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1 **Citizen science and invasive alien species: predicting the detection of the**
2 **oak processionary moth *Thaumetopoea processionea* by moth recorders**
3

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29 **Summary**

30 Invasive alien species are a major cause of biodiversity change and may impact upon human well-
31 being and the economy. If new, potentially invasive, taxa arrive then it is most cost-effective to
32 respond as early in their establishment as possible. Information to support this can be gained from
33 volunteers, i.e. via citizen science. However, it is vital to develop ways of quantifying volunteer
34 recorder effort to assess its contribution to the detection of rare events, such as the arrival of invasive
35 alien species. We considered the potential to detect adult oak processionary moths (*Thaumetopoea*
36 *processionea*) by amateur naturalists recording moths at light traps. We calculated detection rates
37 from the Netherlands, where *T. processionea* is widely established, and applied these to the spatial
38 pattern of moth recording effort in the UK. The probability of recording *T. processionea* in the
39 Netherlands varied across provinces from 0.05-2.4% per species of macro-moth recorded on a list of
40 species (so equalling 1-52% for a list of 30 species). Applying these rates to the pattern of moth
41 recording in the UK: *T. processionea* could be detected (detection >0%), if it were present, in 69%
42 and 4.7% of 10km and 1km squares, respectively. However, in most squares detection probability is
43 low (<1% of 1km squares have annual detection probability of >10%). Our study provides a means to
44 objectively assess the use of citizen science as a monitoring tool in the detection of rare events, e.g.
45 the arrival of invasive alien species, occurrence of rare species and natural colonisation.

46

47

48 Key words: list length analysis, monitoring, volunteer, naturalist, citizen scientist, alien invasive
49 species

50

51 **Highlights**

- 52 • Outbreaks of *Thaumetopoea processionea* could be detected by amateur moth recorders
- 53 • We analysed moth trapping from the Netherlands and applied results to the UK
- 54 • *T. processionea* could be detected, if present, but mostly with low probability
- 55 • This citizen science is valuable for, but insufficient to guarantee, early detection
- 56 • It is important to quantify recorder effort in citizen science

57

58 **Introduction**

59 Globally, invasive alien species are one of the major threats to biodiversity, and they may also impact
60 negatively upon human well-being by affecting ecosystem services and human health (Millennium
61 Ecosystem Assessment 2005; Pejchar & Mooney 2009; Pyšek & Richardson 2010). These impacts
62 can be costly to society, but managing invasive alien species also incurs a cost, which becomes
63 increasing high as the species become established. Therefore, if a potentially-invasive alien species is
64 introduced to an area, early detection is important for effective (and cost-effective) control and
65 eradication (Hobbs & Humphries 1995; Pyšek & Richardson 2010; Blackburn *et al.* 2011). The cost
66 of detecting nascent invasions of alien species can be high (Mehta *et al.* 2007) and is an important
67 consideration when developing optimal strategies for responding to these species (Epanchin-Niell *et*
68 *al.* 2012). Thus establishing low-cost methods to provide large-scale and long-term surveillance for
69 invasive alien species is important.

70 Citizen science, that is the involvement of volunteers in the process of scientific research, including
71 making records of species' occurrences, has great potential for the detection of invasive alien species
72 because it can be an effective method for gaining reports of rare events, including new occurrences of
73 invasive alien species, at a relatively low cost (Dickinson, Zuckerberg & Bonter 2010). One approach
74 is for citizen science participants to monitor fixed plots for the presence of invasive alien species
75 (Maxwell, Lehnhoff & Rew 2009; Crall *et al.* 2011). Success depends on volunteers being effective at
76 detecting and identifying invasive alien species; something that has been tested and repeatedly found
77 to be true (Delaney *et al.* 2008; Gallo & Waitt 2011; Crall *et al.* 2011). This approach requires
78 substantial resources for coordination and volunteer recruitment but, providing all the plot data are
79 submitted, it generates information on the absence of invasive alien species as well as their presence
80 at these locations. However, systematically monitoring pre-defined plots does not address the need for
81 early detection of invasive alien species at large spatial or temporal extents, such as is necessary for
82 those species that are predicted to arrive, but precisely where and when is unknown (e.g. Roy *et al.*
83 2014).

84 An alternative citizen science approach for detecting potential invasive alien species is the
85 opportunistic reporting of observations by the general public. While the probability of arrival of
86 invasive alien species can be modelled (Ibáñez *et al.* 2009), actual arrivals are rare stochastic events.
87 So, while the likelihood of a particular invasive alien species occurring in a particular place at a
88 particular time is almost negligible, when considering a large area over a long-enough time period the
89 overall probability of arrival is much greater. Engaging with the general public and providing tools for
90 data submission is therefore a potentially cost-efficient method for early detection across large spatio-
91 temporal scales (Roy *et al.* 2015).

92 Currently, there are several examples of citizen science in which anyone can record invasive alien
93 species, e.g. Recording Invasive Species Counts (Roy *et al.* 2012), Invaders of Texas (Gallo & Waitt
94 2011) or EDDMapS (Bargeron & Moorhead 2007). These types of projects have the potential to
95 provide good spatial coverage through promotion via the media. However, one of the serious
96 limitations is that typically the data gathered are ‘presence only’ records: an absence of records
97 provides no information on the absence of the species (i.e. the situation with no observers is
98 indistinguishable from the situation with many observers and the species absent). In order to draw
99 inference from the absence of records (e.g. see Isaac *et al.* 2014) it would be extremely valuable to
100 have an assessment of recorder effort, but this is very difficult to quantify. An alternative approach is
101 to rely upon natural history enthusiasts who are already making and submitting records (an activity
102 that falls within the definition of citizen science; Pocock *et al.* 2015), to report sightings of new
103 invasive alien species belonging to their taxon of interest.

