 corresponded to a high exploration score (Figure 3). The second axis (personality score 2, PC2)
 364 and correlated mostly with “mean speed in area 4”, “activity rate in area 4” and the “change of
 365 speed in border area 2” (Table 3). High positive values of PC2 correspond to high activity scores
 366 (Figure 3).

367

368

369

370

371

372

373 **Table 2.** Phenotypic correlation between behavioural variables, with Spearman's rank correlation coefficient Rho
 374 and 95 percent confidence intervals (between square brackets), based on a number of individual values from N =
 375 977 to N = 1009. Correlation coefficients with confidence intervals excluding zero are shown in bold, and
 376 correlation coefficients remaining significantly different from zero after Benjamini and Hochberg correction are
 377 indicated with an asterisk. The personality trait category to which each variable belongs is indicated in brackets:
 378 activity (A), boldness (B) and exploration (E).

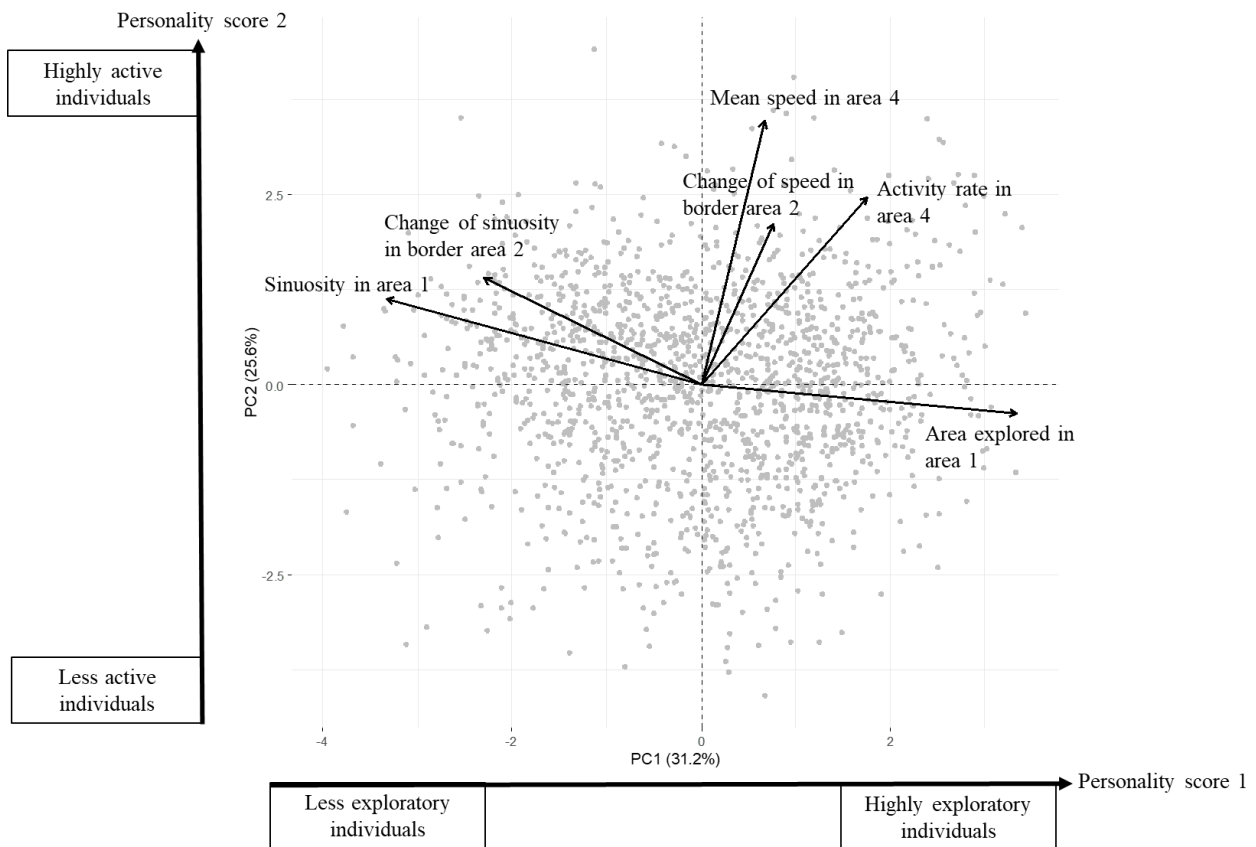
	(B) Change of speed in border area 2	(A) Mean speed in area 4	(A) Activity rate	(B) Change of sinuosity in border area 2	(E) Sinuosity in area 1
(A) Mean speed in area 4	0.31 [0.25; 0.37] *				
(A) Activity rate in area 4	0.10 [0.04; 0.16] *	0.38 * [0.32; 0.43] *			
(B) Change of sinuosity in border area 2	0.11 [0.05; 0.17] *	0.07 [0.01; 0.14] *	-0.12 [-0.18; -0.06] *		
(E) Sinuosity in area 1	-0.07 [-0.14; -0.01] *	0.13 [0.07; 0.19] *	-0.16 [-0.22; -0.10] *	0.38 [0.32; 0.44] *	
(E) Area explored in area 1	0.11 [0.04; 0.16] *	0.01 [-0.04; 0.08]	0.29 [0.23; 0.34] *	-0.28 [-0.34; -0.22] *	-0.56 [-0.61; -0.52] *

379

380 **Table 3.** Parameters from the first two principal components (PC1 and PC2) of the PCA for the behavioural
 381 variables measured. Component loadings represent the relationship between the principal components and the
 382 variables from which they are constructed. The personality trait category to which each variable belongs is
 383 indicated in brackets: activity (A), boldness (B) and exploration (E).

Parameter	PC1	PC2
Eigenvalue	1.87	1.54
Percentage of variance explained	31.23	25.58
Component loading		
<i>(A) Mean speed in area 4</i>	0.16	0.84
<i>(A) Activity rate in area 4</i>	0.43	0.60
<i>(B) Change of speed in area 2</i>	0.19	0.51
<i>(B) Change of sinuosity in area 2</i>	-0.56	0.34
<i>(E) Area explored in area 1</i>	0.81	-0.09
<i>(E) Sinuosity in area 1</i>	-0.81	0.27

384



385 **Figure 3.** Graphical representation of the first two axes of the PCA on individual values (grey points) for
 386 repeatable behavioural traits (in black type).

387 *Are the measured traits correlated with fitness-related traits?*

388 Active females (i.e. those with higher PC2 values) had significantly larger numbers of
 389 offspring and significantly longer tibias (Table 4). Higher rates of exploration (i.e. higher PC1
 390 values) were not significantly correlated with any of the fitness-related traits measured. None
 391 of the behavioural variables or personality scores was significantly correlated with longevity
 392 (Table 4).

