- 1 Quantifying between-individual variation using high-throughput phenotyping of
- 2 behavioural traits in the fruit fly (Drosophila melanogaster)
- 3 Erin L. Macartney¹, Patrice Pottier¹, Samantha Burke¹, Shinichi Nakagawa^{*1}, Szymon M.
- 4 Drobniak*1,2
- 5 * shared senior authorship
- 6
- ⁷ ¹ Evolution and Ecology Research Centre, School of Biological Earth and Environmental
- 8 Science, University of New South Wales, Sydney, Australia
- 9 ² Institute of Environmental Sciences, Jagiellonian University, Krakow, Poland

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- 11 Corresponding authors: s.drobniak@unsw.edu.au & erin.macartney@unsw.edu.au
- 12 Erin L. Macartney (ORCiD: 0000-0003-3866-143X, Twitter: @Erin_Macartney_)
- 13 Patrice Pottier (ORCiD: 0000-0003-2106-6597, Twitter: @Patrice_Science)
- 14 Samantha Burke (ORCiD: 0000-0001-6902-974X)
- 15 Shinichi Nakagawa (ORCiD: 0000-0002-7765-5182, Twitter: @itchyshin)
- 16 Szymon M. Drobniak (ORCiD: 0000-0001-8101-6247, Twitter: @szymekdr)

17

19 Abstract

20 Behavioural phenotyping is often time and labour-intensive, which can come at a cost to sample size and statistical precision. This is particularly a concern given that behaviours are 21 often highly variable within and between individuals, so naturally requires a larger sample 22 23 size. Drosophila melanogaster is a common model system in many research fields, and behavioural observations are frequently required. While D. melanogaster has a rapid 24 lifecycle that enables large numbers of flies to be reared for experiments, they are still 25 subject to methodological bottlenecks for behavioural observations. Additionally, their small 26 27 and delicate bodies make it difficult to observe certain behaviours in real-time, for example, in movement tracking or when performing repeated assays on the same individuals. Here, 28 29 we present a method, pilot data, custom data processing and analysis scripts for highthroughput behavioural phenotyping in *D. melanogaster*, as well as general remarks for 30 31 future studies. We used automatic tracking units to measure three behaviours in the same 32 individuals: locomotor activity, exploratory behaviour in a Y-maze, and habituation to a 33 startle response stimulus. We then examined between-individual variation and trait 34 correlations using our pilot data. Through this, we show that these behaviours are amenable to high-throughput automated tracking, with locomotor activity generating the most 35 36 straightforward and high-quality data. These methods can be used to free up time and 37 labour to allocate to increasing sample sizes and can be used to address a range of biological questions in ecology, evolution, and beyond. 38

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40 Key words: locomotion, startle, habituation, variation, power, precision, automated

41 phenotyping, repeatability

43 Introduction

44 Behaviour is closely tied to how an individual receives and processes information, and thus, 45 how it responds to its environment¹. It is also intricately linked to fitness and can represent individual condition and age². Measuring behavioural traits among individuals provides key 46 47 information regarding population averages and variation while measuring behaviour within individuals allows researchers to quantify behavioural repeatability³, often referred to as 48 'personality' ^{4,5}. Furthermore, measuring multiple behaviours within individuals allows the 49 quantification of between-individual behavioural correlations ⁶, or behavioural syndromes ⁷. 50 Therefore, accurately measuring behaviour is a key part of individual phenotyping with 51 broad implications across many fields of biology. 52 In contrast to genotyping and common morphological phenotyping, quantifying behaviour 53 can be highly laborious and time-consuming. Measurements of individual behaviour 54 55 regularly involve directly observing (either in real-time or through video recordings) and quantifying behaviours across time ^{8–11}. However, this often results in methodological 56 bottlenecks (e.g., an inability to measure enough individuals in the most 57 appropriate/physiologically relevant time-window) and can come at a cost to sample size ¹². 58 59 Moreover, experimental designs aimed at increasing sample size often inadvertently involve the addition of unwanted variation by using small-sample size experimental blocks. While 60 61 the use of randomized blocks can be a highly effective experimental design that can enable researchers to reach appropriate sample sizes ¹³, certain conditions (e.g., temperature, 62 lighting, time of day, unintentional use of different genetic backgrounds or clutches) are not 63 always able to be completely homogenous between blocks, even in laboratory 64 environments. Such compromises that occur within behavioural studies can result in 65 66 decreased precision of estimates, especially when not accounted for correctly during

statistical analysis ¹⁴. This is particularly a concern given that behavioural traits are often
highly variable within and between individuals and thus, naturally require larger sample
sizes ^{15,16}.