104 As a case study, we consider one approach for the detection of the oak processionary moth
105 *Thaumetopoea processionea* (Lepidoptera: Notodontidae) in the UK. *T. processionea* is of current
106 concern to policy makers in the UK because it has become established in west London, following its
107 recent spread in Belgium and the Netherlands (Groenen & Meurisse 2012). *T. processionea* can
108 impact upon human health because the larvae shed urticating setae that can cause allergic reactions
109 such as urticaria, conjunctivitis and respiratory difficulties (Gottschling & Meyer 2006; Fenk, Vogel &
110 Horvath 2007; Mindlin *et al.* 2012). In some parts of the species’ range and at high population
111 densities it can be a defoliator of oak trees (Wagenhoff & Veit 2011) and so potentially could impact
112 upon oak health and biodiversity as well (although this has not occurred in the UK to date).

113 *T. processionea* was accidentally introduced to the UK on imported oak trees (*Quercus* sp.); it was
114 first recorded in west London in 2006 and had expanded its range by about 10km radius by 2011
115 despite control measures, probably mostly by natural dispersal, although human-mediated dispersal is
116 also possible (Townsend 2013). Its gradual spread from its current range is currently monitored by
117 professionals and trained volunteers who undertake visual surveys of the silk nests built by the
118 communal larvae and pheromone trapping for adult male moths (Mindlin *et al.* 2012; Williams *et al.*
119 2013). However, this approach is not suitable for detecting occurrences of the species away from the
120 slowly-expanding distribution in west London (e.g. new introductions to the UK or human-mediated
121 dispersal within the UK) because any such occurrences are unpredictable, requiring the long-term
122 surveillance of very large geographical areas with extremely high financial cost if undertaken by paid
123 surveyors. However, other approaches such as pheromone traps have proved useful to assess spread of
124 similar species (Sharov *et al.* 2002) and could be run by volunteers. In addition, observing larval nests
125 in low density populations is unreliable because they typically occur in the oak canopy and are often
126 hidden by foliage (Townsend 2013), although such biases in detection can be taken into account in
127 data from monitoring schemes (Fitzpatrick *et al.* 2009).

128 In the UK, the Netherlands and elsewhere many thousands of people record moths as a hobby,
129 submitting records to national databases. The use of light traps is an especially popular form of moth
130 recording, partly due to its convenience, e.g. traps can be left running overnight in gardens and
131 catches recorded the following morning (Fry & Waring 2001). These enthusiasts usually record lists
132 of species captured, in particular all the macro-moths captured, similar to the ‘checklist’ approach for
133 opportunistic recording of birds (Sullivan *et al.* 2014). This allows changes in moth prevalence over
134 time to be quantified (Groenendijk & Ellis 2010; Fox *et al.* 2014), but also means that the absence of a
135 species from a list can be considered a non-detection (Isaac *et al.* 2014), i.e. the non-detections can be
136 distinguished from a lack of recording effort. This is not the case for most mass participation citizen
137 science projects where presence-only data are collected and recording effort (including recording
138 absences) is not known. Interpretation of such data becomes increasingly difficult as the species of
139 interest becomes less frequently recorded and often requires recording effort to be inferred, by the
140 recording of related species (Snäll *et al.* 2011; Isaac *et al.* 2014).

141 Our aim in the current project was to use data from a region where *T. processionea* is established (the
142 Netherlands) to calculate the probability that moth recorders detect *T. processionea* when it is present,
143 and then to apply these detection probabilities to the current pattern of citizen science moth recording
144 in the UK. From this we could estimate the probability that moth recorders would provide early
145 detection of *T. processionea* across the UK.

146 **Methods**

147 The Noctua database holds data from volunteer moth recorders in The Netherlands and currently
148 holds 4.5 million records (Groenendijk & Ellis 2010). We extracted data from the Noctua database on
149 moth records during the flight period of *T. processionea* in 2002-2013. *T. processionea* was
150 established in the Netherlands over this period. The flight period was 25 July- 30 August, which was
151 defined as the range of dates where the number of records of *T. processionea* was at least 10% of the
152 maximum number of records per day for the years 2002-2010 and 2012-2013 (the year 2011 was
153 removed due to an apparent artefact in the data; Fig S1). The records in the Noctua database comprise
154 species identity, grid reference, date and recorder name. We aggregated the moth records by ‘species
155 lists’ (Szabo *et al.* 2010), where a species list comprises the moths recorded during one night of moth
156 trapping; specifically we defined a ‘species list’ as a unique combination of 1km grid square and date.
157 We did not use recorder name to distinguish between samples because names are not unique and can
158 be recorded in multiple ways within the database (e.g. with or without initials and first names) and
159 multiple recorders could have submitted the same record (e.g. when they all took part in a group moth
160 trapping event). Considering the unique combination of 1km square and date may occasionally lead to
161 aggregation of separate species lists (where they occurred in the same 1km grid square on the same
162 night), but our experience suggests that this occurs only rarely at the 1km resolution.

163 We then calculated the probability of recording *T. processionea* (OPM) while taking account of the
164 list length (i.e. the average ‘per-species recording probability’: \bar{S}_{OPM}). There is spatial variation in the
165 prevalence of *T. processionea* across the Netherlands, so throughout we undertook analyses separately
166 in each province.

167 To calculate the probability that *T. processionea* had been recorded in a species list we firstly
168 calculated the total probability that *T. processionea* was recorded on a list of length L ($P_{OPM,L}$; eqn 1).

$$169 \quad P_{OPM,L} = N_{OPM,L}/N_{total,L} \quad [\text{eqn 1}]$$

170 where, for a given list of length L , $N_{total,L}$ is the total number of lists and $N_{OPM,L}$ is the number of lists
171 in which *T. processionea* was present.

