

1 **Title page**

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3 A miniaturized threshold-triggered acceleration data-logger for recording burst
4 movements of aquatic animals

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35

36 **Abstract**

37 Animal-borne accelerometers are effective tools for quantifying the kinematics of
38 animal behaviors, such as swimming, running, and flying, under natural conditions.
39 However, quantifying burst movements of small and agile aquatic animals (e.g., small
40 teleost fish), such as during predatory behavior, or while fleeing, remains challenging.
41 To capture the details of burst movements, accelerometers need to sample at a very high
42 frequency, which will inevitably shorten the duration of the recording or increase the
43 size of the device. To overcome this problem, we developed a high-frequency
44 acceleration data-logger that can be triggered by a manually-defined acceleration
45 threshold, thus allowing the selective measurement of animal burst movements. We
46 conducted experiments under laboratory and field conditions to examine the
47 performance of the logger. The laboratory experiment using red seabream (*Pagrus*
48 *major*) showed that the new logger could measure the kinematics of their escape
49 behaviors (i.e., body beat cycles and maximum acceleration values). The field
50 experiment using free-swimming yellowtail kingfish (*Seriola lalandi*) showed that the
51 loggers trigger correctly (i.e., of the 18 burst movements, 17 were recorded by the
52 loggers). We suggest that this new logger can be applied to measure the burst
53 movements of various small and agile animals, whose movements may be otherwise
54 difficult to measure.

55

56 **Keywords:** accelerometer, bio-logging, escape, fast-start, feeding strike, telemetry

57

58 **Introduction**

59

60 Animal-borne accelerometers have been used to estimate energy expenditure (Murchie
61 et al., 2011; Payne et al., 2011), activity patterns (Kawabe et al., 2004; Payne et al.,
62 2016; Sato et al., 2007), and specific behaviors such as the feeding, mating, and
63 spawning (Føre et al., 2011; Tsuda et al., 2006; Watanabe and Takahashi, 2013; Whitney
64 et al., 2010) of aquatic animals. In general, accelerometers record acceleration in a
65 continuous manner at a defined frequency (e.g. 1–100 Hz) or record a defined
66 time-average of the acceleration, either digitally stored or transmitted (Cooke et al.,
67 2016). Subsequently, these acceleration data are often transformed to various
68 components (e.g., dynamic and static accelerations) to estimate the energy budgets and
69 activity, or to carry out classification into more detailed behaviors (Gleiss et al., 2011;
70 Shepard and Wilson, 2008; Tanaka et al., 2001; Wilson et al., 2006).

71 Most previous studies have focused on routine behaviors such as cruising,
72 gliding, and resting. Thus, the sampling frequencies of the accelerometers are typically
73 below 32 Hz (Kawabe et al., 2004; Murchie et al., 2011; O'Toole et al., 2010; Tsuda et
74 al., 2006). However, such low sampling frequencies cannot be used to measure the
75 detailed burst movement dynamics of small and agile animals (e.g., teleost fish), which
76 occur over short time scales (i.e., in the order of 100 ms), despite the fact that these
77 burst movements include ecologically important behaviors such as escape responses and
78 feeding strikes. The *in situ* measurements of such behaviors would provide novel
79 insights into the movement performance, energy expenditure, and survival strategies of
80 animals in complex natural habitats.

81 Recently, Broell et al. (2013) demonstrated that an accelerometer with a

82 sampling frequency of at least over 30 Hz (ideally, 100 Hz) is required to identify the
83 escape responses and feeding strikes of the sit-and-wait predator *Myoxocephalus*
84 *polyacanthocephalus*. Moreover, accelerometers with a sampling frequency of 200 Hz
85 have been demonstrated as useful in distinguishing the feeding behavior of trophic
86 generalist fish on different prey types (Horie et al., 2017; Kawabata et al., 2014). These
87 studies clearly show that the high-frequency accelerometers are useful in measuring the
88 burst movements of agile animals; however, such high frequency sampling rapidly
89 consumes electricity and memory, which inevitably shortens the duration of the
90 recording, or increases its size (e.g., 10 hours in 100 Hz and 135 mAh, ORI400-D3GT
91 from Little Leonardo Co., Tokyo, Japan). Thus, applying high-frequency accelerometers
92 to field studies remains a challenging task.