70 The fruit fly, Drosophila melanogaster is a common model system that is used to address a 71 range of research questions, including those related to behaviour. For example, fruit flies are common in studies of mating and courtship behaviour ^{8,17,18}, aggression ^{10,19}, and 72 learning and memory ^{20–22} using a variety of behavioural assays and responses. The fast life-73 74 history, short generation time, and ease of breeding hundreds of individuals at once are key 75 benefits of using *D. melanogaster* in behavioural studies. However, quantifying behaviour is still subject to procedural bottlenecks due to the time and labour required to observe 76 individual flies directly. An added difficulty lies in the small size and delicate body structure 77 78 of fruit flies, making repeated handling of the same individuals, as well as large numbers of 79 flies in a short time, challenging. One solution to such bottlenecks is automated 80 phenotyping. While not all behaviours are easily amenable to automation (e.g., mating 81 behaviours where relevant movements are often subtle and subject to nuances that may be hard to detect through automation, but see, Dankert et al, 2009²³), some behaviours are likely 82 83 to be easier to quantify through automation, thus easing the bottleneck and freeing up time and labour that can be invested in increasing sample size and statistical power. 84 85 Here, we present a method (including protocol, pilot test data, and analysis scripts) for

quantifying between-individual variation in behaviour in *D. melanogaster* using automated
 high-throughput phenotyping. We employed commercially available phenotyping units (see
 methods) that can perform high-throughput automated behavioural phenotyping in a range
 of model species, including *D. melanogaster*. Accompanying our detailed methods, we

provide test data on three different behaviours measured at an individual level: overall 90 91 locomotor activity, exploratory behaviour in a Y-maze, and habituation to a startle response 92 stimulus. Through automation, we were able to assay all three behaviours in 360 individuals 93 across four days – a sample size that would not have been easily attainable in this period 94 without a large team of full-time researchers. Furthermore, these behaviours occur very 95 rapidly and thus require specialised recording tools to provide sufficient sensitivity and 96 robustness. Therefore, we show that these methods can be used to assay multiple 97 behaviours in the same individuals, enabling the quantification of between-individual 98 variation in multiple behaviours, as well as examining the relationship between these 99 behaviours (e.g., to determine behavioural syndromes).

100 Methods

101 Fly husbandry

Our study used a large, outbred population of Canton-S wild-type fruit flies with overlapping generations maintained at the University of New South Wales, Sydney. The flies were under the 12h:12h photoperiod in a temperature-controlled room maintaining constant thermal conditions (25°C) and humidity of approximately 45-65%. The colony was kept in a 10-litre plastic terrarium with 12 bottles of commercially available food medium (Carolina Biological Supply, Burlington, NC). The three oldest bottles of food were replaced with three bottles of fresh food once a week.

To obtain test individuals, we removed the three newest bottles of food after 24 hours from the stock population and kept each bottle individually until the flies eclosed. This was done to approximately standardise larval density and the age of the individuals to be used in the test. The flies were then aged for six days post-eclosion before being used in the assays. Before each batch of assays, a sample of several dozen flies were transferred into an empty, dry culture bottle and briefly (3-5 minutes) submerged in a bucket of ice to induce a chilling coma. Following coma induction, the flies were transferred into their assay arenas (see Behavioural assays) using aspirators. Before each assay, the flies were allowed to return to full activity for approximately 15 minutes by leaving a loaded set of arenas in a lit 25°C area. Behavioural assays were performed in the mornings and afternoons to avoid the mid-day low-activity period.

120 Each batch of tested flies consisted of individuals repeatedly tested in three different 121 behavioural assays (see Behavioural assays; Fig. 1). Transferring the flies between assays 122 using different arena designs was achieved by briefly inducing chilling comas by placing the 123 arenas into a -20°C freezer until anesthetised. They were then aspirated into the new arena (i.e., between the 48 well plates or y-mazes; see Behavioural assays). Following all tests, the 124 flies were euthanised by leaving them overnight in a freezer (-20°C). The following day all 125 126 individual insects were sexed. In total, we assayed 360 flies arranged into 8 batches (i.e., four days of morning and afternoon batches) of 45 flies each (90 flies per day) (see General 127 remarks on how sample size can be increased further). 128

129

130 Behavioural assays

We employed three types of assays, each performed using a multi-well plexiglass or
polypropylene plate able to hold between 15 and 48 flies (depending on the assay type; Fig.
13. 1). Both male and female flies were randomly allocated within the plates and identities were
tracked across all assays so that we could later link individual behaviours measured in the

same individual. Tracking was achieved by maintaining a strict correspondence betweenarenas in subsequent assays.