172 Following Szabo *et al.* (2010), we expected that the probability of detecting *T. processionea* ($P_{OPM,L}$)
173 on a list would increase with increasing list length (L). This is because list length gives an indication
174 of recording effort, assuming that all recorders record every macro-moth species they identify, which
175 is typical behaviour among moth recorders in north-western Europe. It could be possible to test this
176 assumption quantitatively in the future because biased recording of some species would result in them
177 being more likely to be recorded on shorter lists. In the case of light traps running overnight, ‘effort’
178 is a function of factors including the effectiveness of the moth trap, duration of trapping, number of
179 traps used, weather conditions, moon phase and local habitat. Calculating the per-species probability
180 of recording *T. processionea* ($S_{OPM,L}$) for each category of list length L in each province takes the list
181 length into account (eqn 2).

$$182 \quad S_{OPM,L} = 1 - \exp(\ln(1 - P_{OPM,L}) / L) \quad [\text{eqn 2}]$$

183 Therefore, $S_{OPM,L}$ was calculated for each value of the list length L . We calculated the average $S_{OPM,L}$
184 (eqn 3) across a set of these values of L (i.e. treating each list length category, not the lists themselves,
185 as the data) which met the criteria that: (i) the value of the list length was at least six (i.e. $L > 5$), (ii)
186 there were at least six lists of that list length (i.e. $N_{OPM,L} > 5$ for each value of L), and (iii) there were
187 some/all lists of that list length in which *T. processionea* was absent (i.e. $P_{OPM,L} < 1$). We excluded
188 these three cases because (i, ii) observation of the results (Fig. S1) suggested that estimates of $S_{OPM,L}$
189 tended to be lower than expected when the list lengths were very short or few lists were included in
190 the category of length L , and (iii) in these cases $S_{OPM,L}$ was constrained to be one and appeared to be
191 biased high. From \bar{S}_{OPM} for each province, we could back-calculate the estimated probability of
192 recording *T. processionea* for a list of length L ($\hat{P}_{OPM,L}$) as one minus the probability of not detecting
193 *T. processionea* (eqn 4).

$$194 \quad \bar{S}_{OPM} = \frac{1}{M} \sum_{i=1}^M S_{OPM,L} \quad [\text{eqn 3}]$$

195 where M is the subset of values of the list length as described in the text

196
$$\hat{P}_{OPM,L} = 1 - (1 - \bar{S}_{OPM})^L \quad [\text{eqn 4}]$$

197 We then applied the values of \bar{S}_{OPM} obtained from data from the Netherlands to the pattern of moth
 198 recording across the UK. Specifically, we calculated estimated detection rate in the UK (\hat{D} : eqn 5), by
 199 combining (1) the probability of recording *T. processionea* per recording event (\bar{S}_{OPM}) for
 200 Netherlands providences, with (2) the recording effort in the UK (i.e. the list length and frequency of
 201 recording). We extracted information on all recording events between 25 July and 30 August from the
 202 UK National Moth Recording Scheme database (Fox *et al.* 2010), which currently holds over 20
 203 million records. We therefore assumed that the flight period of *T. processionea* was the same in the
 204 UK as the Netherlands. There can be a lag in the UK from record submission and verification by
 205 county recorders to acceptance into the database, so to minimise this effect we considered the records
 206 for the ten-year period 2000-2009. As for the Netherlands dataset, a recording event was defined as
 207 the list of species recorded in a unique combination of 1km grid square and date. Therefore, for any
 208 region (e.g. a 1km square) and any year, we knew the length (L) of each list ($n = 1$ to the total N lists
 209 in that region) and so could calculate, across all lists and for a given value of \bar{S}_{OPM} , the estimated
 210 probability of detecting *T. processionea* (\hat{D} ; eqn 5). Note that \hat{D} is scale-free, so it can be calculated at
 211 any extent. However, it does assume that the selected value of \bar{S}_{OPM} is appropriate over the whole of
 212 each region (e.g. a whole 1km or 10km square). For the results presented here we calculated the
 213 average \hat{D} across the years 2000-2009.

214
$$\hat{D} = 1 - \prod_{n=1}^N [(1 - \bar{S}_{OPM})^{Ln}] \quad [\text{eqn 5}]$$

215 **Results**

216 **The probability of recording *T. processionea* in the Netherlands**

217 Our dataset for moth recording in the Netherlands between 25 July and 30 August in 2002-2013
 218 comprised 53 781 lists (i.e. unique combinations of 1km grid square and date) of 417 614 individual
 219 species records. *T. processionea* was recorded 2 640 times (i.e. it comprised 0.6% of species records
 220 and occurred on 4.9% of lists).

221 The probability of recording *T. processionea* per recording event ($P_{OPM,L}$) increased with increasing
 222 list length (L), as we expected (Fig. 1 a-l). The average per-species detection probability (\bar{S}_{OPM}),
 223 calculated from a subset of all the list lengths (Fig. 1 and S2) was back-calculated to the observed list
 224 length ($\hat{P}_{OPM,L}$) and showed a good fit to the observed data (Fig. 1).

225 We found that provinces varied in the average per-species probability of recording *T. processionea*
 226 (Fig. 1 m and n). The two provinces in the south-east of the Netherlands, where *T. processionea* had
 227 been established longest, had per-species detection probabilities of 2.1-2.4% (i.e. this was the chance
 228 that a new species on a list at a recording event would be *T. processionea*; Fig. 1k-l). This equates to

229 47-52% chance of recording *T. processionea* when a recording event obtained a list of 30 species. The
230 four provinces with medium detection rates had an average per-species probability of recording of
231 about 1.4% (Fig. 1 g-j), equating to a 34% chance of recording *T. processionea* for a list of 30 species.
232 Finally those provinces with the lowest detection rate, the per-species detection rate varied from 0.05
233 to 0.4% (Fig. 1 a-f), so for a list of 30 species there was a 1-11% chance of detecting *T. processionea*.

234 **The probability of recording *T. processionea*, if it was present, in the UK**

235 The number of species lists recorded in the UK during the flight period of *T. processionea* (25 July-30
236 August, i.e. assumed to be the same as in the Netherlands) between 2000 and 2009 was 136 344
237 (range per year: 9 753-15 369) with a total of 1 618 661 individual species records. *T. processionea*
238 was not recorded on any list in this dataset, even though it was present in western London from 2006
239 and had been recorded at various sites on the south coast of England as a presumed immigrant from
240 continental Europe. There were lists from 2 119 (69%) of the 3 055 10km squares in the UK during
241 25 Jul-30 August 2000-2009 (Fig. 2) and 12 190 (4.7%) of 256 663 1km grid squares in the UK, i.e.
242 for each 10km square, on average only five of the 100 1km squares had records. Squares with lists
243 were distributed across the UK although parts of Scotland and Northern Ireland were relatively
244 sparsely covered (Fig. 2).