93 To overcome this problem, we developed a data-logger, which selectively
94 records acceleration signals based on a manually-defined threshold (Event logger),
95 which reduces electricity and memory requirements, and thus enables us to measure the
96 burst movements of animals for a relatively long period (e.g., 5-day battery life with a
97 sampling frequency of 500 Hz and 10 burst movements per day) despite its small size
98 (7.7 g in air). A similar selective recording system was used to measure the predatory
99 behavior of the piscivorous pike *Esox lucius* in a laboratory setting (Van Deurs et al.,
100 2017). However, the details of the logger system, the measurement performance of the
101 logger, and its practicality for field measurements were not provided. In the present
102 study, we describe the selective recording system of the Event logger, present results
103 from the laboratory performance test conducted on the escape response of the red
104 seabream *Pagrus major*, and show the results of the field performance test conducted on
105 the yellowtail kingfish *Seriola lalandi*.

106

107 **Materials and Methods**

108

109 **Event logger system**

110 The Event logger consists of two different types of 3-axis accelerometers. One
111 accelerometer detects the threshold excess (Detection accelerometer) and the other one
112 records data (Recording accelerometer), as shown in Fig. S1a. The Detection
113 accelerometer is continuously active, while the Recording accelerometer is inactive
114 unless the Detection accelerometer detects any burst movements signals (i.e., exceeding
115 a set threshold). The Detection accelerometer measures accelerations in the rate of 400
116 Hz. One absolute value is manually set as a threshold (any value from 1.00 to 3.95 g),
117 and applied to the absolute values of all 3-axes accelerations. Once the Recording
118 accelerometer becomes active, it records data for a manually set time period (any time
119 period, in the order of 1 s). Then, it reverts to being inactive. If the acceleration exceeds
120 the threshold when the Recording accelerometer is still active, the Detection
121 accelerometer ignores it. The measurement range of the Recording accelerometer is \pm
122 16 g with 16 bit resolution, and its sampling frequency can be set manually from 1 to
123 1,000 Hz. The battery life of the Event logger depends on the activation frequency and
124 the recording period per activation (e.g., 5-day battery life with 10 activations per day,
125 and a recording of 10 s per activation). The size, mass, and discharge capacity of the
126 Event logger is $29 \times 11 \times 15$ mm, 7.7 g, and 85 mAh, respectively (Fig. S1b). For
127 interested users, the Event logger is available at <http://www.biologging-solutions.com>.

128

129 **Calibration of Detection accelerometer in the Event logger**

130 Inherently, accelerometers have slightly different raw values among units, and the bias
131 level can drift by the process of production, temperature fluctuation, large shock, etc.
132 Therefore, the accelerometers need to be calibrated in advance. We calibrated the
133 Recording accelerometer of the Event logger by referencing the gravitational
134 acceleration and calibrated the Detection accelerometer by the method described below.

135 To find the formula for calibrating the Detection accelerometer, we examined
136 the correspondence between actual acceleration values and the occurrence of the Event
137 logger's activation. The Event logger was attached to a similar-sized conventional
138 3-axis acceleration logger, which was continuously active (hereafter, Reference logger;
139 $29 \times 11 \times 15$ mm, 7.2 g, Biologging Solutions Inc., Tokyo, Japan). The sampling
140 frequencies of both loggers were set to 500 Hz. These two loggers were packaged and
141 thrust, by hand, at various intensities (approximately 1–6 g). We set four different
142 threshold values, namely 1.6, 2.4, 3.2, and 3.95 g, in the Event logger, and thrust 50
143 times in each threshold value. Logistic regression analysis was used to obtain the
144 calibration formula. The occurrence of the Event logger's activation was designated, in
145 binary, as 1 (activated) and 0 (not activated); these values were used as the objective
146 variable. The set threshold and maximum acceleration values, recorded by the reference
147 logger, were considered as explanatory variables. In addition, we calculated the time lag
148 between the time when the acceleration exceeded the threshold, and the time when the
149 recording was initiated.

150

151 **Experiment 1: Laboratory performance test using red seabream**

152 To examine the measurement performance of the logger, we attached the logger on *P.*
153 *major* in a tank and measured its escape response.

