

1 **Tracktor: image-based automated tracking of animal movement and** 2 **behaviour**

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4 Vivek Hari Sridhar^{1,2}, Dominique G. Roche³, Simon Gingins^{1,2}

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6 ¹ *Department of Collective Behaviour, Max Planck Institute for Ornithology, Konstanz, Germany*

7 ² *Department of Biology, University of Konstanz, Konstanz, Germany*

8 ³ *Institute of Biology, University of Neuchâtel, 2000 Neuchâtel, Switzerland*

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10 E-mail for correspondence: vivekhsridhar@gmail.com

11

12 **Abstract**

13 1. Automated movement tracking is essential for high-throughput quantitative analyses of
14 the behaviour and kinematics of organisms. Automated tracking also improves replicability
15 by avoiding observer biases and allowing reproducible workflows. However, few automated
16 tracking programs exist that are open access, open source, and capable of tracking
17 unmarked organisms in noisy environments.

18

19 2. *Tracktor* is an image-based tracking freeware designed to perform single-object tracking
20 in noisy environments, or multi-object tracking in uniform environments while maintaining
21 individual identities. *Tracktor* is code-based but requires no coding skills other than the user
22 being able to specify tracking parameters in a designated location, much like in a graphical
23 user interface (GUI). The installation and use of the software is fully detailed in a user
24 manual.

25

26 3. Through four examples of common tracking problems, we show that Tracktor is able to
27 track a variety of animals in diverse conditions. The main strengths of Tracktor lie in its
28 ability to track single individuals under noisy conditions (e.g. when the object shape is
29 distorted), its robustness to perturbations (e.g. changes in lighting conditions during the
30 experiment), and its capacity to track multiple individuals while maintaining their identities.
31 Additionally, summary statistics and plots allow measuring and visualizing common metrics
32 used in the analysis of animal movement (e.g. cumulative distance, speed, acceleration,
33 activity, time spent in specific areas, distance to neighbour, etc.).

34

35 4. *Tracktor* is a versatile, reliable, easy-to-use automated tracking software that is
36 compatible with all operating systems and provides many features not available in other
37 existing freeware. Access *Tracktor* and the complete user manual here:

38 <https://github.com/vivekhsridhar/tracktor>

39

40 **Keywords:** collective behaviour, fast start escape response, locomotion, kinematics, choice
41 experiment, video analysis software

42 **1. Introduction**

43 Quantifying animal movement is central to answering numerous research questions in
44 ethology, behavioural ecology, ecophysiology, biomechanics, neuroscience and evolutionary
45 ecology. Movement is also increasingly used as a proxy to evaluate sub-lethal effects of
46 pharmaceuticals, climate change and other anthropogenic stressors on animals, thereby
47 improving our understanding of contemporary environmental and health issues of global
48 importance (Wong & Candolin, 2015; Tucker et al., 2018). As such, researchers in diverse
49 fields of science require methods allowing non-invasive measurement of movement, in
50 different environments, across many animals simultaneously, and with high precision and
51 accuracy (Dell et al., 2014).

52

53 The field of computer vision now allows researchers to automatically track subjects in a
54 variety of contexts (Dell et al., 2014). Such automated tracking techniques drastically reduce
55 the time and effort required to extract image-based data, allowing researchers to generate
56 large and highly detailed datasets of animal movement and behaviour (e.g. Mersch, Crespi,
57 & Keller, 2013; Attanasi et al., 2014). Methods for automated tracking abound in the
58 computer vision literature, yet researchers often lack the know-how to implement these
59 solutions for their specific tracking needs. Without a programming background,
60 implementing such techniques can be daunting, and manual coding via direct observation
61 often remains a preferred alternative, particularly for small-scale projects (Gomez-Marin,
62 Paton, Kampff, Costa, & Mainen, 2014). However, while direct observation can produce
63 informative data, it is often time-consuming, subjective, and challenging to replicate.
64 Various proprietary software packages have been developed to provide simple automated
65 tracking options (e.g. Noldus, Spink, & Tegenlenbosch, 2013), but these are expensive and

66 often out of reach for many research groups. Clearly, there is a need for automated tracking
67 solutions that are free, user-friendly and reliably provide image-based data acquisition
68 solutions to record movement and behaviour.