245 Applying the per-species recording probabilities from the Netherlands to the UK showed the coverage
246 of squares at different detection thresholds (Table 1; Fig. 1). There was a greater than 0% chance of
247 moth-recorders detecting *T. processionea*, if it had been present, in 69% of 10km squares, but only
248 4.7% of 1km squares, in the UK (Table 1). However, considering the situation with higher detection
249 thresholds, the overall coverage is lower and patchy (Table 1; Fig. 1); when considering the threshold
250 of $\hat{D} > 50\%$ (i.e. chances are *T. processionea* would be recorded, if it was present, in any year with the
251 pattern of recording effort during 2000-2009) then only 5.5% of 10km squares and <0.1% of 1km
252 squares meet this criteria (Table 1; Fig. 2).

253 However, for the outbreaks in their earliest stages, occurrence will be at a much smaller spatial extent
254 than the 10km square. The range (area of the minimum convex polygon) of *T. processionea* in west
255 London in 2009 was just 58km² (Fig. 3). Finer resolution analysis of the data within a 50km square
256 covering west London where *T. processionea* is established, shows how recording effort is
257 distributed. At the resolution of 10km squares, most squares have a 10-50% annual probability of
258 detecting *T. processionea*. However, actual recording occurs at a much finer resolution (i.e. within
259 1km squares, by the definition of a recording event used in the current study). Within the 50km
260 square, most of the 1km squares have a 0% probability of detecting *T. processionea* showing the
261 importance of considering spatial resolution of recording effort relative to invasive species range size.

262 **Discussion**

263 Currently citizen science is promoted as a potential method for conducting cost-effective
264 environmental monitoring, including the early detection of invasive alien species and disease (Tree
265 Health and Plant Biosecurity Expert Taskforce 2012; Dickinson *et al.* 2012; Roy *et al.* 2015).
266 ‘Opportunistic’ recording can produce data which is suitable to monitor many species when recording
267 is via a ‘checklist’ approach or when non-detections can be inferred (Snäll *et al.* 2011; Sullivan *et al.*
268 2014; Isaac *et al.* 2014), but is less useful as the focal species becomes less frequently recorded.
269 Interpreting the results of projects in which people submit records of potentially invasive alien species
270 (i.e. presence-only data from mass participation citizen science) is difficult because recorder effort
271 cannot usually be quantified. It is important to distinguish lack of records due to the species being
272 absent from a lack of recorders. In this study, by considering volunteers who record the target species
273 as a by-product of general recording, we were able to estimate the probability that volunteers
274 recording macro-moths would detect the moth oak processionary, *T. processionea*.

275 From our findings in this study we draw two conclusions. Firstly, across much of the UK there is a
276 greater than zero probability that moth recorders will detect *T. processionea* if it is present; therefore
277 this form of ‘citizen science’ could be useful for its early detection. Secondly, the actual probability of
278 detecting *T. processionea* is low and patchy across the UK, especially at fine spatial resolutions (i.e.
279 within 1km grid squares), so this form of monitoring is unlikely to be sufficient in providing early
280 detection of *T. processionea*. The environment in the Netherlands (where we parameterised the
281 model) is not a perfect match to the UK (where we applied the model), but we are confident that it is
282 similar enough for our results to provide a good indication of the likely detection of *T. processionea*
283 by moth recorders in the UK. Given the way naturalists record moths at light traps, it is unlikely that
284 this distinctive species would be missed or mis-identified, if present, but lack of awareness could
285 contribute to mis-identifications leading to non-detections for more cryptic or less distinctive species.
286 Overall, maps of quantified recording effort (e.g. Fig. 2 for the amateur naturalists considered in this
287 study) could be combined with maps of hazard, e.g. *T. processionea* arrival or spread (Cowley,
288 Johnson & Pocock 2015), if such maps were available, to optimise the targeting of additional
289 recording effort, e.g. professional monitoring or targeted advertising.

290 Volunteers who record moths do so for a range of motivations, including their own enjoyment,
291 connection with nature and wanting to contribute to scientific knowledge (e.g. Fox *et al.* 2014). The
292 early detection of invasive alien species is a by-product of this recording rather than an intended aim.
293 Other people may have different motivations for taking part in the search for and reporting of *T.*
294 *processionea*, e.g. arboriculturists, land managers, local council staff and householders concerned
295 about human health impacts. These will all contribute to reporting, so the overall situation for
296 effective early detection is not as pessimistic as it might seem from our analysis. However, as we have
297 stressed, this additional recording effort cannot be easily quantified, meaning that it is not possible to

298 predict detection probability, and so it is difficult to effectively manage resources to strategically
299 optimize detection (Hauser & McCarthy 2009).

300 **Asymmetry of information and data flow**

301 If *T. processionea* is not detected then, as we have discussed, it is important to assess the probability
302 that it was present but not detected. However, the converse is very different. If *T. processionea* is
303 detected, then it is important for decision makers that the information is available as quickly as
304 possible in order to determine appropriate action. Currently in Great Britain (GB) there is an alert
305 system for early detection of invasive alien species (Roy *et al.* 2012, 2015), which has an organized
306 structure to support rapid data flow (Fig. 4). There are three potential bottlenecks to data flow. The
307 first is the submission of a record by the observer. Websites and especially smartphone apps facilitate
308 the reporting of potential target species (August *et al.* 2015), but rely on people being aware of and
309 utilising them: communication is important. The second potential bottleneck is the verification of
310 records by experts (volunteers or professionals). A successful public awareness campaign can result in
311 a large number of misidentified records and, even if supporting information (e.g. photographs) are
312 submitted, resources are still needed to support this (Roy *et al.* 2015). The third potential bottleneck is
313 the onward flow of data to those who are able to mount an appropriate response. Inter-operable data
314 systems are an ambition (Graham *et al.* 2008) but the proliferation of individual citizen science
315 projects can put efficient data flow under risk, and so it is incumbent upon project organizers to
316 consider this as utmost importance.