154 Four *P. major* were obtained from a local fish hatchery and transported to the
155 Institute for East China Sea Research at Nagasaki University, Japan. The fish were held
156 in 500 L circular polyethylene tanks (100 cm diameter × 75 cm height), with an aeration
157 apparatus and flow-through seawater at a temperature of 19.9–24.2 °C. The mean body
158 mass and total length of the fish were 2.50 kg (range: 1.96–3.05 kg) and 52.5 cm (range:
159 49.9–54.8 cm), respectively.

160 We attached the logger package incorporating the Event logger and the
161 reference logger onto the *P. major*. The fish were first anaesthetized using 0.1%
162 2-phenoxyethanol; then, the logger was attached using two wiry plastic strings and two
163 2-cm round stainless washers, which served as anchors. The strings were first inserted
164 through the tag, and then through the anterior dorsal musculature, after passing through
165 two syringes. Finally, they were anchored in place by round stainless washers. The
166 tagging procedure never exceeded 1 min. The loggers' x, y, and z axes were aligned to
167 the lateral (rightward), longitudinal (forward), and vertical (upward) coordinates of the
168 fish body, respectively. The sampling frequencies of both loggers were set to 500 Hz.
169 The threshold value of the Event logger was set to 2.0 g, since the excess of this value
170 rarely occurred during the routine movements of other fish species, namely,
171 *Epinephelus ongus* (Kawabata et al., 2014) and *Seriola quinqueradiata* (Noda et al.,
172 2013). The recording period of the Event logger was set to 5 s per activation.

173 The experiment was performed in a 3,000 l circular FRP tank (193 cm diameter
174 × 73 cm height), which was filled with seawater to a depth of 30 cm. The fish were
175 individually introduced into the tank and were allowed to acclimate for approximately
176 20 h. Then, we plunged a hand net into the water, near the fish, and induced the escape
177 response. For each fish, the escape response was elicited 8–10 times, in 30-minute

178 intervals, and a total of 34 trials were carried out across the four fish. To determine the
179 timing at the initiation of the escape response, the fish movements were simultaneously
180 recorded dorsally, by a high-speed video camera (HAS-L1; Detect Co., Tokyo, Japan),
181 at 500 frames s⁻¹. The water temperature was 20.6–22.5 °C.

182

183 Data analysis

184 We used 11 of the 34 trials, while the remaining 23 trials were omitted either because
185 the fish did not show any escape response against the hand net, the response was
186 disturbed by the tank wall or intensive waves, or the hand net obscured the fish body.

187 Maximum and minimum acceleration values, and oscillation cycles, are
188 important variables for estimating locomotor performance and categorizing animal
189 behaviors (Broell et al., 2013; Kawabe et al., 2004). Since latency existed between the
190 initiation of escape response and the recording initiation of the Event logger, we
191 examined whether the latency was short enough to precisely measure these variables. To
192 measure the exact timing of escape response initiation and the recording initiation of the
193 Event logger, we synchronized the acceleration signals of the Event logger to those of
194 the reference logger by finding the optimal time difference with the least squares
195 method. The escape response of fish consists of three distinct stages based on body
196 bends (Weihs, 1973). Peak acceleration usually occurs during the initial bend, or stage 1
197 of the escape response (Domenici, 2009). Therefore, the peak acceleration timings
198 during stage 1 were used in the analysis.

199 To specify the sampling frequency required for the precise measurement of the
200 above variables, we examined the effect of sampling frequency on the measured
201 variables. The accelerations of different sampling frequencies (1–500 Hz) were obtained

202 by downsampling the 500 Hz acceleration signals from the Event logger. The maximum
203 and minimum values of the downsampled accelerations were then compared to those of
204 the raw 500 Hz accelerations. We also calculated the minimum sampling frequency
205 required for detecting the oscillation cycles of the acceleration signals, which usually
206 reflects the tail beat frequencies of fish (Kawabe et al., 2003). At least two points within
207 an oscillation wavelength are necessary to measure the oscillation cycle; thus, we
208 regarded the required sampling frequency as reciprocal to one half of the wavelength. In
209 this analysis, we used x-axis accelerations, since tail beats produce mainly lateral
210 accelerations (Kawabe et al., 2003),

211

212 **Experiment 2: Field performance test using yellowtail kingfish**

213 We examined the accuracy of the Event logger's recording initiation system under
214 natural conditions by attaching an Event logger and reference logger to free-ranging *S.*
215 *lalandi* and comparing the obtained acceleration values. This experiment was conducted
216 off the Neptune Islands Group (Ron and Valerie Taylor) Marine Park, Australia (S35°
217 14', E136° 04') from October to November 2015.