393

394

395

396

397 **Table 4.** Phenotypic correlation between behavioural traits (behavioural variables and personality scores) and
 398 other life history traits (with Spearman's rank correlation coefficient Rho and 95% confidence intervals (between
 399 square brackets) calculated from 959 individual values). Correlation coefficients with confidence intervals
 400 excluding zero are shown in bold, and correlation coefficients that remained significantly different from zero after
 401 Benjamini and Hochberg correction are indicated with an asterisk. The personality trait category to which each
 402 variable belongs is indicated in brackets: activity (A), boldness (B) and exploration (E).

	Offspring number	Longevity	Tibia length
Behavioural variables			
<i>(A) Mean speed in area 4</i>	0.20 [0.14; 0.26] *	-0.05 [-0.12; 0.03]	0.19 [0.12; 0.25] *
<i>(A) Activity rate in area 4</i>	-0.01 [-0.08; 0.06]	-0.06 [-0.13; 0.02]	-0.07 [-0.13; 0.00]
<i>(B) Change of speed in border area 2</i>	0.13 [0.06; 0.19] *	-0.08 [-0.15; 0.00]	0.16 [0.10; 0.21] *
<i>(B) Change of sinuosity in border area 2</i>	0.11 [0.04; 0.17] *	0.002 [-0.07; 0.08]	0.05 [-0.01; 0.11]
<i>(E) Area explored in area 1</i>	-0.05 [-0.11; 0.01]	-0.05 [-0.13; 0.02]	-0.02 [-0.09; 0.04]
<i>(E) Sinuosity in area 1</i>	0.01 [-0.05; 0.07]	0.05 [-0.02; 0.13]	0.05 [-0.01; 0.11]
Personality scores			
<i>Exploration score 1 (PCI)</i>	-0.01 [-0.07; 0.06]	-0.05 [-0.13; 0.03]	-0.03 [-0.09; 0.03]
<i>Activity score 2 (PC2)</i>	0.17 [0.10; 0.23] *	-0.01 [-0.10; 0.07]	0.15 [0.09; 0.21] *

403

404 *Are the measured traits heritable?*

405 Broad-sense heritability estimates for behavioural traits and personality scores ranged
 406 from 0.01 to 0.11. Confidence intervals excluded zero for all traits linked to activity and
 407 exploration, whereas they included zero for the two traits linked to boldness (Table 5). Fitness-
 408 related traits (offspring number, tibia length and longevity) displayed broad-sense heritability
 409 ranging from 0.04 to 0.28, with all confidence intervals excluding zero (Table 5).

410

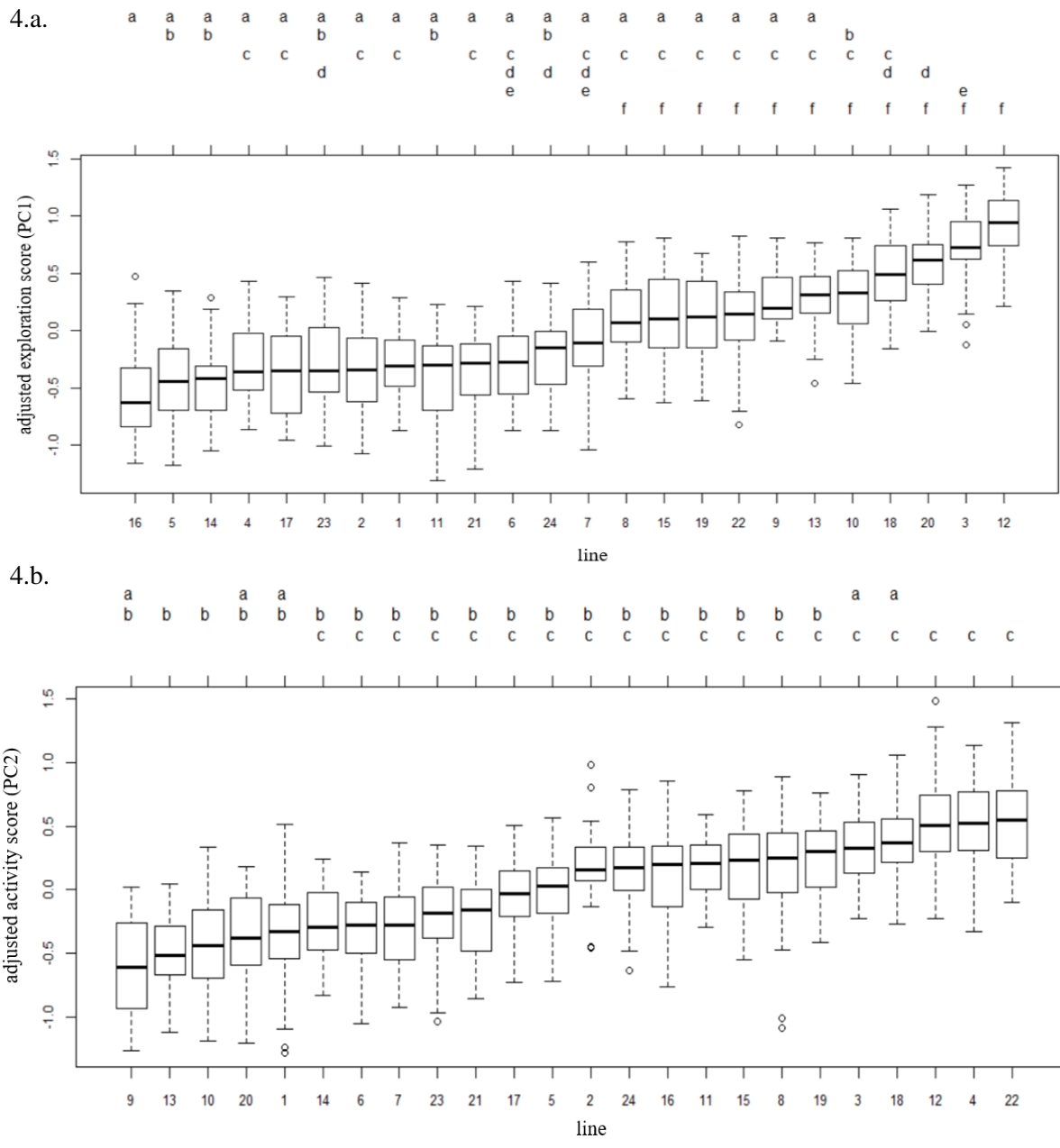
411 **Table 5.** Broad-sense heritability (H^2) of traits measured with 95% confidence intervals (between square brackets).
 412 Heritability estimates are shown in bold if their 95% confidence interval did not include zero. The personality trait
 413 category to which each behavioural variable belongs is indicated in brackets: activity (A), boldness (B) and
 414 exploration (E).