317 **Using citizen science as a tool for detection of rare events**

318 In the current study we have specifically considered the effectiveness of volunteers to provide
319 information on the presence and absence of a target species, in this case *T. processionea*, which can
320 be compared to other methods for the detection of rare events (Table 2). Typically, active surveillance
321 (which could be by professionals or volunteers) is considered when seeking to model the optimal
322 monitoring strategies for early detection of rare events (Maxwell, Lehnhoff & Rew 2009). However,
323 passive surveillance by the general public (or a trained subset thereof) has the potential to permit the
324 long-term, large-scale surveillance of rare events at relatively little cost (Pocock *et al.* 2013); the
325 public are potentially a resource “ready to act as the need arises” (Cooper *et al.* 2007). It is most likely
326 to be successful when the rare events are very noticeable or directly impact people, and is dependent
327 upon having a high public profile, e.g. extensive media coverage. This approach has been deemed
328 successful in the past (Aitkenhead 1981; Hesterberg *et al.* 2009) even though it is not possible to
329 directly assess the recorder effort. Alternatively, people can become involved with focussed
330 monitoring, e.g. by deploying and checking pheromone traps (Sharov *et al.* 2002) although, as with
331 other approaches, detection probability still needs to be considered (Fitzpatrick *et al.* 2009) and the
332 issue of people not reporting absences remains problematic. Also, as citizen science continues to

333 develop, further research on participants' motivations (Rotman *et al.* 2012; Nov, Arazy & Anderson
334 2014) will enhance our ability to effectively use citizen science as a tool for the detection of rare
335 events (Pocock *et al.* 2013).

336 **Conclusion**

337 There is great enthusiasm for citizen science and its role in environmental monitoring. Citizen science
338 clearly does have a role to play in the early detection of invasive alien species, and can also be applied
339 to other rare events such as occurrence of wildlife disease (Kulasekera *et al.* 2000; Hesterberg *et al.*
340 2009), unusual weather (<http://www.cocorahs.org>) and landslips
341 (<https://britishgeologicalsurvey.crowdmap.com/>). When assessing results from such projects it is
342 important to quantify the recorder effort in order to distinguish the absence of records (because there
343 are no recorders) from the absence of the event (even though potential recorders were present).
344 However with presence-only data this is often hard to achieve. The approach in this study was to
345 quantify recording effort by moth recorders and use this to estimate the probability of detecting an
346 invasive alien moth, *T. processionea*, if it was present. Although moth recorders are just one subset of
347 the potential recorders, it shows that there is a chance of recording *T. processionea* across much of the
348 UK, but that the chance is often quite small, making records from moth recorders a valuable, but not
349 sufficiently effective, component of an early detection network for *T. processionea*. This result is
350 relevant to other 'rare events' including the detection of rare or highly threatened resident species and
351 newly-colonising species. Citizen science in all its forms is bound to play an increasing role in
352 detection of rare events but it requires thoughtful enthusiasm rather than hype to ensure that it
353 provides many opportunities for excellent cost-effective science.

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365 Natural Environment Research Council. The Non-Native Species Secretariat provided invaluable
366 support in the development of the Alert system.

367

368 **Appendix**

369 Figure S1. The phenology of *Thaumetopoea processionea* in the Netherlands, based on the number of
370 records in the Noctua database.

371 **Figure S2.** The per-species recording probability for *Thaumetopoea processionea* (S_{OPM}) in each
372 province in the Netherlands.

373

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492

493

494 **Table 1.** The percentage of total 10km and 1km grid squares in the UK which meet the criteria for the
 495 annual probability of detecting *T. processionea* if it was present (\hat{D}), based on the per species
 496 probability of recording *T. processionea* (\bar{S}) in the Netherlands (2002-2013) and the pattern of moth-
 497 recording in the UK (2000-2009). The different values of \bar{S} are taken from the different providences
 498 in the Netherlands and are assumed to be a function of the local density of *T. processionea*, with very
 499 low to low values considered to be most relevant to situations where *T. processionea* is in the early
 500 stages of establishment

Per-species probability of recording (\bar{S})	Percentage of 10km grid squares				Percentage of 1km grid squares			
	Very low (0.05%)	Low (0.39%)	Medium (1.4%)	High (2.4%)	Very low (0.05%)	Low (0.39%)	Medium (1.4%)	High (2.4%)
Threshold for predicted detection probability (\hat{D})								
>0%	69.4	69.4	69.4	69.4	4.7	4.7	4.7	4.7
>1%	30.0	51.1	57.5	59.7	0.5	1.8	2.5	2.7
>10%	6.5	24.9	36.8	42.3	0.1	0.3	0.7	0.9
>50%	0.2	5.5	12.4	15.4	<0.1	<0.1	0.1	0.1