218 Three fish (98, 99, and 102 cm TL) were caught by hand line and tagged on
219 board. Fish were placed on a rubber mat and their head was covered with a wet towel
220 while the gills were continuously ventilated with a saltwater hose. In compliance with
221 local animal ethics procedures, fish were not anaesthetized. We attached the logger
222 package consisting of a radio tag, an Event logger, a reference accelerometer logger
223 (ORI400-D3GT, 12 mm diameter × 45 mm length, 9 g, 135 mAh; Little Leonardo Co.,
224 Tokyo, Japan), and a time-scheduled release mechanism (Watanabe et al., 2004) to each
225 fish. The tagging procedure of the logger package was similar to the Experiment 1. The

226 fish were released promptly after the tagging, and the tagging procedure never exceeded
227 3 min. The sampling frequency, threshold value, and per-activation recording period of
228 the Event logger, were set to 500 Hz, 2.0 g, and 10 s, respectively. The reference logger
229 was set to continuously measure 3-axis accelerations at 20 Hz. Approximately 45, 37,
230 and 18 hours after release, the logger packages popped off from the fish, and emerged
231 afloat for recovery.

232 Two of the three Event loggers worked properly and were thus used in the data
233 analysis. We synchronized the acceleration signals from the Event logger to those from
234 the reference logger using the least squares method. Subsequently, we examined
235 whether the Event loggers were accurately activated, by comparing the activation events
236 with the threshold excess in the acceleration signals obtained from the reference loggers.
237 The maximum and minimum acceleration values, and minimum wavelength obtained
238 through the Event loggers were compared to those obtained through the reference
239 loggers.

240

241

242 **Results**

243

244 **Calibration of Detection accelerometer in the Event logger**

245 In all four set threshold values, the maximum accelerations were higher in the trials
246 where the Event logger was activated compared to trials where the Event logger was not
247 activated. The acceleration range at the 1.6 and 2.4 g thresholds, however, slightly
248 overlapped between the activated and non-activated trials (Table S1). The calibration
249 formula was obtained by a logistic regression and successfully classified 95.5 %

250 (191/200) of thrust events (Fig. S2). The mean time lag between the instance when the
251 acceleration exceeded the threshold and the instance when the recording was initiated
252 was approximately 1.11×10^{-2} s (Table S1).

253

254 **Experiment 1: Laboratory performance test using red seabream**

255 The Event logger initiated recording at $1.64 \times 10^{-2} \pm 1.21 \times 10^{-2}$ s (mean \pm SD) after the
256 initiation of the escape response. In all of the 11 trials, the initiation occurred before the
257 first half of the stage 1 escape response, which indicated that the recording latency was
258 short enough to measure the oscillation cycle of the acceleration signals with precision.
259 Successful recording rates of maximum and minimum acceleration values by the Event
260 logger in the stage 1 escape response are shown in Fig. 1. Of the 11 trials, 11 maximum
261 (100 %) and six minimum (55 %) x-axis accelerations, 10 maximum (91 %) and 5
262 minimum (45 %) y-axis accelerations, and nine maximum (82 %) and two minimum
263 (18 %) z-axis accelerations were successfully recorded by the Event logger. Depending
264 on which of the two came first, the maximum or the minimum acceleration values
265 occurred at $8.55 \times 10^{-3} \pm 2.54 \times 10^{-3}$ s (mean \pm SD) after the escape response initiation.

266 The absolute values of maximum and minimum accelerations became smaller
267 as the sampling frequency decreased. The median value of the downsampled
268 acceleration signals became less than 95% of the original values at 50.0–166.7 Hz and
269 less than 75 % of the original values at 22.7–100.0 Hz (Figure 4). The minimum
270 wavelength median of the acceleration in the x axis was 4.8×10^{-2} s, and the required
271 sampling frequency for detecting the oscillation cycle was estimated at 41.7 Hz.