	H^2 [95% CI]
Behavioural variables	
(A) Mean speed in area 4	0.11 [0.05; 0.18]
(A) Activity rate in area 4	0.02 [0.00; 0.04]
(B) Change of speed in border area 2	0.01 [0.00; 0.03]
(B) Change of sinuosity in border area 2	0.01 [0.00; 0.03]
(E) Area explored in area 1	0.06 [0.02; 0.10]
(E) Sinuosity in area 1	0.06 [0.02; 0.11]
Personality scores	
Exploration score 1 (PC1)	0.08 [0.03; 0.13]
Activity score 2 (PC2)	0.05 [0.02; 0.10]
Fitness-related traits	
Offspring number	0.12 [0.05; 0.19]
Tibia length	0.05 [0.01; 0.09]
Longevity	0.28 [0.14; 0.39]

415

416 *Do personality traits differentiate between lines?*

417 We found significant differences in personality scores between lines (Figure 4.a and
 418 4.b), and the 24 lines were distributed along the first two axes of the PCA (Figure 5). We were
 419 therefore able to distinguish between lines that were very active and exploratory (e.g., lines 3
 420 and 12), and lines that were less active and exploratory (e.g., lines 14 and 21); we were also
 421 able to distinguish between lines that were very exploratory but not very active (e.g., lines 9
 422 and 10) and lines that were active but not very exploratory (for example line 4).



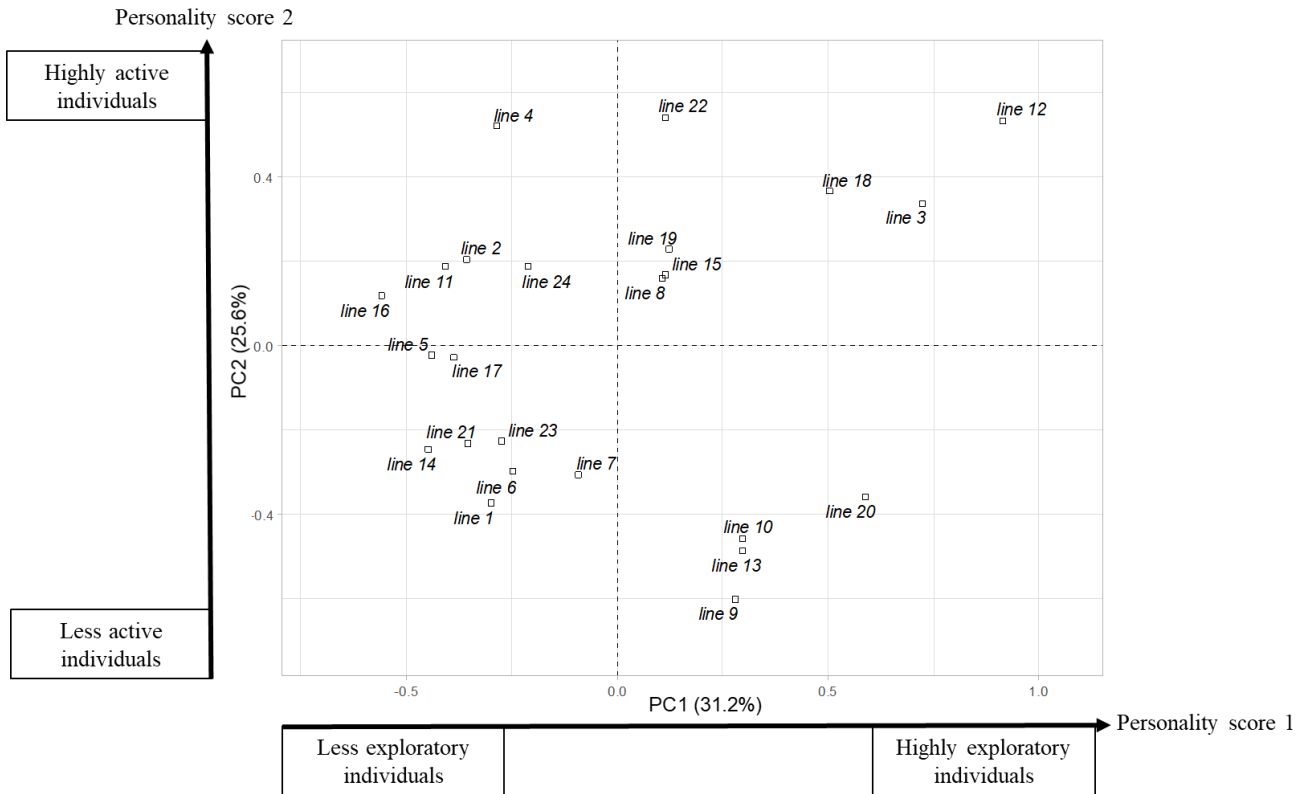
423

424 **Figure 4.** Boxplot of the adjusted values of personality score 1 (Fig. 4.a) and personality score 2 (Fig. 4.b) after
 425 the elimination of variation due to individual, subline and session effects, and compact letter display after Tukey
 426 all-pair comparisons. Two lines with no letters in common are considered to be significantly different (with a p-
 427 value <0.05).

428

429

430



431

432 **Figure 5.** Distribution of the mean points for the 24 lines (centroids) along the first two axes of the PCA.

433 *Are personality traits genetically correlated with fitness-related traits?*

434 The only genetic correlation between personality scores and fitness-related traits that
 435 remained significant after FDR correction was the positive correlation between exploration
 436 score (PC1) and offspring number (Table 6).

437

438

439

440

441

442

443 **Table 6.** Genetic correlation between personality and other life history traits (Spearman's rank correlation
 444 coefficient, based on the trait estimates of 24 near-isogenic lines, with associated p-values in brackets). Correlation
 445 coefficients with confidence intervals excluding zero are shown in bold, and correlation coefficients that remained
 446 significantly different from zero after Benjamini and Hochberg correction are indicated with an asterisk. The
 447 personality trait category to which each behavioural variable belongs is indicated in brackets: activity (A), boldness
 448 (B) and exploration (E).