137 Locomotion tracking: In this assay, individual flies were loaded into 1 cm deep, round, 138 transparent arenas arranged into a 48-well plate. Because one of the assays (the Y-maze) provided space to test 45 flies (3 plates of 15 flies each), the 48-well plates were always 139 140 filled with 45 flies whilst leaving the last three arenas empty. For the purposes of this study, 141 we have used 48-well cell-culture plates (NEST Biotechnology Co., Ltd.; China); the choice of 142 this brand was dictated by the negligible gap between the well rim and lid, which prevents 143 flies from escaping their allocated well and migrating to adjacent wells. Locomotion of 144 individual flies was followed by recording their position (see the Phenotyping units section for technical details) in short intervals, which was then used to calculate the distance 145 travelled by each fly in a set time interval. In our assay, the flies were tracked for 5 146 consecutive intervals, 10 minutes each (a total of 50-minutes observation). 147

Light-off startle assay: In this assay, we have used the established startle response of fruit 148 flies in response to a brief light-off stimulus ^{24,25}. The response is elicited by a short 'light-off' 149 150 pulse lasting 15 milliseconds (ms) and can be measured as a sudden and very fast movement (often involving flight) of the flies. The stimulus was delivered by an optogenetic 151 light-conducting plate mounted beneath the 48-well plate (the same plates that were used 152 for the locomotion assays). The device (part of the ZANTIKS MWP phenotyping system, see 153 154 the 'Phenotyping units' section for more details) consists of a plexiglass plate coupled with a 155 set of several bright LEDs. The diodes emit green light (530 nm), flooding the experimental arenas with over 7000 lux of light. The system is connected to the phenotyping unit and 156 delivers stimuli in the form of brief light-off pulses. 157

The objective of our experiment was to study habituation to the light-off stimulus. Thus, in
our assay, we subjected the flies to three consecutive 15-ms light-off pulses, 1 second apart.
The startle response (distance covered by each fly) was recorded within a 1-second interval
following each light-off pulse.

162 Y-maze tracking: The third assay was performed to explore D. melanogaster behaviour in a simple 3-arm maze ^{26–29}. The test was performed in a different type of arena: flies were 163 loaded into small plexiglass blocks with a y-shaped forking channel etched inside of them. 164 The maze was covered by a sliding coverslip that allowed for easy loading and unloading of 165 166 mazes with flies. Each plate contained 15 y-mazes. Tracking of flies' behaviour in the mazes 167 lasted 30 minutes. Apart from recording the time spent by flies in each of the three maze arms we have also recorded each arm crossing event (i.e., a fly crossing from one maze arm 168 to another, through the central "neutral zone"). 169

170 The protocol we used was based on a comparative study looking at Y-maze exploration behaviour in mice, zebrafish, and fruit flies ²⁹. Briefly, the assay generates a sequence of 171 zone locations visited by flies. Transition between two zones is classified as either left (L) or 172 173 right (R) turn. Analysis involves creating a sliding window along the sequence of turns executed by each fly and grouping the turns in triplets ('trigrams'). The RRR and LLL trigrams 174 are then classified as full repetitions (i.e., 3 consecutive turns in one direction), and RLR/LRL 175 176 trigrams are classified as full alternations (sequences of 3 turns, each in the opposite 177 direction to the previous). Proportions of full alternations and full repetitions are calculated 178 as basic indices of Y-maze exploration by dividing respective counts by the total number of 179 recorded turn triplets. Similar analyses involving automated tracking of fruit flies in a nonbaited Y-maze have already been performed (e.g., in the context of neuronal control of
 handedness²⁷), but our method offers considerably higher time efficiency and throughput.

182 *Phenotyping units*

All behavioural tests were performed using automated tracking units produced by Zantiks (Cambridge, UK). The units we employed were from the WMP series, suitable for tracking of small-sized animals such as small insects, crustaceans, fish larvae, etc. Each unit consists of a computer that controls its operation, an experimental chamber that can host experimental arenas (and can be connected to a temperature control unit, able to maintain internal chamber temperature in a narrow, set range) and a camera system able to track animal movement in arenas inserted to the chamber.