272

273 **Experiment 2: Field performance test using yellowtail kingfish**

274 Each of the acceleration signals recorded by the two reference loggers, namely,
275 packages 1 and 2, exceeded the set threshold nine times each. The Event logger of
276 package 1 was activated in all nine excesses, while that of package 2 was activated in
277 eight of the nine excesses. The maximum acceleration value of the remaining one
278 excess was 2.42 g. In addition to the nine and eight activations, both Event loggers were
279 activated another five times each, when the acceleration measured by the reference
280 loggers surged, but did not exceed the threshold (mean 1.72 ± 0.20 g, N=10, Fig. 3).
281 With the exception of the maximum x-axis acceleration values and minimum x- and
282 y-axes values in package 2, the maximum and minimum values obtained through the
283 Event loggers were larger and smaller, respectively, than those obtained via the
284 reference loggers (Table S2).

285

286

287 **Discussion**

288 The experiment with red seabream showed that the Event logger recorded accelerations
289 during most parts of the escape responses, with sufficient temporal resolutions.
290 Although there was latency in initiating the recording, which prevented accurate
291 recordings of minimum acceleration values during the stage 1 escape response, the
292 logger successfully recorded the maximum acceleration values and oscillation cycles
293 with high accuracy. The difference in successful recording rates, between maximum and
294 minimum acceleration values, was related to the asymmetrical waveform of the
295 acceleration signals (see Fig. 1). The asymmetrical waveform may be due to the
296 gravitational, propulsive and centripetal accelerations, and to the logger attachment site
297 (i.e., left side of the fish body); however, further research is required to explicitly clarify

298 the cause of this asymmetry.

299 The latency in the record initiation was a combination of time lag from escape
300 initiation to threshold detection and time lag from threshold detection to record
301 initiation. Although the former time lag can be modified by a set threshold value, the
302 latter time lag is mechanically fixed as the mean of 1.11×10^{-2} s (Table S1). Because the
303 first peak value (maximum or minimum) occurred at 8.55×10^{-3} s after the escape
304 response initiation, recording both maximum and minimum values during the stage 1
305 escape response of this species was difficult with the present system. Nonetheless, the
306 time to peak acceleration should be species- and behavior-specific; thus, the latency
307 could be short enough to record both maximum and minimum acceleration values for
308 other studies. Additionally, the latency will be shorter as sensor technology continues to
309 advance.

310 The escape response of red seabream included abrupt acceleration changes
311 and thus more than 166.7 Hz of sampling frequency was required to measure the exact
312 peak accelerations, and more than 41.7 Hz for estimating tail beats. This result is
313 consistent with the conclusion by Broell et al. (2013), by which accelerometers with a
314 sampling frequency of more than 30 Hz are required to identify the escape and
315 predatory behaviors of fish.

316 The results of the field study show that the Event logger was accurately
317 activated in response to the threshold excess. The Event logger failed to detect one
318 threshold excess; however, this could have been caused by a slight positional difference
319 between the Event logger and the reference logger. The Event logger was also activated
320 several times, when the acceleration measured by the reference logger did not exceed
321 2.0 g. Although these activations could be caused by false detections by the Detection

322 accelerometer (see Table S1), it is more likely that the activations are related to the 400
323 Hz samplings of the Detection accelerometer, which allows the detection of momentary
324 high acceleration with higher probabilities than the 20 Hz samplings of the reference
325 logger (See Fig. 2).

326 The Event logger system is similar to on-board data processing, in the sense of
327 discarding unnecessary data before recording. Generally, on-board data processing
328 operates by summarizing the information inside the tags, which allows us to save
329 battery and minimize memory usage. Previous studies demonstrated that the
330 acceleration data could be processed into activity level (Payne et al., 2016) or
331 occurrence of specific behaviors, such as opening jaws, stroking flippers, and
332 borrowings (Adachi et al., 2014; De Almeida et al., 2013; Naito et al., 2013). However,
333 few on-board processing methods have been reported to determine the types of burst
334 movements, possibly due to the low sampling frequencies of accelerometers, and due to
335 the limited numbers of established processing algorithms (but see (De Almeida et al.,
336 2013; Horie et al., 2017). In addition, such processing would limit the scope of analysis,
337 since it cannot provide the various kinematic variables simultaneously. In contrast, the
338 Event logger works with a simple algorithm, which records certain raw acceleration
339 data with a high sampling frequency, and thus allows post-hoc analysis from various
340 kinematic aspects. Such raw high-frequent acceleration data would be especially useful
341 for exploring burst behaviors, which have not been studied well. Therefore, we believe
342 that this new logger will contribute to the exploration of burst movements in various
343 animals, and to the further development of animal-borne accelerometer methods.