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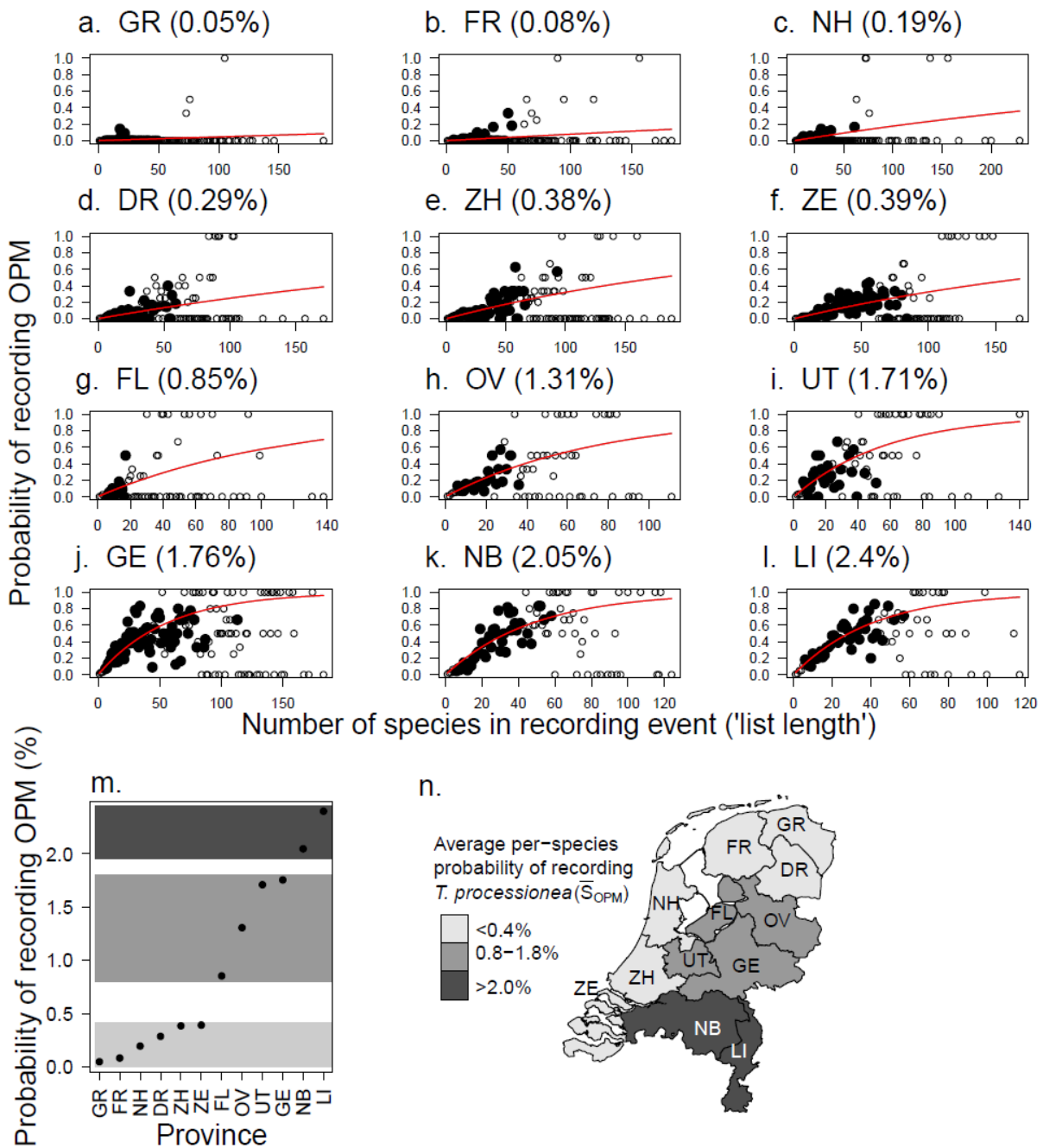
504 **Table 2.** A framework for considering the role of citizen science in the detection of rare events, such
 505 as invasive or rare species

Type of recording	Opportunistic surveillance (presence only records of target species)	Opportunistic surveillance (as a byproduct of recording other events, e.g. other species occurrences)	Systematic surveillance (monitoring by volunteers)	Active surveillance (by professionals)
Participants	General public = mass participation citizen science	Volunteers already (recording the other events)	Participants undertaking regular monitoring at known locations and known times	Contracted surveyors; they may be actively searching an area or undertaking regular monitoring at fixed sites
Recording effort	Presence-only records, so recording effort is very difficult to assess	Can be assessed by current recording of species that are not the intended target	Protocols mean that efforts can be prescribed and known	Surveyors are under contract so (in theory) their effort can be quantified and managed
Opportunities	The potential for large-scale long-term monitoring at low cost	It is supported by the enthusiasm and motivation of those already engaged in recording other events	Volunteers can be as accurate as professionals (and this can be tested) and provide cost-efficient long-term monitoring	Surveyors are under contract so they are instructed where to survey
Challenges	Sustaining interest; Regular promotion; Feedback essential but time-consuming Responding to mis-identifications; recording effort is difficult to quantify	Promoting rapid submission of records of target events; ensuring that records are dealt with efficiently and passed on to stakeholders	Requires resources to recruit and retain participants; unlikely to detect first occurrence of a rare event unless the location of such events are predictable and locations selected to match	Incurs a direct (often large) on-going cost to employ people

506

507 **Figure 1.** The probability of recording *T. processionea* depends on the number of species per
 508 recording event and varies by the province in the Netherlands. In a-l the circles show the proportion of
 509 recording events of each list length in which *T. processionea* was recorded. The line shows the
 510 estimate that was back-calculated from the average per-species recording probability (given in the title
 511 of each graph along with the two-letter code for the province name) calculated as the average from a
 512 subset of the data (shown as the points that are filled (see text for details). For completeness the
 513 remaining data not used in the calculation are showed as open circles). Provinces are ordered by
 514 increasing per-species probability of recording *T. processionea*. The average per-species recording
 515 probability in the provinces occurs in three bands (m), which are distributed as shown in (n).