69

70 A variety of open access tracking software already exist, designed to provide solutions for
71 generic (e.g. Pérez-Escudero, Vicente-Page, Hinz, Arganda, & Polavieja, 2014; Rodriguez et
72 al., 2018) and specific tracking problems (e.g. Branson, Robie, Bender, Perona, & Dickinson,
73 2009; Hewitt et al., 2018). However, automated tracking is achieved via different
74 approaches, each one with its own strengths and weaknesses. Here, we propose Tracktor, a
75 simple, open-source tracking software that can track multiple individuals simultaneously,
76 single individuals in noisy environments, and produces a range of plots and summary
77 statistics. Below, we detail its functioning, give examples of applications, and discuss its
78 strengths and limitations. We also provide a detailed section on optimising video acquisition
79 in the supplementary material.

80

81 **2. Software**

82 Tracktor is a python-based program that uses the OpenCV library (<https://opencv.org/>) for
83 image processing. Briefly, Tracktor functions as a three-step process (Fig. 1): first, it
84 automatically detects objects in videos through adaptive thresholding, and isolates the
85 animals from other objects based on their size; second, it separates individuals from each
86 other using k-means clustering; third, it maintains individual identities between frames
87 using the Hungarian algorithm (Kuhn, 1955). A detailed description of the software is
88 provided in the electronic supplementary material (ESM). There is no GUI (Graphical User
89 Interface); all the commands are directly sent from scripts. However, all relevant input

90 parameters are clearly explained and grouped in a designated cell at the start of the code,
91 making Tracktor easy to use (Fig. S1). Additionally, the code is separated into sections and
92 thoroughly annotated such that it is clear and accessible, even for an audience with minimal
93 or no coding skills. For example, users with basic knowledge of the R statistical language will
94 readily be able to use Tracktor. A detailed, step-by-step manual guiding users from
95 installation to implementation is available on the Git repository:

96 <https://github.com/vivekhsridhar/tracktor>.

97

98 **3. Examples of applications**

99 We provide four examples of common animal tracking problems that have successfully been
100 solved with Tracktor. Each example is fully replicable and comprises annotated code to track
101 the animals and produce relevant graphs and summary statistics. These examples can serve
102 as the basis for solving similar tracking problems and can be readily adjusted to suit other
103 problems with minimal modifications.

104

105 **3.1 Fish fast-start escape response (1 individual)**

106 Fast-start (i.e. burst) swimming performance in fishes is typically assessed by filming their
107 escape behaviour in response to a mechano-acoustic stimulus falling inside the
108 experimental tank (Domenici & Blake, 1997). When the stimulus hits the water surface,
109 waves are created that propagate across the tank (Fig. 2a). These disturbances are
110 problematic for automated tracking because water movements can be mistakenly detected
111 as the animal, and several traditional image processing techniques (e.g. background
112 subtraction) cannot be used in these conditions. As a result, most researchers code videos
113 manually (e.g. Gingins, Roche, & Bshary, 2017; Shi et al., 2017), which is time consuming and

114 prone to measurement error. Using a high speed (1000 fps) video of an escape response by
115 the cleaning goby *Elacatinus prochilos*, we show that Tracktor can differentiate between
116 animal movements and those of other shapes detected as potential objects. We provide
117 summary statistics that are specific to fast-start analyses, and implement the Savitzky-Golay
118 filter (Savitzky & Golay, 1964) to smooth noise and avoid erroneous measurements (Fig. 3a-
119 c).

120

121 **3.2 Two-choice flume experiment (1 individual)**

122 Researchers in behavioural ecology and ecotoxicology are often interested in measuring an
123 animal's avoidance or attraction to a given stimulus (e.g. olfactory or visual cues). Such
124 choice tests are generally carried out by placing the test subject in a central arena with
125 various choice options on different sides. The time spent in the vicinity of each option is
126 then used as a proxy to measure preference or avoidance. Several experiments are based
127 on this approach, for example to evaluate mate choice, kin and individual recognition, social
128 tendencies, numerical abilities, and reaction to chemical cues (e.g. Hurst et al., 2001;
129 Gómez-Laplaza & Gerlai, 2011; Fischer & Frommen, 2013; Jutfelt, Sundin, Raby, Krång, &
130 Clark, 2017; Macario, Croft, Endler, & Darden, 2017). Here, we use Tracktor to track the
131 movements of a single cleaning goby (*Elacatinus prochilos*) in a two-current choice flume
132 with water containing the smells of different habitats. We first track the movements of the
133 fish throughout the video. Then, we select an area by drawing a rectangle on the image, and
134 measure the proportion of time the fish spent in the designated area (Fig. 2b).