344

345

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350

351

352 **Author contributions**

353 Y.K. developed the data logger and conceived the study. A.M. and Y.K. conducted the
354 calibration experiment. N.N., A.M., and R.K. conducted the laboratory experiment with
355 the red seabream. N.P., C.H., and Y.Y.W. conducted the field experiment on the
356 yellowtail kingfish. N.N. and Y.K. analyzed the data. N.N. and Y.K. drafted the
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358

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368

369

370 **Ethical approval**

371 Animal care and experimental procedures were approved by the Animal Care and Use
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373 ECSE15-13), in accordance with the Regulations of the Animal Care and Use
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459 **Figure captions**

460

461 **Fig. 1**

462 (a-c) Typical acceleration signals (a, x-axis; b, y-axis; c, z-axis) during the escape
463 response of red seabream. The black solid and dashed lines indicate the initiation of
464 recording and the end of the stage 1 escape response, respectively. Blue and red lines
465 indicate the instances when acceleration reached their maximum and minimum
466 acceleration values, respectively. (d) Boxplot of record initiation latency, and the
467 instances when acceleration reached maximum and minimum acceleration values. The
468 boxes, upper and lower whiskers indicates IQR, maximum value or $Q_3 + 1.5 \text{ IQR}$, and
469 minimum value or $Q_1 - 1.5 \text{ IQR}$, respectively. The red line indicates the median latency
470 of record initiation

471

472 **Fig. 2**

473 The effect of sampling frequency on measured maximum and minimum acceleration
474 values during the escape response. The vertical axes show the ratio of the downsampled
475 acceleration data to the original 500 Hz acceleration data. The solid black line and upper
476 and lower gray lines indicate median, maximum and minimum values, respectively

477

478

479 **Fig. 3**

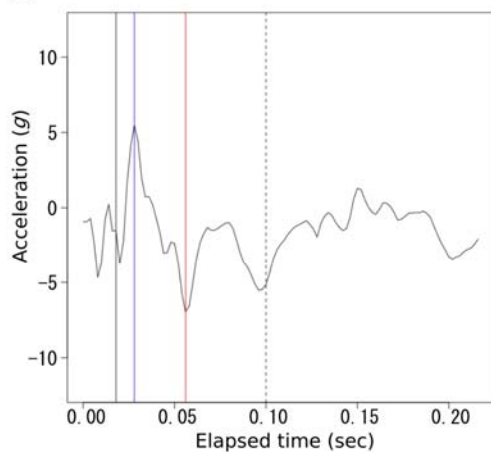
480 Activations of the Event loggers attached onto the two free-ranging yellowtail kingfish
481 (a, b). The activation events (red arrows) are shown against the maximum absolute
482 values of the 3-axis accelerations recorded by the 20 Hz reference loggers

483

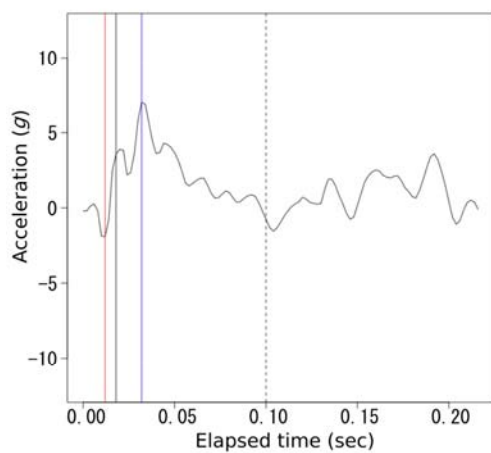
484 Fig. 1

485 (a)