	Offspring number	Longevity	Tibia length
Behavioural variables			
(A) Mean speed in area 4	-0.16 [-0.54; 0.33]	-0.32 [-0.62; 0.06]	0.51 [0.21; 0.70]
(A) Activity rate in area 4	-0.53 [-0.78; -0.14]	0.15 [-0.28; 0.54]	-0.21 [-0.62; 0.28]
(B) Change of speed in border area 2	-0.03 [-0.44; 0.37]	-0.45 [-0.72; -0.10]	0.35 [-0.05; 0.63]
(B) Change of sinuosity in border area 2	0.29 [-0.11; 0.63]	0.15 [-0.31; 0.56]	0.19 [-0.20; 0.54]
(E) Area explored in area 1	-0.60 [-0.79; -0.26] *	0.01 [-0.40; 0.43]	-0.28 [-0.65; 0.16]
(E) Sinuosity in area 1	0.63 [0.33; 0.82] *	0.25 [-0.20; 0.62]	-0.03 [-0.43; 0.39]
Personality scores			
Exploration score 1 (PC1)	-0.64 [-0.83; -0.29] *	-0.09 [-0.52; 0.32]	-0.19 [-0.58; 0.24]
Activity score 2 (PC2)	-0.10 [-0.52; 0.39]	-0.22 [-0.66; 0.19]	0.41 [0.01; 0.67]

449

450 **Discussion**

451 We investigated whether animal personality could be used to develop or improve
 452 phenotyping methods for the BCA *Trichogramma evanescens*. We first developed an
 453 automated phenotyping method based on automated pathway analysis, providing a set of
 454 behavioural trait measures that proved repeatable over time and heritable (i.e. personality traits).
 455 We then identified differences in life history strategies between individuals as behavioural traits
 456 were correlated together and combined them into personality scores, which were correlated with
 457 other life history traits. Finally, we observed differences in personality scores between the 24
 458 genotypes of *T. evanescens* and found a negative genetic correlation between exploration and
 459 fecundity.

460 *Evidence of personality traits in Trichogramma evanescens*

461 Personality has never before been assessed in a species as small as *Trichogramma*
462 *evanescens*. Based on other video-tracking studies in other species (Branson et al., 2009;
463 Charalabidis, Dechaume-Moncharmont, Petit, & Bohan, 2017), we designed and developed a
464 video-tracking approach measuring a large number of variables relating to the movements of *T.*
465 *evanescens* individuals during their presence in the different areas (white, black and border
466 areas) within an experimental arena. Here, we chose to work on seven variables that (i) could
467 be calculated with methods commonly used in trajectory and movement studies (speeds,
468 trajectory length and sinuosity estimates) (Mclean & Volponi, 2018) and (ii) we considered to
469 be associated with some of the commonly defined personality traits defined by Réale et al.
470 (2007): boldness, exploration and activity.

471 For each of the seven behavioural variables, we assessed repeatability, broad-sense
472 heritability and phenotypic and genetic correlations between personality traits and between
473 these traits and other life history traits, according to methods generally used in animal
474 personality studies (Réale et al., 2007). For six of the seven behavioural variables, we observed
475 significant repeatability (R) (values ranging from 0.10 to 0.35, Table 1). These six variables
476 could therefore be considered as personality traits. The R values obtained were within the range
477 of R values commonly observed for behavioural traits, although most were lower than the mean
478 R value obtained for animal behaviour (0.35) (Bell et al., 2009). However, personality has rarely
479 been studied in parasitoid insects, and a recent study on the parasitoid wasp *Venturia canescens*
480 reported a similar R value for activity and a lower R value for exploration (about 0.10, whereas
481 we obtained R values for exploration-related variables of 0.18 and 0.24(Gomes et al., 2019)).

482 The broad-sense heritability of the variables (ranging from 0.06 to 0.11, Table 5) was
483 lower than the mean value for animal behaviour (0.24) in the meta-analysis by Dochtermann et
484 al. (2019). Stirling et al. (2002) found no significant differences in heritability between
485 behavioural and life-history traits in their meta-analysis, whereas we found that heritability

486 values for personality traits were lower than heritability values of two classical fitness-related
487 traits (offspring number and longevity) in *T. evanescens* (Table 5).

488 Behavioural traits could be grouped together into two continuums or behavioural
489 syndromes (Denis Réale et al., 2007; Sih et al., 2004; Sih, Cote, Evans, Fogarty, & Pruitt, 2012):
490 a continuum extending from individuals with low levels of exploratory behaviour to highly
491 exploratory individuals, and a continuum extending from individuals with low levels of activity
492 to highly active individuals (Figure 3). Bold (or shy) behaviour and active behaviour have been
493 shown to be correlated with fecundity traits in several species (Biro & Stamps, 2008), but rarely
494 in insects (Monceau et al., 2017). In this study, we found a weak but significant phenotypic
495 correlation between behavioural traits, fecundity and body length, as shy or active females
496 produced more offspring, and had longer tibias (Table 4). The positive correlation between
497 activity (with the variable “mean speed”) and the length of tibia is quite intuitive, as it should
498 be easier for individuals with longer tibia to cover larger distance. Moreover, bigger females
499 would have more energy to spend for both offspring production and activity. However, although
500 these positive correlations might have been expected, they are equivocal in the literature and
501 seem to depend on the function of personality traits in a given species (Biro & Stamps, 2008;
502 Gu, Hughes, & Dorn, 2006). We can note that the variable for shyness on which we found a
503 phenotypic correlation with fecundity and tibia length is the “change of speed in border area
504 2”, which is also directly linked with speed abilities. Finally, an analysis of genetic correlations
505 showed that the lines with the most exploratory individuals had the smallest numbers of
506 offspring (Table 6). These correlations seem to be compatible with the pace-of-life syndrome
507 (POLS) hypothesis, a currently debated hypothesis (Royauté, Berdal, Garrison, &
508 Dochtermann, 2018), according to which, behavioural traits are related to morphological,
509 physiological and other life-history traits (Réale et al. 2010).

510 *Potential of personality traits for use in genetic improvement of ~~the~~ biocontrol agents*

511 *Trichogramma evanescens*

512 In this study, our aim was to evaluate the possibility of using personality traits as traits
513 of interest in biological control, and of integrating these traits into genetic improvement
514 programmes for the BCA *T. evanescens*. The six repeatable behavioural traits we measured
515 were correlated with each other, and could be combined into two continuums. For each
516 individual and continuum, we estimated a personality score corresponding to the position of the
517 individual along the continuum, a common method in animal personality studies (Mazué et al.,
518 2015; Monceau et al., 2017). We found that it was possible to capture a large proportion of the
519 behavioural trait variance with two scores (36.2% of the total variance explained by personality
520 score 1, and 26.4% explained by personality score 2). This finding highlights the utility of
521 calculating a few synthetic indices (or scores), rather than measuring large numbers of
522 variables, to obtain relevant information for BC. We therefore systematically present our results
523 considering all the traits individually and summarized as two personality scores.