135

136 **3.3 Spider courtship behaviour (2 individuals)**

137 Researchers often place two or more individuals in the same arena and record dyadic
138 interactions, such as approach and avoidance behaviours. Such tracking problems are
139 common in mate choice experiments, aggression, and sociability tests (Jolles, Boogert,
140 Sridhar, Couzin, & Manica, 2017; Macario et al., 2017). Here we simultaneously track a male
141 and a female spider (*Nephila senegalensis*) during courtship and provide distance between
142 the two. In this example, Tracktor simultaneously tracks individuals of very different sizes
143 (male *Nephila* are much smaller than females) and is robust to disturbances in the system.
144 The key to successfully tracking objects of different sizes simultaneously is to ensure that
145 the recording area is uncluttered since size thresholding cannot be used to its fullest extent
146 when non-target objects of similar sizes are present in the arena (see ESM). Towards the
147 end of the video sequence, the experimenter enters the field of view, casting a shadow on
148 the setup. This does not affect tracking performance because Tracktor uses adaptive
149 thresholding to cope with changes in lighting while disregarding the experimenter based on
150 a specified size rule (see ESM).

151

152 **3.4 Termite collective behaviour (8 individuals)**

153 Studying collective behaviour requires recording the movements of multiple individuals
154 simultaneously, resulting in large datasets. Automated tracking is extremely useful in this
155 context, and many researchers already use this approach for their research by designing
156 their own custom tracking solutions (Tunstrøm et al., 2013; Jolles et al., 2017). Here, we
157 show that Tracktor can simultaneously track 8 termites (*Constrictotermes cyphergaster*) in a
158 petri dish while maintaining individual identities. The challenge in this example is to
159 correctly identify individuals: when two termites perform trophallaxis, thresholding results
160 in the merging of their shapes. The k-means clustering technique implemented in Tracktor

161 (see ESM) allows separating contours into a fixed number of clusters corresponding to the
162 number of individuals in the video. Summary statistics are provided as standard measures of
163 collective states such as polarisation and rotation of the group (Fig. 2e; see Couzin, Krause,
164 James, Ruxton, & Franks, 2002; Tunstrøm et al., 2013).

165

166 **4. Strengths & limitations**

167 Tracktor aims to provide an automated tracking tool that can be applied to many animals
168 without the use of tags. Similar types of software (e.g. idTracker, Pérez-Escudero et al.,
169 2014; BioTracker, Mönck et al., 2018; ToxTrac, Rodriguez et al., 2018) already exist, and we
170 do not claim that Tracktor outperforms other software in all conditions. Rather, we
171 emphasise that tracking problems and solutions are numerous and varied; hence, Tracktor
172 provides an alternative that can fill the needs of many researchers when other software are
173 not adequate. For instance, to our knowledge, Tracktor is the first tool to provide a robust
174 and complete automated analysis of a fish fast-start escape response.

175

176 One of the common computer vision techniques to detect objects in videos is background
177 subtraction, which relies on the assumption that the background is static while the animals
178 to be tracked are mobile. Background subtraction has the advantage that it can deal with
179 complex backgrounds and uneven illumination. However, if the animals remain static for a
180 significant amount of time, detection will fail. For example, one of the eight termites
181 (example 3.4) remained immobile throughout the recording. Additionally, it is not possible
182 to use this technique when the background itself is moving, such as when measuring escape
183 responses in fish (example 3.1). Under such conditions, Tracktor is thus likely to perform
184 better than software using background subtraction (e.g. BioTracker).

185

186 We recognize that Tracktor's lack of a GUI might initially discourage potential users from
187 using the software. However, the growing prevalence of basic scripting/programming skills
188 among evolutionary ecologists (e.g. in R) suggests that the absence of a GUI will not
189 constitute a major impediment to most interested users, particularly given the simplicity of
190 Tracktor's code and the detailed user manual we provide alongside the software. A (python)
191 code-based program also offers the additional advantage that it can easily be modified and
192 improved by the community and run on all operating systems, which is often not the case
193 for software with a GUI (e.g. ToxTrac).