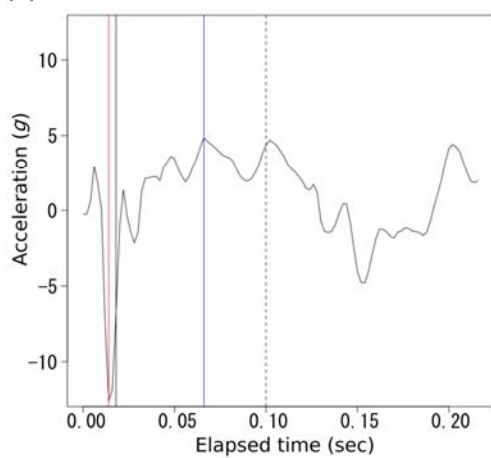
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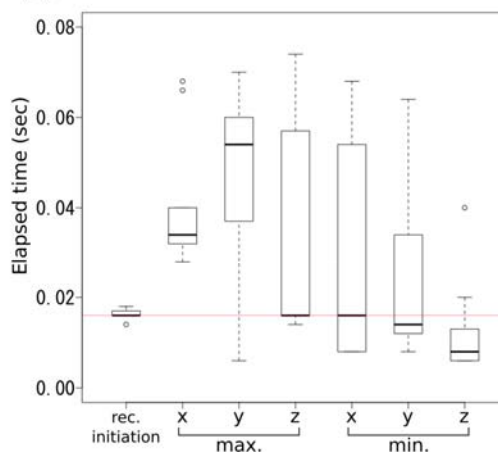
(b)



(c)

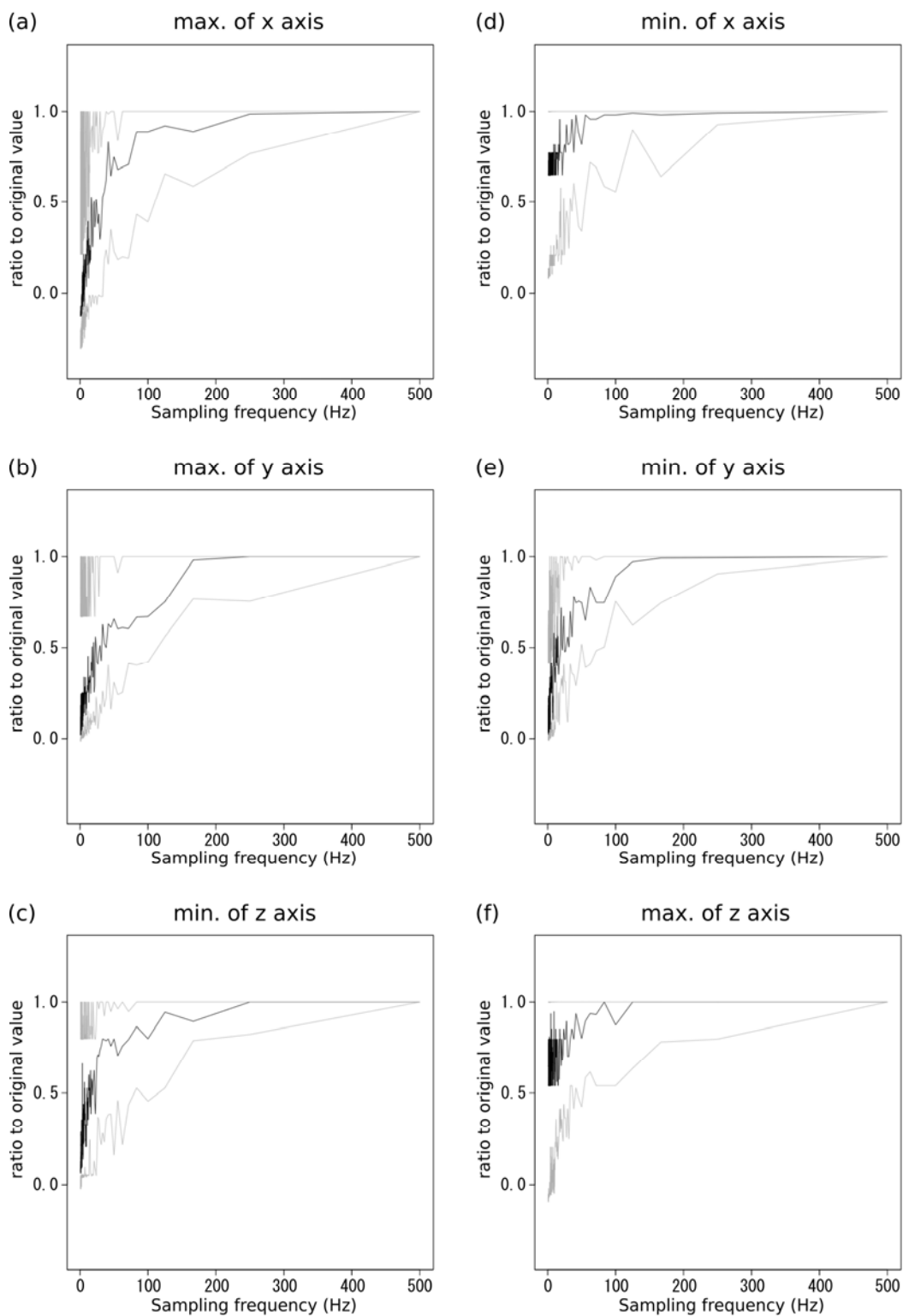


(d)