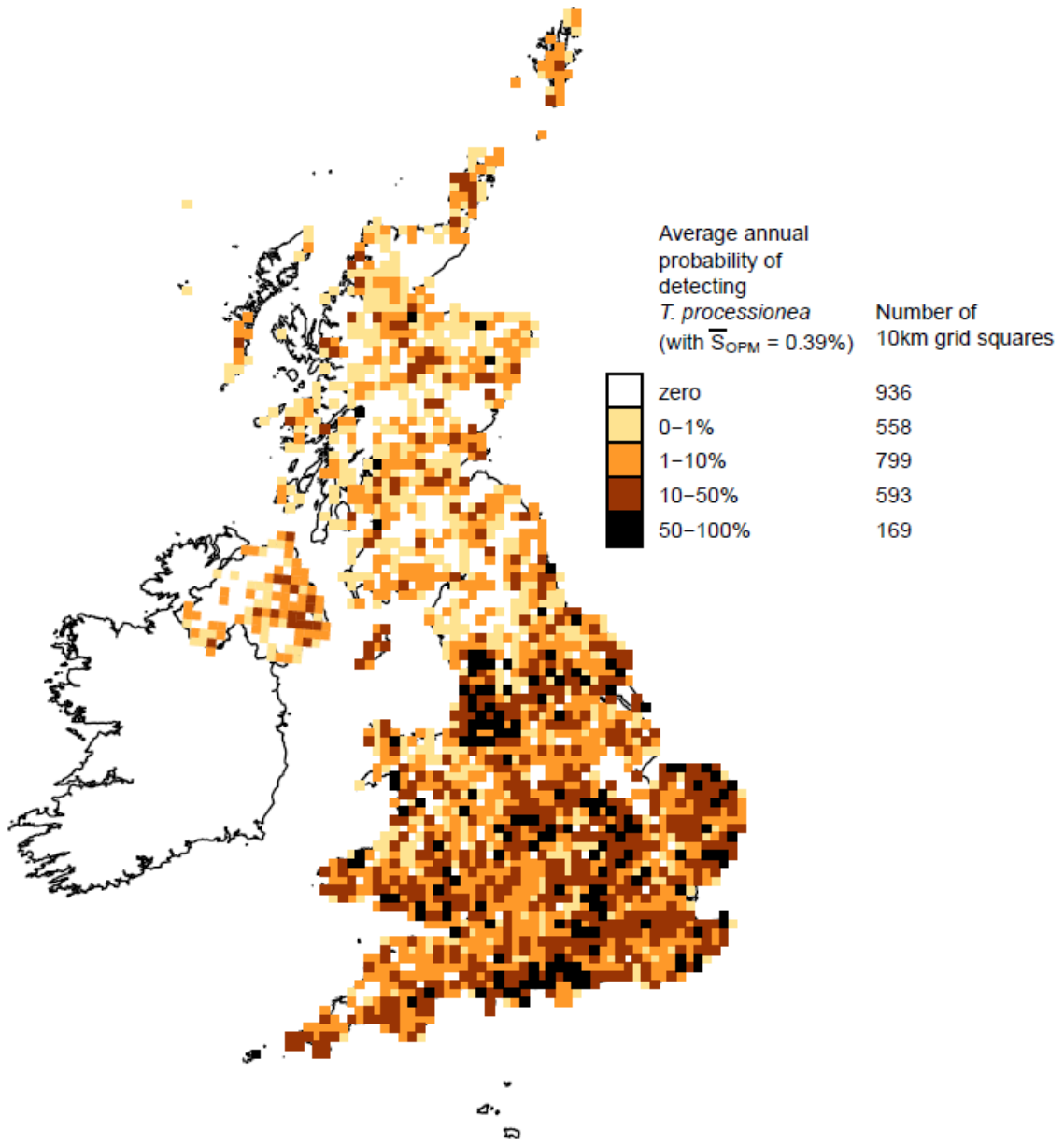
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518 **Fig. 2.** The average annual probability of detecting *T. processionea* (\hat{D}), if it were present, in 10km
519 grid squares in the UK based on the observed recording effort during 25 July-30 August in 2000-
520 2009. The results are shown when considering a low per-species probability of recording *T.*
521 *processionea* ($\bar{S}_{OPM}=0.0039$), based on modelling from the Netherlands (Fig. 1).

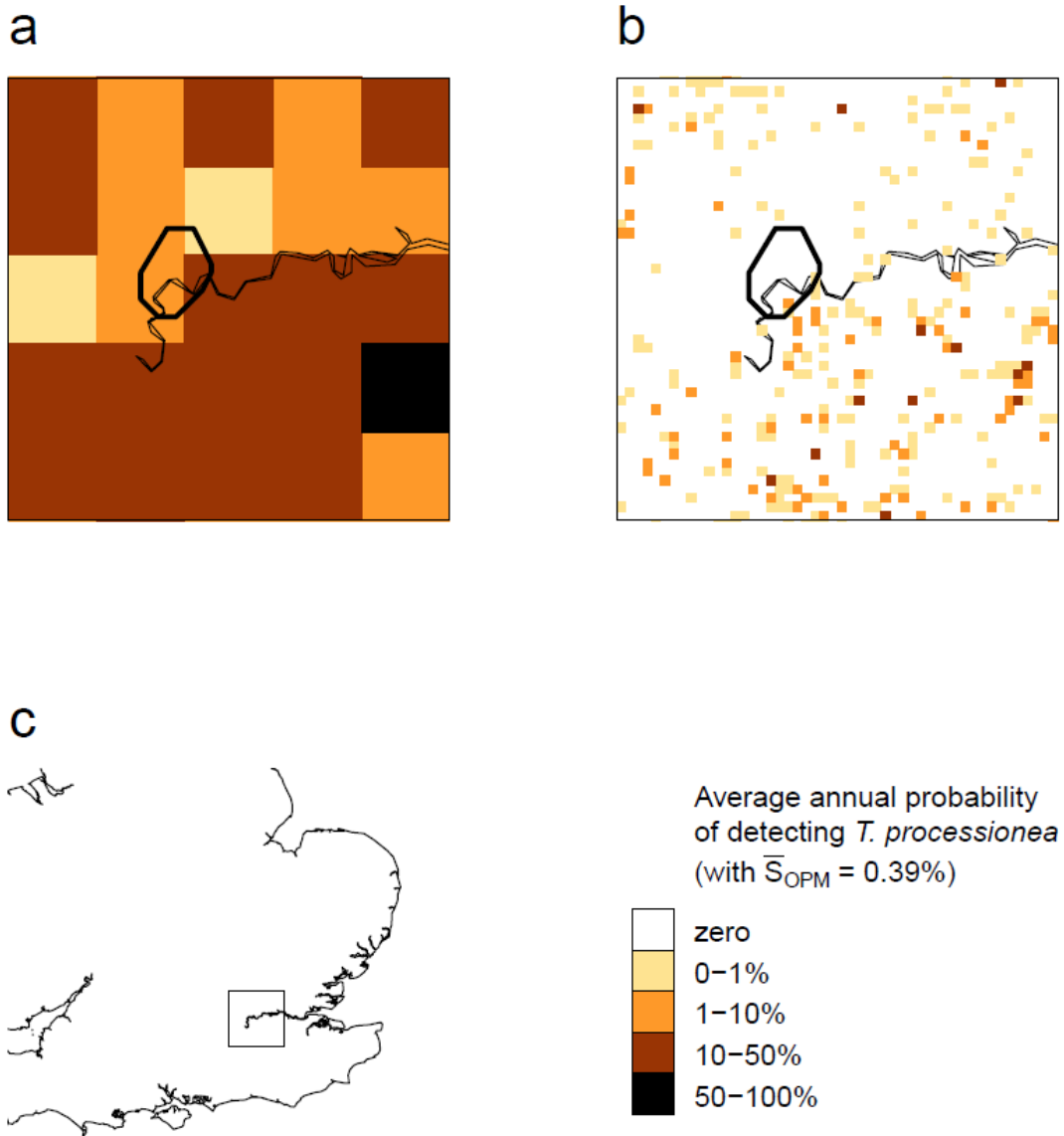
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525 **Figure 3.** The probability of detecting *T. processionea*, if it was present, in (a) 10km and (b) 1km grid
 526 squares in a 50km square containing the current range of *T. processionea* in west London (thick black
 527 outline is the minimum convex polygon of the range of *T. processionea* in 2009) based on the average
 528 recording effort by moth recorders during 25 July-30 August in 2000-2009 and a low probability of
 529 recording *T. processionea* in the Netherlands (Fig. 1). (c) The box indicates the area magnified in a
 530 and b.