524 The relevance of the behavioural traits or synthetic scores to the context of BC was
525 demonstrated by the phenotypic correlations between these traits and scores and the traits
526 classically measured in BC (fecundity, longevity and body length) (Hopper et al., 1993; Prezotti
527 et al., 2004; Roitberg et al., 2001; Smith, 1996). In this study, active females (i.e. with high
528 values for “mean speed in area 4” and “personality score 2”) produced more offspring and had
529 longer tibias (Table 4). By contrast, we found that bold females (i.e. with low values for “change
530 of speed in border area 2” and “change of sinuosity in border area 2”) produced a small number
531 of offspring (Table 4). In several species, activity and boldness behaviours have been shown to
532 be correlated with traits of ecological importance, such as dispersal (Sih et al., 2004), which is
533 also a trait linked to field efficiency in BC (Fournier & Boivin, 2000). Our results indicate that
534 active females produce more offspring, which is predictive of a high degree of efficiency in

535 rearing conditions and, in the case of parasitoids, in the field. Note, however, that we did not
536 assess survival or body condition in the offspring. The same females also displayed shy
537 behaviour. The impact of a shy behaviour on an individual's field efficiency would depend on
538 the agrosystem conditions. Indeed, in the presence of high densities of predators intraguild
539 predation may occur (Bennett, Gillespie, Shipp, & VanLaerhoven, 2009; Dumont et al., 2018).
540 In this scenario, shy parasitoid individuals (i.e. the intraguild preys) might be less predated as
541 they might be less willing to take risks, compared to bold individuals. However, in situations
542 where intraguild predation is not a challenge, bolder individuals, more willing to take risks,
543 could be faster in finding resources (i.e. egg patches in the case of *Trichogramma* species).
544 Therefore, Further studies are, ~~therefore,~~ required to assess the full ecological relevance ~~of~~
545 ~~these lines of the lines we studied-~~ in BC. The relevance of the variables measured will be
546 confirmed only if they are shown to be correlated with BC performance in industrial and field
547 and/or greenhouse conditions.

548 Most of our data analyses aimed to evaluate the added value of the measured behavioural
549 traits for genetic improvement strategies, breeding programs. We found that personality scores
550 differ among isogenic lines (Figure 4.a. and Figure 4.b.) and that these differences highlight
551 contrasted behaviours, as evidenced by their distribution along the two personality scores in
552 Figure 5. This may make it possible to differentiate between these behaviours and to select for
553 them, should they prove relevant in terms of BC efficiency. We also observed a negative genetic
554 correlation between the personality score relating to exploration and offspring production. It
555 will probably be important to take this trade-off into account in BC, as it may oppose
556 performance in rearing and performance in the field. Indeed, as for activity and boldness,
557 exploration behaviours are also correlated with traits linked to field efficiency in BC, such as
558 dispersal (Fournier & Boivin, 2000; Sih et al., 2004).

559 Given these results, and the ease with which all the traits can be assessed and personality
560 scores obtained through short (90 seconds) automated video-tracking measurements, the new
561 method described here may provide useful criteria for the selection of candidate BCA taxa
562 (populations, strains, sibling species, etc.) or for quality control purposes. However, the high
563 level of intra-isogenic line variability observed (Figure 4.a and Figure 4.b), accounting for the
564 relatively low broad-sense heritability of the traits and scores (between 0.01 and 0.11; Table 5),
565 constrains the use of this method, as it may be necessary to phenotype large numbers of
566 individuals for reliable comparisons between taxa or reared populations. The low heritability
567 also constitutes an obstacle to the implementation of ambitious experimental evolution
568 programmes. Oriented experimental evolution may be fastidious for traits displaying such a
569 high degree of environmentally induced variability. As a comparison, breeding programmes for
570 livestock animals generally make use of traits with higher heritability. Heritability values for
571 morphological, physiological, behavioural or other traits linked to fitness and considered in
572 these breeding programmes generally range from 0.17 to 0.70 in sheep, pigs, cows and fish
573 (Juengel et al., 2019; Kavlak & Uimari, 2019; Moretti, de Rezende, Biffani, & Bozzi, 2018;
574 Vargas Jurado, Leymaster, Kuehn, & Lewis, 2016). However, in order to select traits with low
575 heritability values, the method of genomic selection is already used for livestock animals
576 (e.g.(Hayes, Bowman, Chamberlain, & Goddard, 2009). This method is based on the
577 phenotyping and genotyping of a high number of individuals in order to establish a statistical
578 equation between the genotype and the phenotype. Based on this equation, it is then possible to
579 predict the phenotype of an individual, knowing only its genotype (Hayes et al., 2009). This
580 method has never been applied to BCA, but has been recently suggested as a promising
581 application to BCA selection (Leung et al., 2020), and could help considering behavioural –
582 and personality – traits in BCA selection programs.

583 **Conclusion**

584 In conclusion, the use of methods and concepts of animal personality to develop
585 phenotyping methods and associated data analyses for BC led to the rapid phenotyping of traits
586 rarely used in BC that were repeatable, heritable and correlated with fitness-related traits. Our
587 results also provide support to investigate the interest of animal personality in other BCA
588 species (parasitoids or predators). However, it will be possible to consider the actual potential
589 of these traits and of the phenotyping method satisfactory only after investigating the
590 relationships between the laboratory-measured traits and BC performance indices in real BC
591 situations, in industrial production settings or in field releases. This first study has driven the
592 launch of large-scale field experiments, which are currently underway and aim to generate field-
593 release performance indices.

594 **Acknowledgements**

595 We would like to thank all the members of staff at Institut Sophia Agrobiotech who
596 helped us when we were overwhelmed by the sheer numbers of *Trichogramma* individuals in
597 our experimental tubes, including Léa Tregoat-Bertrand in particular. Special thanks to Maxime
598 Dahirel for helpful advice for statistical analysis, and to Morgane Bequet-Rennes and Simon
599 Vaïsse, who contributed to this study by running previous experiments that led to our decision
600 to design this new phenotyping assay. We thank Marta Montserrat, Bart A. Pannebakker and
601 François Dumont for their thoughtful comments. We thank the biocontrol company BIOLINE,
602 in particular Paloma Martinez and Benoît Franck, and the INRAE biological resource centre
603 “Egg-Parasitoids Collection” (CRB EP-Coll, Sophia Antipolis) for providing field-sampled
604 populations for the establishment of some of the isogenic lines used in this study. We thank the
605 Microscopy Platform of the Sophia Agrobiotech Institute, INRA, UNS, CNRS, UMR 1355-
606 7254, INRA PACA, Sophia Antipolis for providing access to instruments and technical advice.
607 This work was funded by the the French “Agence Nationale de la Recherche” under the

608 “Investissements d'Avenir UCAJEDI” project with reference n ° ANR-15-IDEX-01. Silène
609 Lartigue received financial support from the French Ministry in charge of Agriculture.