194

195 Like many other software packages, Tracktor will not perform well if there are occlusions, or
196 if animals go in and out of the camera's field of view. Solving such problems requires
197 algorithms that maintain identity by recognising individual features (e.g. idTracker) or assign
198 IDs to tracklets when tracks are broken (e.g. ToxTrac). Such solutions require more
199 computing power and result in longer processing times. Therefore, they were not
200 implemented in the design of Tracktor.

201

202 **5. Conclusion**

203 Tracktor is a simple, rapid and versatile solution for many commonly encountered image-
204 based tracking problems in animal behaviour. The software is robust to disturbances and
205 allows users to track one or many individuals under a range of experimental conditions.
206 Automated image-based tracking solutions such as Tracktor can greatly improve the study
207 of animal movement and behaviour by reducing video processing time, avoiding observer
208 bias, and allowing transparent, reproducible workflows.

209

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214 software and provided feedback.

215

216 **Authors' contributions**

217 All authors designed and planned the study and contributed in writing the manuscript.
218 V.H.S. developed the software.

219

220 **Data accessibility**

221 The data used here are available at <https://github.com/vivekhsridhar/tracktor>.

222

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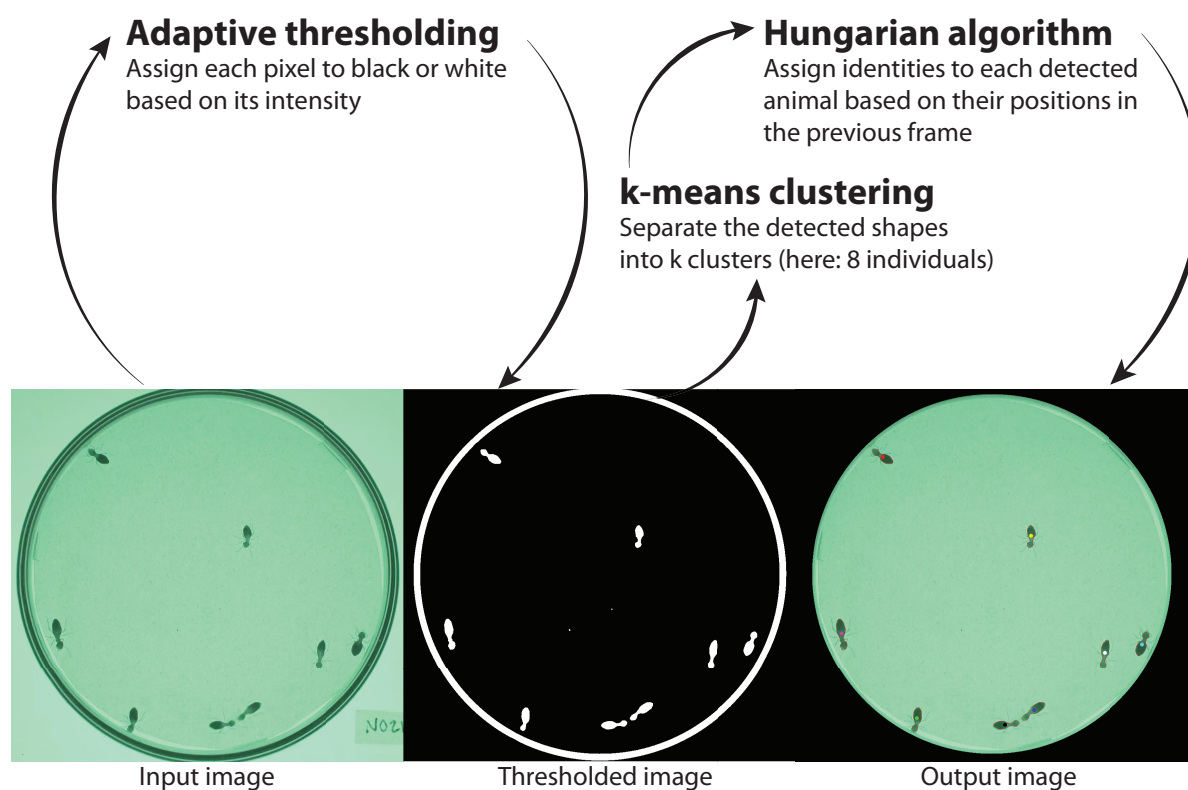
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302
303

304 **Figures**

305



306

307

308 **Figure 1.** Overview of the different stages of the tracking procedure implemented in

309 Tracktor, illustrated here for tracking 8 termites in a petri dish (example 3.4). First, adaptive

310 thresholding is used to binarise the input image. A size rule is used to identify objects of

311 interest (here, the largest object, the petri dish, is eliminated based on the size rule).