190 All the arenas used in our assays had the same format (i.e., dimensions of a standard ELISA 191 multi-well plate) and were placed inside each unit on a raised stand (locomotion and Y-maze 192 assays) or the designated optogenetic stand (light-off startle assay). To be able to track the 193 animals, the unit requires a correct definition of experimental arenas (a bitmap "asset" file mapping regions of the recorded image to specific experimental arenas) and an auto-194 195 reference process that removes images of actual animals from the immobile background 196 (thus allowing them to be traced using the actual experiment). The auto-reference stage 197 was programmed into each assay and lasted 10 minutes (locomotion and Y-maze assays) or 5 minutes (light-off startle assay) to make sure that within this interval each individual had 198 moved in its arena. 199

Experimental procedures in the units are controlled by a scripting language (Zanscript) –
 scripts describing the three assays used in our study can be found in the GitHub repository
 (https://github.com/elmacartney/Dmel_methods).

We have used 4 phenotyping units in total. Three of them were used to run Y-maze assays 203 (40 minutes), whereas the fourth one was used to run the locomotion (30 minutes) and 204 205 startle (10 minutes) assays concurrently to the Y-maze assays on another set of flies. Such 206 arrangement of units means that different batches of flies were subjected to subsequent 207 tests in different orders (Y-maze was the first performed assay in some flies while locomotion and startle assays were first in others). Any potential variation introduced by 208 209 differing orderings of the assays is captured by appropriate grouping variables that can be 210 introduced as random effects in statistical analyses.

211

212 Data analysis

The light-off startle and locomotion data were used without additional data collation, i.e., 213 214 the respective responses (see above for details) were used in downstream procedures. Since 215 the files produced by Zantiks units have a particular format (a header section with technical 216 details, followed by the actual data that is formatted according to the script run on the machine, then a footer), data processing involved parsing each file to extract the most 217 218 relevant information (e.g., experimental unit ID, assay ID, run date, formatted data matched 219 with the numbers of experimental arenas). Parsing steps used in each assay type are 220 presented in the GitHub repository (https://github.com/elmacartney/Dmel_methods; please see the R directory for a detailed RMarkdown document). Pre-processing of the 221 startle assay outputs were based on an earlier study applying the same test ²⁵. The details of 222 223 file parsing may differ if the Zanscript responsible for performing the experiment is modified 224 or additional data is being generated: in such cases, it is recommended to test the parsing section of the code to ensure correct columns are extracted from the raw data files. 225

The Y-maze behavioural test outputs required more sophisticated processing. Our protocol 226 is based on a modified analysis from Cleal et al.²⁹. In brief, the analysis extracted all maze 227 228 arm switches observed during the assay (i.e., transitions between two maze arms, termed 229 'zone changes'). The switches were then classified as left (L)- or right (R)-turns, and 230 sequences of consecutive L/R-turns extracted from a sliding window moving along the recorded sequence of Y-maze locations were assembled into trigrams. We then calculated 231 232 the proportion of alternating (LRL or RLR) vs. sequential (LLL, RRR) vs. partial (RRL, LLR) 233 movements in the total count of all possible maze explorations (which also include returning 234 to the same zone).

235 Resulting response variables were analysed using linear mixed-effects models with a 236 gaussian response (continuous variables, e.g., the distance travelled by an individual within an arena) or generalised mixed-effects models with a Poisson error distribution (count 237 variables, e.g., the number of alteration sequences in a Y-maze). Mixed models were fitted 238 using the *lme* function in *nlme* package³⁰ in R (R Core Team, 2021). Each model contained 239 240 fixed effects of sex and date. Experimental batch ID was included as a random effect. For the models analysing locomotion activity, fly movement was quantified in five repeated 241 242 measurement bins of 10 minutes. In these models we also included the random effect of individual fly ID to accommodate for this source of dependence. Continuous variables were 243 log-transformed wherever needed (based on the visual inspection of model residuals) and 244 245 zero-centred. To test for potential effects of fly fatigue we have also generated an additional 246 variable indicating whether a given assay was performed as the first or second assay on a given batch of flies (in all cases, the flies were either tested in the Y-maze, and then for 247 248 locomotion and startle response, or in the reversed order; locomotion always preceded the startle assay). Finally, we also ran a multivariate model including the locomotor activity, 249

- alternation and repetition indices, and handedness (i.e., asymmetry in choosing left versus
- 251 right runs in the Y-maze) to estimate cross-trait correlations and demonstrate that
- 252 usefulness of our protocol in identifying potential behavioural syndromes. The model was fit
- 253 in MCMCglmm using an uninformative inverse-Wishart prior for variance and flat normal
- 254 priors for means.