610 **Conflict of interest disclosure**

611 The authors of this article declare that they have no financial conflict of interest with the content
612 of this article. Jérôme Moreau, François-Xavier Dechaume-Moncharmont and Vincent
613 Calcagno belong to the panel of *PCIEcology* recommenders.

614 **Data and R code**

615 Data tables and code needed to re-do the analyses and figures are available on Zenodo
616 (<http://doi.org/10.5281/zenodo.4058218>).

617 **References**

- 618 Abràmoff, M. D., Magalhães, P. J., & Ram, S. J. (2004). Image processing with imageJ.
619 *Biophotonics International*, 11(7), 36–41.
- 620 Amat, I., Desouhant, E., Gomes, E., Moreau, J., & Monceau, K. (2018). Insect personality:
621 what can we learn from metamorphosis? *Current Opinion in Insect Science*, 27, 46–51.
- 622 Ayvaz, A., Karasu, E., Karabörklü, S., Tunçbilek, A. Ş. (2008). Effects of cold storage, rearing
623 temperature, parasitoid age and irradiation on the performance of *Trichogramma*
624 *evanescens* Westwood (Hymenoptera: Trichogrammatidae). *Journal of Stored Products*
625 *Research*, 44, 232–240.
- 626 Bates, D., Maechler, M., Bolker, B., & Walker, S. (2015). Fitting linear mixed-effects models
627 using lme4. *Journal of Statistical Software*, 67(1), 1–48.
- 628 Bell, A. M., Hankison, S. J., & Laskowski, K. L. (2009). The repeatability of behaviour: a meta-
629 analysis. *Animal Behaviour*, 77(4), 771–783.
- 630 Benjamini, Y., & Hochberg, Y. (1995). Controlling the false discovery rate: a practical and
631 powerful approach to multiple testing. *Journal of the Royal Statistical Society: Series B*
632 *(Methodological)*, 57(1), 289–300.
- 633 Bennett, J. A., Gillespie, D. R., Shipp, J. L., & VanLaerhoven, S. L. (2009). Foraging strategies
634 and patch distributions: Intraguild interactions between *Dicyphus hesperus* and *Encarsia*
635 *formosa*. *Ecological Entomology*, 34(1), 58–65.
- 636 Biro, P. A., & Stamps, J. A. (2008). Are animal personality traits linked to life-history
637 productivity? *Trends in Ecology and Evolution*, 23(7), 361–368.
- 638 Bolnick, D. I., Amarasekare, P., Araújo, M. S., Bürger, R., Levine, J. M., Novak, M., ...
639 Vasseur, D. A. (2011). Why intraspecific trait variation matters in community ecology.
640 *Trends in Ecology and Evolution*, 26(4), 183–192.
- 641 Branson, K., Robie, A. A., Bender, J., Perona, P., & Dickinson, M. H. (2009). High-throughput
642 ethomics in large groups of *Drosophila*. *Nature Methods*, 6(6), 451–457.
- 643 Briese, D. T. (2000). Classical biological control. *Australian Weed Management Systems*, 161–
644 192.
- 645 Charalabidis, A., Dechaume-Moncharmont, F. X., Petit, S., & Bohan, D. A. (2017). Risk of
646 predation makes foragers less choosy about their food. *PLoS ONE*, 12(11), 1–18.
- 647 Cõnsoli, F. L., & Parra, J. R. P. (1995). Effects of constant and alternating temperatures on
648 *Trichogramma galloi* Zucchi (Hym., Trichogrammatidae) biology II.-parasitism capacity
649 and longevity. *Journal of Applied Entomology*, 119, 667–670.
- 650 Cote, J., Fogarty, S., Weinersmith, K., Brodin, T., & Sih, A. (2010). Personality traits and
651 dispersal tendency in the invasive mosquitofish (*Gambusia affinis*). *Proceedings of the*
652 *Royal Society B: Biological Sciences*, 277(1687), 1571–1579.
- 653 Dingemanse, N. J., Kazem, A. J. N., Reale, D., & Wright, J. (2009). Behavioural reaction
654 norms : animal personality meets individual plasticity. *Trends in Ecology and Evolution*,
655 25(2), 81–89.
- 656 Dochtermann, N. A., Schwab, T., Anderson Berdal, M., Dalos, J., & Royauté, R. (2019). The
657 heritability of behavior: a meta-analysis. *Journal of Heredity*, 110(4), 403–410.