312 Second, k-means clustering is used to separate objects if they are merged in the thresholded

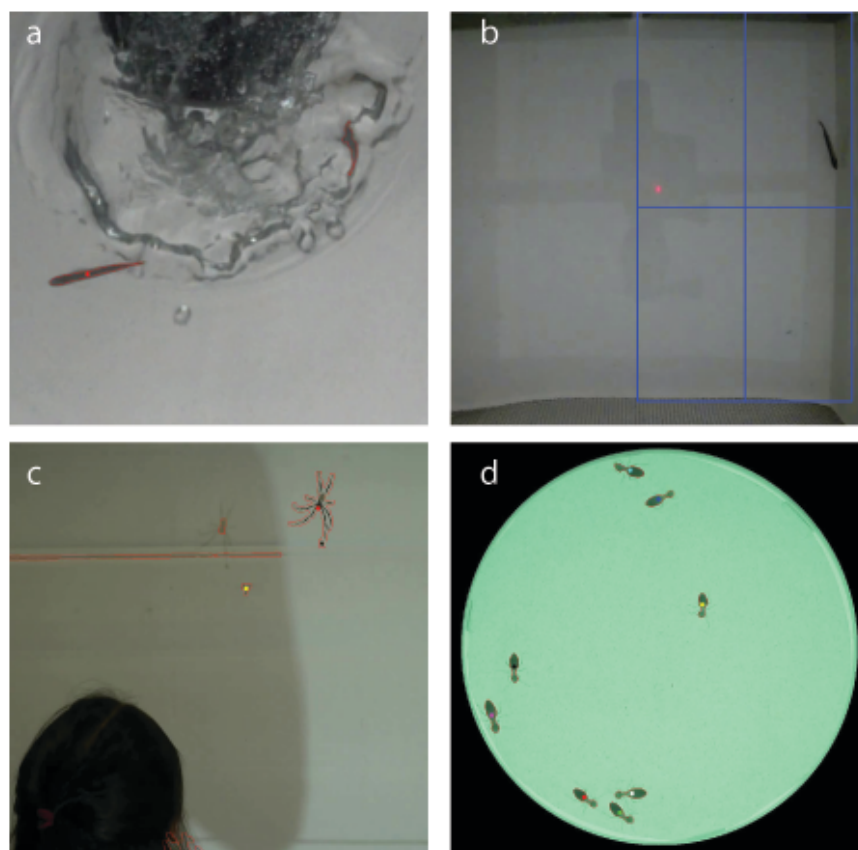
313 image. Third, individual identities are assigned using the Hungarian algorithm. Some steps

314 are skipped depending on the tracking problem (e.g. k-means clustering is not implemented

315 in examples 3.1, 3.2 and 3.3). See the ESM for a more detailed description of the software

316 design.