255 Results

256 Locomotion tracking

257 Individual flies exhibited ample variation in their locomotor activity, as measured by the

total distance travelled by each individual within the 10-minute bins (Fig. 2A, Fig. S1).

259 Locomotion behaviour was repeatable across five consecutive assay bins: a linear mixed 260 model looking at locomotor activity and partitioning variation in this trait into the betweenand within-individual variance indicated the existence of substantial repeatability (ICC = 261 0.36, LRT χ^2 = 412.6, df = 1, p < 0.001). Experimental block explained 17% of the overall 262 variance. When allowed to differ between the sexes via a sex-specific heterogenous 263 (co)variance structure (assuming cross-sex correlations of zero), inter-individual variance 264 tended to be larger in males than in females (Table 1; LRT χ^2 = 16.6, df = 1, p < 0.001). 265 Residual variance was also heteroscedastic between the sexes (Table 1; LRT χ^2 = 34.3, df = 1, 266 p < 0.001.) Due to our pilots reasonably small sample size (sample size can easily be 267 268 increased by extending the number of batches; also see General remarks), this result should however be treated with caution as the model had problems reaching convergence. 269

Males and females differed in their overall locomotor activity, with males being notably more mobile, in addition to exhibiting greater variance in activity (Table 1 and Fig. 2B). Flies not moving at all may be indicative of fly exhaustion linked to prolonged handling. We tested for this by checking whether the fraction of flies not moving at all in the locomotion assay (distance = 0) differed between batches of flies tested as first or second. Fraction of zero-distance individuals differed between sequence positions (0.6% for flies tested first for locomotion, 8.9% for flies tested second; $\chi^2 = 76.8$, p < 0.001, df = 1). However, when

- 277 included in the linear model for distance covered, sequence positions did not generate any
- 278 differences (estimate: -0.03, *t* = -0.22, *df* = 350, *p* = 0.81).

279

Y-maze

282	Pure repetition (e.g., LLL, RRR) and alternation (e.g., LRL, RLR) behaviours were the most
283	prevalent (Fig. 3A), with alternations having the highest frequency of all trigram types.
284	Both proportions of repetitions and alternations were highly variable at the between-
285	individual level (Fig. S2). They also exhibited a high fraction of zeroes (i.e., individuals that
286	did not exhibit repetition or alternation behaviours; see General remarks for possible
287	discussion of this issue, and the Discussion for details; Fig. S2).
288	Average locomotor activity tended to be positively associated with the overall proportion of
289	repetition trigrams (Fig. S3 A), a pattern that was absent for the proportion of alterations
290	(Fig. S3 B).
291	When expressed as the absolute number of alternation behaviours (rather than as a
292	proportion), the occurrence of alternation patterns was strongly positively related to
293	average locomotor activity of individual flies, and the strength of this relationship varied
294	between sexes (Table 2, Fig. 3B).
295	Sexual dimorphism was visible in all types of Y-maze behaviour (Fig. S4). Interestingly, the
296	sexes tended to differ in the extent of biased handedness (i.e., asymmetry in choosing left
297	versus right runs in the Y-maze; Fig. S4 C). Batches of flies tested in the Y-maze as second in
298	the sequence tended to be more explorative as they produced significantly more alteration
299	trigrams (GLM with Poisson error: 1.72, $Z = 19.42$, $p < 0.001$) and repetition patterns (0.53, Z
300	= 7.34, <i>p</i> < 0.001).

302 Light-off startle

Patterns in startle response habituation were not as clear as in the other two measured
variables. Most importantly, in over 60% of cases the flies did not exhibit any measurable
movement response to the light-off stimulus (i.e., the movement score was zero across all
three light-off pulses). Subsequent analyses included only individuals that moved
significantly in any of the three pulses (i.e., that had non-zero movement distance when
summed across the three light-off pulses).

309 The overall pattern indicated that many individuals, according to expectation, decreased

their movement response during the second exposure to the light-of stimulus. However,

this "habituation" trend was rarely maintained during the third exposure (Fig. 4, see also Fig.S5).

Formal analysis using the post-stimulus distance data revealed no clear decreasing linear trend among the three consecutive stress exposures (Table 3). Males did not differ from females in their average startle response nor did the sexes differ in their habituation slope (Table 3).