- 658 Dochtermann, N. A., Schwab, T., & Sih, A. (2014). The contribution of additive genetic
659 variation to personality variation: heritability of personality. *Proceedings of the Royal*
660 *Society B: Biological Sciences*, 282(1798).
- 661 Dumont, F., Aubry, O., & Lucas, E. (2018). From evolutionary aspects of zoophytophagy to
662 biological control. *Frontiers in Ecology and Evolution*, 6, 1–15.
- 663 Dumont, F., Réale, D., & Lucas, E. (2017). Isogroup selection to optimize biocontrol increases
664 cannibalism in omnivorous (Zoophytophagous) bugs. *Insects*, 8(3).
- 665 Eilenberg, J., Hajek, A., & Lomer, C. (2001). Suggestions for unifying the terminology in
666 biological control. *BioControl*, 46(4), 387–400.
- 667 Fournier, F., & Boivin, G. (2000). Comparative dispersal of *Trichogramma evanescens* and
668 *Trichogramma pretiosum* (Hymenoptera: Trichogrammatidae) in relation to
669 environmental conditions. *Population Ecology*, 29(1), 55–63.
- 670 Godfray, H. C. J., Beddington, J. R., Crute, I. R., Haddad, L., Lawrence, D., Muir, J. F., ...
671 Toulmin, C. (2010). Food security: the challenge of feeding 9 billion people. *Science*, 327,
672 812–818.
- 673 Gomes, E., Desouhant, E., & Amat, I. (2019). Evidence for risk-taking behavioural types and
674 potential effects on resource acquisition in a parasitoid wasp. *Animal Behaviour*, 154, 17–
675 28.
- 676 Gosling, S. D. (2001). From mice to men: what can we learn about personality from animal
677 research? *Psychological Bulletin*, 127(1), 45–86.
- 678 Gu, H., Hughes, J., & Dorn, S. (2006). Trade-off between mobility and fitness in *Cydia*
679 *pomonella* L. (Lepidoptera: Tortricidae). *Ecological Entomology*, 31(1), 68–74.
- 680 Hartig, F. (2019). DHARMA: Residual Diagnostics for Hierarchical (Multi-Level / Mixed)
681 Regression Models. Retrieved from <https://cran.r-project.org/package=DHARMA>
- 682 Hassan, S. A. (1993). The mass rearing and utilization of *Trichogramma* to control
683 lepidopterous pests, achievements and outlook. *Pesticide Science*, 37, 387–391.
- 684 Hayes, B. J., Bowman, P. J., Chamberlain, A. J., & Goddard, M. E. (2009). Invited review:
685 Genomic selection in dairy cattle: Progress and challenges. *Journal of Dairy Science*,
686 92(2), 433–443.
- 687 Hervé, M. (2020). RVAideMemoire: testing and plotting procedures for biostatistics. Retrieved
688 from <https://cran.r-project.org/package=RVAideMemoire>
- 689 Hopper, K. R., Roush, R. T., & Powell, W. (1993). Management of genetics of biological-
690 control introductions. *Annual Review of Entomology*, 38, 27–51.
- 691 Hothorn, T., Bretz, F., & Westfall, P. (2008). Simultaneous inference in general parametric
692 models. *Biometrical Journal*, 50(3), 346–363.
- 693 Juengel, J. L., Hickey, S. M., Clarke, S. M., Cullen, N. G., McEwan, J. C., & Dodds, K. G.
694 (2019). Heritability of ram mating success in multi-sire breeding situations. *Animal*, 13(5),
695 917–923.
- 696 Kassambara, A., & Mundt, F. (2019). factoextra: extract and visualize the results of multivariate
697 data analyses. Retrieved from <https://cran.r-project.org/package=factoextra>

- 698 Kavlak, A. T., & Uimari, P. (2019). Estimation of heritability of feeding behaviour traits and
699 their correlation with production traits in Finnish Yorkshire pigs. *Journal of Animal*
700 *Breeding and Genetics*, 136(6), 484–494.
- 701 Kralj-fiser, S., & Schuett, W. (2014). Studying personality variation in invertebrates : why
702 bother ? *Animal Behaviour*, 91, 41–52.
- 703 Kruitwagen, A., Beukeboom, L. W., & Wertheim, B. (2018). Optimization of native biocontrol
704 agents, with parasitoids of the invasive pest *Drosophila suzukii* as an example.
705 *Evolutionary Applications*, 11(9), 1473–1497.
- 706 Le, S., Josse, J., & Husson, F. (2008). FactoMineR: an R package for multivariate analysis.
707 *Journal of Statistical Software*, 25(1), 1–18.
- 708 Leung, K., Ras, E., Ferguson, K. B., Ariëns, S., Babendreier, D., Bijma, P., ... Pannebakker, B.
709 A. (2020). Next-generation biological control: the need for integrating genetics and
710 genomics. *Biological Reviews*, 95(6), 1838–1854.
- 711 Lirakis, M., & Magalhães, S. (2019). Does experimental evolution produce better biological
712 control agents? A critical review of the evidence. *Entomologia Experimentalis et*
713 *Applicata*, 167(7), 584–597.
- 714 Lommen, S. T. E., Jong, P. W. De, & Pannebakker, B. A. (2017). It is time to bridge the gap
715 between exploring and exploiting : prospects for utilizing intraspecific genetic variation to
716 optimize arthropods for augmentative pest control – a review. *Entomologia Experimentalis*
717 *et Applicata*, 168, 108–123.
- 718 Lynch, M., & Walsh, B. (1998). *Genetics and Analysis of Quantitative Traits*. Sunderland, MA:
719 Sinauer.
- 720 Mackauer, M. (1976). Genetic problems in the production of biological control agents. *Annual*
721 *Review of Entomology*, 21(1), 369–385.
- 722 Mazué, G. P. F., Dechaume-Moncharmont, F. X., & Godin, J. G. J. (2015). Boldness-
723 exploration behavioral syndrome: Interfamily variability and repeatability of personality
724 traits in the young of the convict cichlid (*Amatitlania siquia*). *Behavioral Ecology*, 26(3),
725 900–908.
- 726 Mclean, D. J., & Volponi, M. A. S. (2018). trajr : An R package for characterisation of animal
727 trajectories, 440–448
- 728 Michalko, R., Pekár, S., & Entling, M. H. (2019). An updated perspective on spiders as
729 generalist predators in biological control. *Oecologia*, 189(1), 21–36.
- 730 Mills, Nicholas J., & Kean, J. M. (2010). Behavioral studies, molecular approaches, and
731 modeling: methodological contributions to biological control success. *Biological Control*,
732 52(3), 255–262.
- 733 Mills, Nick J, & Wajnberg, E. (2008). Optimal foraging behaviour and efficient biological
734 control methods. *Behavioural Ecology of Insect Parasitoids–From Theoretical*
735 *Approaches to Field Applications*, 3–30.
- 736 Monceau, K., Dechaume-Moncharmont, F. X., Moreau, J., Lucas, C., Capoduro, R., Motreuil,
737 S., & Moret, Y. (2017). Personality, immune response and reproductive success: an
738 appraisal of the pace-of-life syndrome hypothesis. *Journal of Animal Ecology*, 86, 932–
739 942.