531

532

533 **Figure 4.** Summary of the Great Britain (GB) Alert system for early detection of invasive alien
 534 species. (1) After a suspected observation is submitted via a website, smartphone app or email, (2) an
 535 automatic alert allows a data checker to (3) initially review the record and (4) update the database if it
 536 is incorrect. Otherwise, suspect records are (5) submitted for rapid verification by a species expert
 537 and, if verified as correct, (6) stakeholders are alerted to take appropriate action.

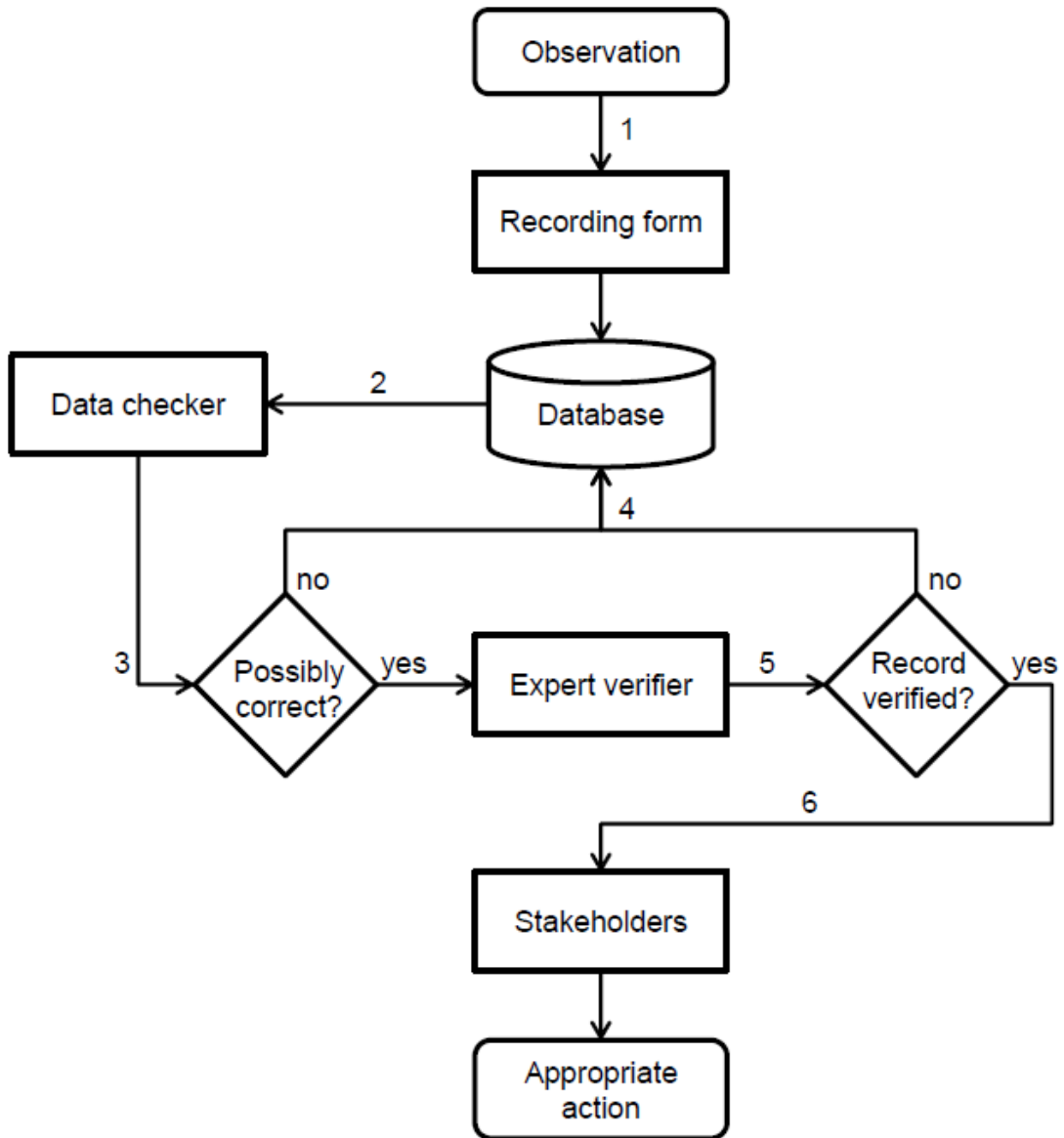


Fig. 4.

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