Although a random slopes analysis might not be numerically stable and robust with the sample size used in this pilot, a simple extension of a mixed model used to analyse habituation to include random slopes (i.e., slopes of the habituation response randomly varying between measured individuals) indicated the existence of non-negligible individual variation (LRT comparing the intercept-only and random slopes models: $\chi^2 = 10.36$, df = 1, p= 0.005).

323 Discussion

Our pilot study and its results indicate that certain types of behaviour in small model 324 325 organisms such as *Drosophila melanogaster* are amenable to automatic phenotyping and 326 can become a promising, new avenue in high-throughput analysis of animal behaviours. 327 Here we show that coupling accurate movement tracking with custom-made scripting 328 language creates a powerful system capable of measuring and recording several types of 329 behaviours that can be relevant in evolutionary and ecological contexts. Our study also 330 indicates that repeated measurements on the same individuals are logistically and 331 technically possible (but also see General remarks for comments on how to improve this 332 aspect of a *Drosophila melanogaster* high-throughput phenotyping study). To date, similar 333 analyses were done on considerably smaller numbers of individuals, mostly due to technical considerations. Increasing numbers of assayed individuals brings such assays to a 334 335 completely new level, where evolutionary questions centred on between-individual 336 variability and its components can be addressed. Note that the number of individuals used 337 here were designed as a pilot study and it is possible to increase the sample size even further in more formal, hypothesis-driven studies (see General remarks). 338 339 Of all three applied assays, the general locomotor activity assay was the most 340 straightforward and generated the best quality data. It also exhibited the highest levels of between-individual differentiation and lowest levels of potentially difficult statistical issues 341 (such as zero-inflation of the response). This can, in part, be due to the simplicity of this 342 343 assay, but also due to the relatively longer measurement period (i.e., ensuring that at least 344 some non-zero amount of the measured behaviour is observed during the assay). Our results clearly suggest that there are sexual differences both in the average levels of this 345 346 behaviour, and in the between-individual variability (repeatability). Locomotor activity can be used as a measure of exploratory behaviour³². Its deterioration could also be used in 347

toxicity, thermal limits, aging, and other similar assays where mobility and movement
patterns of an organism are used as quantitative proxies of an individual's response to
experimentally applied stress.

351 The Y-maze exploration assay demonstrated that the proposed method also allows for real-352 time processing of the behavioural data, e.g., to generate Y-maze zone data. Such assays could easily be expanded (e.g., by equipping the Y-maze with olfactory and other sensory 353 354 stimuli), providing high-throughput ways of assaying learning behaviours, aversion responses or effectiveness of knockouts targeted to specific sensory-regulating genes^{33,34}. 355 356 Our Y-maze output data were considerably zero-inflated, which could be an intrinsic feature of the data (and hence could be modelled with appropriate zero-inflated 357 Poisson/binomial/beta distributions) or could result from technical considerations (see 358 General remarks). Nevertheless, Y-maze exploration indices tended to correlate with overall 359 levels of movement activity of the flies, suggesting consistency in behavioural patterns 360 361 measured by different assays.

362 The most challenging assay, the light-off startle analysis, gave the weakest and most 363 ambiguous results. Surprisingly, a large portion of flies in this assay did not react to the stressful light-off stimulus at all, raising questions about the generality of this assay as 364 presented in earlier studies²⁵. Part of the lack of uniformity in the data from this assay and 365 366 previously published accounts could be due to the flies not being entirely synchronised in 367 terms of their age (i.e., while all the eggs used to generate the focal flies were laid within 24 368 hours of each other, flies could have varied in the time it took for them to eclose into adults). Youth and uniform age seem to be some of the more important factors in 369 determining the success of the light-off response habituation assay²⁴. 370

371	Because the studied traits were relatively simple, we do not argue that they represent a
372	one-to-one correspondence with complex behaviours occuring in wild contexts. Characters
373	considered in our paper should be regarded as proxies, or isolated components, of more
374	complex behaviours. Although likely not seen in natural settings (e.g., exploration of a small,
375	confined space; movement in a narrow, unbaited triple-arm maze), these traits represent
376	simplified behavioural units that expose variation underlying more complex behaviours.
377	More importantly, all measured traits seem to be weakly to moderately linked together
378	(Table 4, all correlations but one are statistically significant) – which supports their
379	relevance, and emphasises the potential of the proposed protocol in identifying broader
380	behavioural syndromes.
381	General remarks
382	Our report clearly emphasizes that high-efficiency phenotyping of labile characters – such as
383	multiple behavioural traits – is possible even in small and delicate organisms such as the
384	fruit fly. Combining automatic tracking in real-time with carefully designed miniature arenas
385	and flexible experiment-scripting language, as applied in our study, provides a new
386	approach to the challenge of large, high-throughput phenotyping of individual small animals
387	such as Drosophila melanogaster (commonly used due to their fast lifecycle). Following our
388	tests, we have further refined the methods applied in this study to increase the
389	reproducibility, efficiency, and ease of the performed procedures. Some of the
390	improvements that could be considered include:
391	- Using inbred or otherwise isogenic lines of flies instead of repeatedly testing
392	individual animals. Repeated tests on individuals (e.g., performing the same assay