- 740 Moretti, R., de Rezende, M. P. G., Biffani, S., & Bozzi, R. (2018). Heritability and genetic
741 correlations between rumination time and production traits in Holstein dairy cows during
742 different lactation phases. *Journal of Animal Breeding and Genetics*, 135(4), 293–299.
- 743 Nachappa, P., Margolies, D. C., Nechols, J. R., & Campbell, J. F. (2011). Variation in predator
744 foraging behaviour changes predator-prey spatio-temporal dynamics. *Functional Ecology*,
745 25(6), 1309–1317.
- 746 Nachappa, P., Margolies, D. C., Nechols, J. R., & Morgan, T. J. (2010). Response of a complex
747 foraging phenotype to artificial selection on its component traits. *Evolutionary Ecology*,
748 24(4), 631–655.
- 749 Nakagawa, S., & Cuthill, I. C. (2007). Effect size, confidence interval and statistical
750 significance: A practical guide for biologists. *Biological Reviews*, 82(4), 591–605.
- 751 Nakagawa, S., & Schielzeth, H. (2010). Repeatability for Gaussian and non-Gaussian data: a
752 practical guide for biologists. *Biological Reviews*, 85, 935–956.
- 753 Prezotti, L., Parra, J. R. P., Vencovsky, R., Coelho, A. S. G., & Cruz, I. (2004). Effect of the
754 size of the founder population on the quality of sexual populations of *Trichogramma*
755 pretiosum, in laboratory. *Biological Control*, 30(2), 174–180.
- 756 R Core Team. (2019). R: A language and environment for statistical computing. Vienna: R
757 Foundation for Statistical Computing. Retrieved from <https://www.r-project.org/>
- 758 Réale, D., Garant, D., Humphries, M. M., Bergeron, P., Careau, V., & Montiglio, P. O. (2010).
759 Personality and the emergence of the pace-of-life syndrome concept at the population
760 level. *Philosophical Transactions of the Royal Society of London. Series B, Biological*
761 *Sciences*, 365(1560), 4051–4063.
- 762 Réale, Denis, Reader, S. M., Sol, D., McDougall, P. T., & Dingemanse, N. J. (2007). Integrating
763 animal temperament within ecology and evolution. *Biological Reviews*, 82(2), 291–318.
- 764 Robin, D. C., & Marchand, P. A. (2020). Macroorganismes de biocontrôle en France, état des
765 lieux. *Innovations Agronomiques*, 79, 425–439.
- 766 Roitberg, B. D. (2007). Why pest management needs behavioral ecology and vice versa.
767 *Entomological Research*, 37, 14–18.
- 768 Roitberg, B. D., Boivin, G., & Vet, L. E. M. (2001). Fitness, parasitoids, and biological control:
769 an opinion. *The Canadian Entomologist*, 133, 429–438.
- 770 Royauté, R., Berdal, M. A., Garrison, C. R., & Dochtermann, N. A. (2018). Painless life? A
771 meta-analysis of the pace-of-life syndrome hypothesis. *Behavioral Ecology and*
772 *Sociobiology*, 72–64.
- 773 Royauté, R., Buddle, C. M., & Vincent, C. (2014). Interpopulation Variations in Behavioral
774 Syndromes of a Jumping Spider from Insecticide-Treated and Insecticide-Free Orchards.
775 *Ethology*, 120(2), 127–139.
- 776 Schöller, M., & Hassan, S. A. (2001). Comparative biology and life tables of *Trichogramma*
777 *evanescens* and *T. cacoeciae* with *Ephesia elutella* as host at four constant temperatures.
778 *Entomologia Experimentalis et Applicata*, 98(1), 35–40.
- 779 Sforza, R. (2010). Recherche et évaluation des candidats pour la lutte biologique classique. In
780 *La lutte biologique. Vers de nouveaux équilibres écologiques* (Quae-Educa, pp. 280–285).

- 781 Sih, A., Bell, A., & Johnson, J. C. (2004). Behavioral syndromes: an ecological and
782 evolutionary overview. *Trends in Ecology and Evolution*, *19*(7), 372–378.
- 783 Sih, A., Cote, J., Evans, M., Fogarty, S., & Pruitt, J. (2012). Ecological implications of
784 behavioural syndromes. *Ecology Letters*, *15*, 278–289.
- 785 Smith, S. M. (1996). Biological control with *Trichogramma*: advances, successes, and potential
786 of their use. *Annual Review of Entomology*, *41*, 375–406.
- 787 Stirling, D. G., Reale, D., & Roff, D. A. (2002). Selection, structure and the heritability of
788 behaviour. *Journal of Evolutionary Biology*, *15*, 277–289.
- 789 Stoffel, M. A., Nakagawa, S., & Schielzeth, H. (2017). rptR: repeatability estimation and
790 variance decomposition by generalized linear mixed-effects models. *Methods in Ecology
791 and Evolution*, *8*(11), 1639–1644.
- 792 Suverkropp, B. P., Bigler, F., & Van Lenteren, J. C. (2001). Temperature influences walking
793 speed and walking activity of *Trichogramma brassicae* (Hym., Trichogrammatidae).
794 *Journal of Applied Entomology*, *125*(6), 303–307.
- 795 Tabone, E., Bardon, C., Desneux, N., & Wajnberg, E. (2010). Parasitism of different
796 *Trichogramma* species and strains on *Plutella xylostella* L. on greenhouse cauliflower.
797 *Journal of Pest Science*, *83*(3), 251–256.
- 798 Thibierge. (2015). *Innovations Agronomiques* 46 (2015), 27-37, 46, 27–37.
- 799 van Lenteren, J. C. (2012). The state of commercial augmentative biological control: Plenty of
800 natural enemies, but a frustrating lack of uptake. *BioControl*, *57*(1), 1–20.
- 801 van Ooers, K., & Sinn, D. L. (2011). Toward a basis for the phenotypic gambit: Advances in the
802 evolutionary genetics of animal personalities. In *From genes to animal behavior* (pp.
803 165–183). Springer, Tokyo.
- 804 Vargas Jurado, N., Leymaster, K. A., Kuehn, L. A., & Lewis, R. M. (2016). Estimating
805 heritability of wool shedding in a cross-bred ewe population. *Journal of Animal Breeding
806 and Genetics*, *133*(5), 396–403.
- 807 Wajnberg, E. (2004). Measuring genetic variation in natural enemies used for biological
808 control: why and how? In *Genetics, Evolution and Biological Control*.
- 809 Wajnberg, E. (2009). Genetics of the behavioral ecology of egg parasitoids. In *Egg Parasitoids
810 in Agroecosystems with Emphasis on Trichogramma* (pp. 149–165). Springer, Dordrecht.
- 811 Wajnberg, E., Bernstein, C., & Alphen, J. J. M. Van. (2008). *Behavioral ecology of insect
812 parasitoids: from theoretical approaches to field applications*. Blackwell Publishing.
- 813 Wajnberg, E., Curty, C., & Jervis, M. (2012). Intra-population genetic variation in the temporal
814 pattern of egg maturation in a parasitoid wasp. *PLoS ONE*, *7*(9).
- 815 Wajnberg, E., Roitberg, B. D., & Boivin, G. (2016). Using optimality models to improve the ef
816 fi cacy of parasitoids in biological control programmes. *Entomologia Experimentalis et
817 Applicata*, *158*, 2–16.
- 818 Willer, H., & Lernoud, J. (2019). *The world of organic agriculture: Statistics and emerging
819 trends 2019*. Bonn.
- 820 Wolf, M., & Weissing, F. J. (2012). Animal personalities : consequences for ecology and

821 evolution. *Trends in Ecology & Evolution*, 27(8), 452–461.