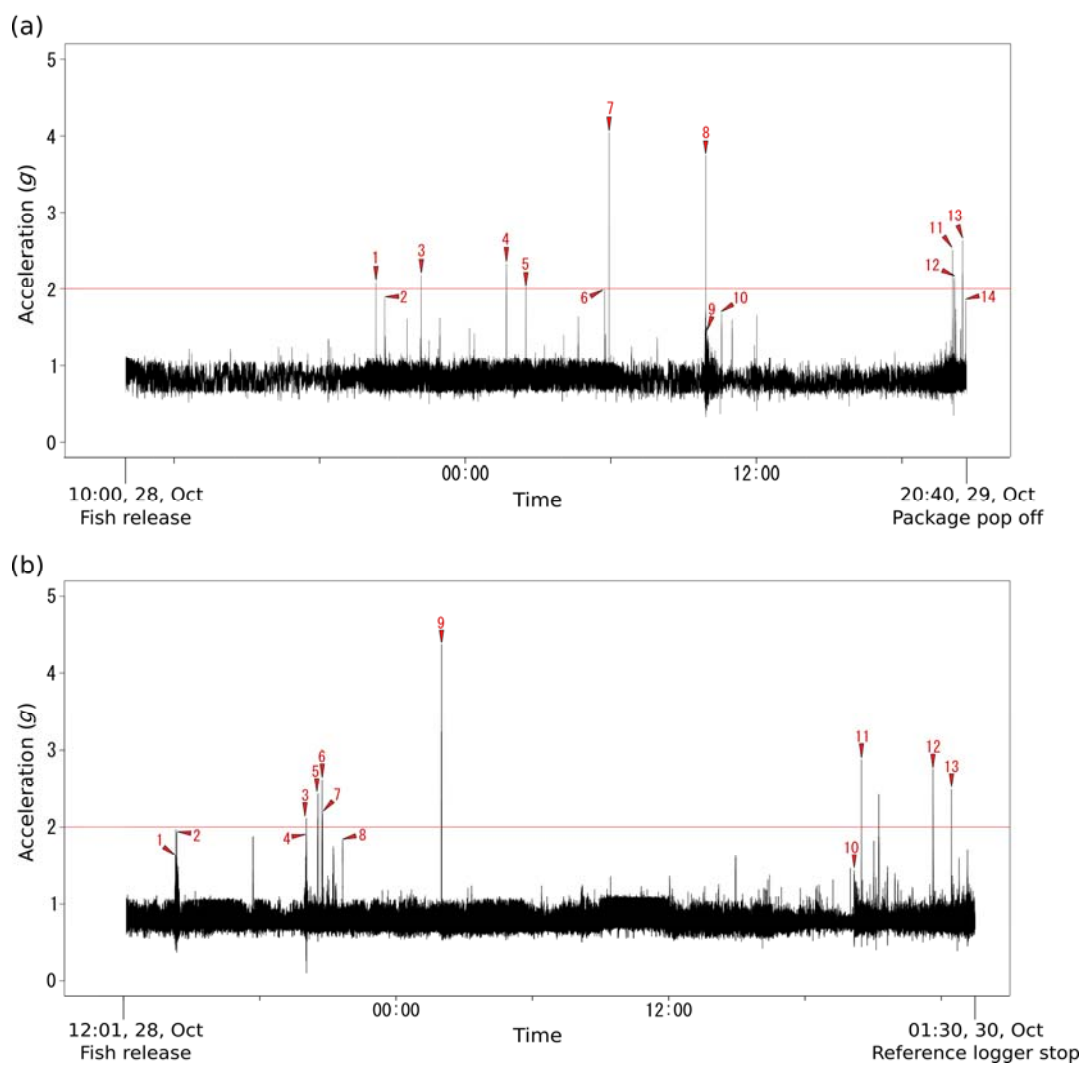


487 Fig. 2

488



489 Fig. 3



490