- 393 across multiple time points in their lifespan or performing sequential tests on
- 394 specific individuals) can be problematic as flies are delicate and easy to damage

during transfer between different arenas. Such handling also requires repeated 395 exposure to cold or CO₂ to induce a temporary coma. Our data indicate that some 396 397 fatigue effects may be present (flies tested for locomotion at the end showed an excess of individuals not moving at all), but effects are not as obvious as one might 398 expect (flies tested last in Y-mazes showed more exploratory behaviour). However, 399 similarly replicated data can be obtained by replacing individual (i.e., genetically 400 401 distinct) flies with multiple individuals coming from inbred lines, e.g., the Drosophila Genetic Reference Panel ^{35,36}. In such a case, individual flies from one line can be 402 deemed unique genotypes (e.g., 'individuals' if maintained under identical 403 404 conditions) and so can be tested only once in a given type of behavioural assay. Considerable time can be saved by housing flies in individual vials divided into 405 406 complete batches to be assayed on a given testing occasion. Doing so greatly reduces the time needed to isolate the required pool of tested flies and removes the 407 408 need of anaesthetising a large population of flies to subsample it for a specific number of individuals. Such a strategy also makes it easy to age-synchronise all 409 410 individuals, which may be key to reducing nuisance variability and improving the reproducibility of certain tests (e.g., the light-off startle assay ²⁴). 411 Refraining from using CO² to anesthetise flies can reduce the time needed for the 412 animals to regain full activity³⁷. Instead, using an ice bath to stun the flies and cool 413 down all working surfaces (e.g., the arenas/wells into which the flies are loaded) 414 provides enough immobilisation without compromising flies' activity or neuronal 415 performance in the longer term. 416

Flies should be assayed outside of their mid-day low-activity period which may be
 difficult when performing large numbers of assays on many flies. To avoid this, we

propose ensuring that batches are run early morning and late afternoon or by 419 batching the flies into several smaller cohorts maintained at 12:12 photoperiods 420 421 shifted in relation to each other by a few hours. By shifting the photoperiod, flies are 422 always tested in their 'morning' even when the assays are run over midday. We suspect that increased zero-inflation (i.e., excess of immobile flies) observed in 423 several variables measured in this study could be due to assaying some of the flies 424 425 after their morning activity peak. Targeting the (relative) morning hours in all 426 measured flies and standardising the measurement time window with respect to the fly circadian cycle should provide improvements in their overall activity levels and 427 428 eliminate unnecessary sources of trait variability.

Using mutants altered in target traits linked to the assayed behaviours could be an
 effective validation method that would enable stricter identification of behaviour
 components playing role in our protocols.

432 In summary, we demonstrate that the fruit fly – a fast growing and fast reproducing model organism – provides a study system highly suitable for large, high-throughput phenotyping, 433 which will likely lie at the centre of variation-focused behavioural biology. We provide a 434 suite of simple assays that can easily be conducted in most *Drosophila melanogaster* 435 laboratories with minimum adjustments to standard procedures and protocols. Proposed 436 methods are cost-effective, logistically flexible, and can be modularised to improve their 437 438 robustness. We believe that further development of such techniques will soon become one 439 of the key elements of evolutionary and behavioural biology, where targeting variancerelated questions requires sample sizes and replication levels not achievable with 440 441 conventional phenotyping approaches.