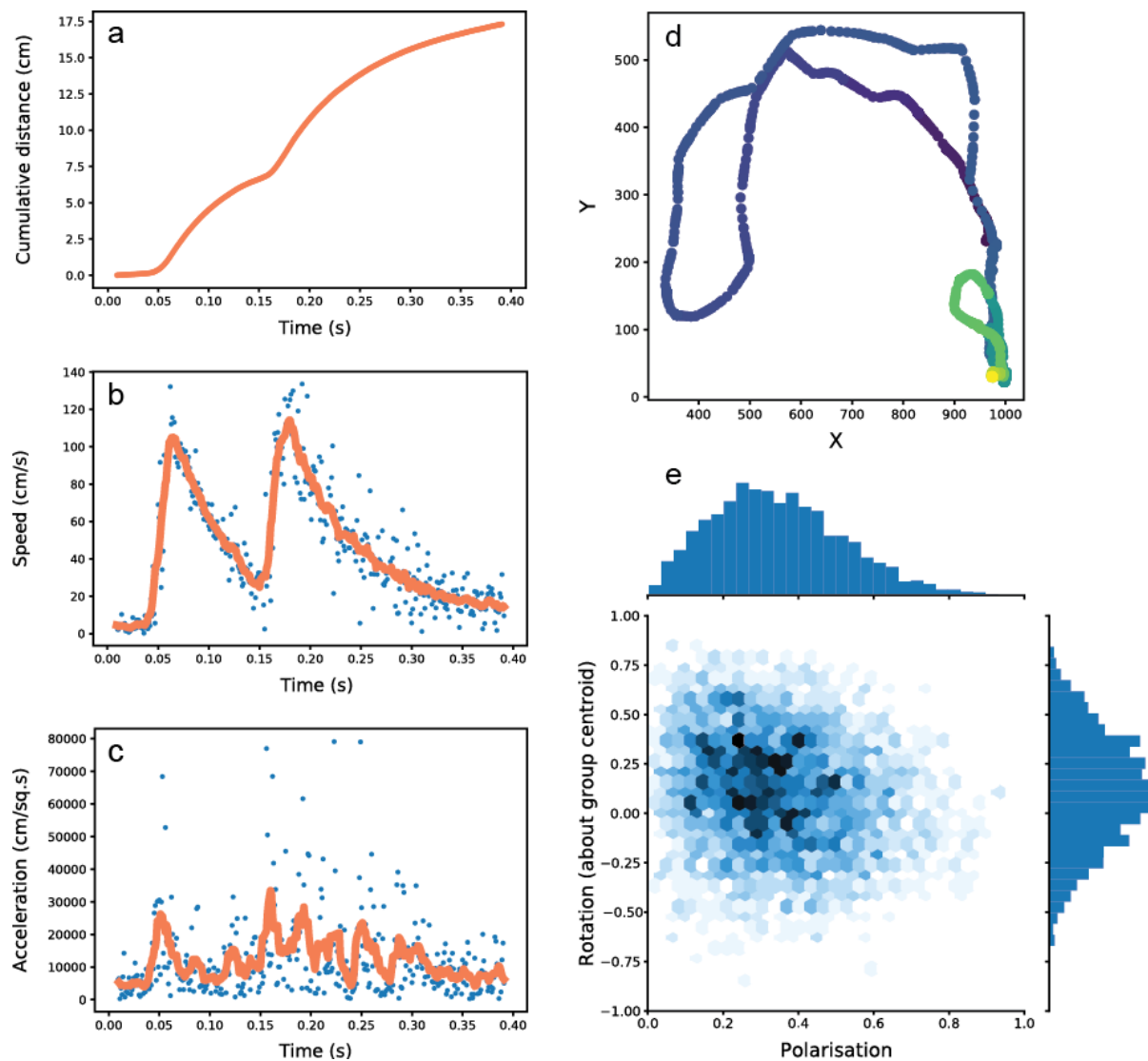
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319

320 **Figure 2.** Examples of four different tracking problems solved with Tracktor: (a) a fast-start
321 escape response by a single fish in an environment disturbed by a falling stimulus (example
322 3.1); (b) a single fish in a two-choice flume experiment with the region of interest depicted
323 as a blue rectangle (example 3.2); (c) two spiders engaging in courtship in an environment
324 disturbed by the experimenter (example 3.3); (d) collective behaviour of eight termites in an
325 undisturbed environment (example 3.4).



326

327

328 **Figure 3.** Example plots of summary statistics outputted by Tracktor. For the fish fast-start
329 escape experiment (example 3.1), panels illustrate: **a)** cumulative distance covered [cm] vs.
330 time [s], **b)** speed [cm s^{-1}] vs. time [s], and **c)** acceleration [cm s^{-2}] vs. time [s]. Blue dots are
331 the raw data; orange lines represent the smoothed values of speed and acceleration using
332 the Savitzky-Golay filter. Panel **d)** is the track [XY positions in pixels] for the flume choice
333 experiment (example 3.2). Panel **e)** shows rotation vs. polarisation to assess the group's
334 dynamical state in the termite collective behaviour experiment.