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451 Declaration of interest statement

452 The authors have no competing interests to declare.

453 Data availability statement

- 454 All data and code can be found on Github at
- 455 https://github.com/elmacartney/Dmel_methods

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547 Appendices

- 548 Fig. S1. Variation in individual locomotor activity in all assayed flies. Each vertical line
- 549 represents the maximum and minimum activity score from 5 quantification repeats included
- in the assay. Dots are outliers (i.e., observations outside of the 1.5*IQR).
- 551 Fig. S2. Distribution the observed counts of repetition and alternation trigrams from the Y-
- 552 maze assay.
- 553 Fig. S3. Sex-specific relationships between the proportions of repetition and alternation
- 554 trigrams in all possible trigrams (from the Y-maze assay), and the average locomotor activity
- of each individual (from the locomotor activity assay).
- 556 Fig. S4. Sex-specific patterns in Y-maze behaviour. The plot shows male and female
- 557 proportions of repetition (A) and alternation (B) trigrams, as well as sex-specific handedness
- 558 (C; zero means perfect symmetry and random choice of turn direction, negative values
- 559 signify right-turn bias).
- Fig. S5. Light-off startle behaviour averaged over all assayed flies. Points are arithmetic
 means of distance covered at each stimulus exposure (1st, 2nd, 3rd; x-axis), bars represent
 95% confidence intervals.

563

564 Tables

565	Table 1 . Results from a general linear mixed model looking at patterns driving variation in
566	overall locomotor activity. Fixed effects estimates are provided with SEs and relevant p-
567	values. Random effects estimates are reported as relevant SD. Heteroscedasticity was
568	tested using the multiplicative parametrisation (i.e., female residual SD equals 0.612 x 1
569	while male residual SD equals 0.612 x 1.248 = 0.7638).

Fixed effects					
Term	Estimate	SE	df	t	p-value
Intercept	-0.355	0.280	1424	-1.265	0.206
Sex (Males)	0.471	0.077	351	6.092	<0.001
Date (Day 2)	0.328	0.395	2	0.830	0.494
Random effects					
Term	SD				
Individual ID (Males)	0.506				
Individual ID (Females)	0.713				
Residual	0.612				
Residual heteroscedasticity					
Females	1.000				
Males	1.248				

- **Table 2.** Generalised linear model analysis (Poisson error distribution) of the total
- 573 alternations' count in relation to sex and average individual locomotor activity. All effects
- are reported with their SE, Z statistics and p-values.

	Estimate	SE	Z	p-value
Intercept	0.294	0.183	1.604	0.109
Sex (Males)	0.793	0.199	3.978	<0.001
Average locomotor activity	0.006	0.001	5.928	<0.001
Sex * Locomotor Activity	-0.005	0.001	-4.783	<0.001

- 576 Table 3. General linear mixed model analysis of variation in "habituation" slopes. Fixed
- 577 effects' estimates are reported with relevant SE, t-statistics, and p-values. Random effects
- 578 are provided as standard deviations.

Fixed effects

Term	Estimate	SE	df	t	p-value
Intercept	3.308	1.028	216	3.218	0.002
Sex (Male)	0.890	1.534	104	0.581	0.563
Startle stimulus sequence no.	0.314	0.482	216	0.651	0.516
Sex * sequence no.	-0.811	0.727	216	-1.116	0.266

Random effects

Term	Estimate
Individual ID (Intercept)	5.281
Individual ID (habituation slope)	2.579
Intercept-Slope correlation	-0.939
Residual	3.881

579

- 580 Table 4. Correlations (with 95% CI intervals) for a sample of 4 measured traits,
- 581 demonstrating the potential of identifying behavioural syndromes.

582

	Locomotory activity	Proportion of alternations	Proportion of repetitions	Handedness
Locomotory activity	1	-	-	-
Proportion of alternations	-0.07 [-0.13; - 0.01]	1	-	-
Proportion of repetitions	0.12 [0.07; 0.19]	-0.04 [-0.10; 0.03]	1	-
Handedness	-0.14 [-0.21; - 0.09]	0.10 [0.04; 0.17]	-0.18 [-0.23; - 0.11]	1

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589 Figures



590

591 **Fig. 1.** Overview of the experimental setup demonstrating the steps undertaken with each batch of

592 45 flies.



Fig. 2. Variability in the average locomotion activity (mean of five measurement bins) of *Drosophila melanogaster*. The figure shows the histogram of activity for all flies (A) and sex-specific patterns in
 average locomotor activity (B). In both cases the response was log-transformed.



Fig. 3. The distribution of all possible types of trigrams in the Y-maze exploration assay (A) and the
relationship between the count of alternation trigrams (LRL and RLR) and the average locomotor
activity of each fly (B).



Fig. 4. Habituation to light-off startle response in a sample of flies that exhibited non-zero locomotor
activity in at least one of the three measurement timesteps. Horizontal axis of each subplot specifies
the three subsequent measurement events; vertical axis presents total distance travelled in the time
interval used to register the light-off startle response (measured in number of pixels travelled).