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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 58km<sup>2</sup> (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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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

374 **References**

- 375 Aitkenhead, P. (1981) Colorado beetle - recent work in preventing its establishment in Britain. *EPPO Bulletin*,  
376 **11**, 225–234.
- 377 August, T., Harvey, M., Lightfoot, P., Kilbey, D., Papadopoulos, T. & Jepson, P. (2015) Emerging technologies  
378 for biological recording. *Biological Journal of the Linnean Society*, **115**, 731–749.
- 379 Barger, C.T. & Moorhead, D.J. (2007) EDDMapS: Early detection and distribution mapping system for the  
380 South-east Exotic Pest Plant Council. *Wildland Weeds*, **Fall 2007**, 4–8.
- 381 Blackburn, T.M., Pyšek, P., Bacher, S., Carlton, J.T., Duncan, R.P., Jarošík, V., Wilson, J.R.U. & Richardson,  
382 D.M. (2011) A proposed unified framework for biological invasions. *Trends in Ecology & Evolution*, **26**,  
383 333–9.
- 384 Cooper, C.B., Dickinson, J.L., Phillips, T. & Bonney, R. (2007) Citizen science as a tool for conservation in  
385 residential ecosystems. *Ecology and Society*, **12**, 11.
- 386 Cowley, D.J., Johnson, O. & Pocock, M.J.O. (2015) Using electric network theory to model the spread of oak  
387 processionary moth, *Thaumetopoea processionea*, in urban woodland patches. *Landscape Ecology*, **30**,  
388 905–918.
- 389 Crall, A.W., Newman, G.J., Stohlgren, T.J., Holfelder, K. a., Graham, J. & Waller, D.M. (2011) Assessing  
390 citizen science data quality: an invasive species case study. *Conservation Letters*, **4**, 433–442.
- 391 Delaney, D.G., Sperling, C.D., Adams, C.S. & Leung, B. (2008) Marine invasive species: validation of citizen  
392 science and implications for national monitoring networks. *Biological Invasions*, **10**, 117–128.
- 393 Dickinson, J.L., Shirk, J., Bonter, D., Bonney, R., L Crain, R., Martin, J., Phillips, T. & Purcell, K. (2012) The  
394 current state of citizen science as a tool for ecological research and public engagement. *Frontiers in*  
395 *Ecology and the Environment*, **10**, 291–297.
- 396 Dickinson, J.L., Zuckerberg, B. & Bonter, D.N. (2010) Citizen science as an ecological research tool: challenges  
397 and benefits. *Annual Review of Ecology, Evolution, and Systematics*, **41**, 149–172.
- 398 Epanchin-Niell, R.S., Haight, R.G., Berec, L., Kean, J.M. & Liebhold, A.M. (2012) Optimal surveillance and  
399 eradication of invasive species in heterogeneous landscapes. *Ecology Letters*, **15**, 803–812.
- 400 Fenk, L., Vogel, B. & Horvath, H. (2007) Dispersion of the bio-aerosol produced by the oak processionary  
401 moth. *Aerobiologia*, **23**, 79–87.
- 402 Fitzpatrick, M., Preisser, E., Ellison, A. & Elkinton, J.S. (2009) Observer bias and the detection of low-density  
403 populations. *Ecological Applications*, **19**, 1673–1679.
- 404 Fox, R., Oliver, T.H., Harrower, C., Parsons, M.S., Thomas, C.D. & Roy, D.B. (2014) Long-term changes to the  
405 frequency of occurrence of British moths are consistent with opposing and synergistic effects of climate  
406 and land-use changes. *Journal of Applied Ecology*, **51**, 949–957.
- 407 Fox, R., Randle, Z., Hill, L., Anders, S., Wiffen, L. & Parsons, M.S. (2010) Moths count: recording moths for  
408 conservation in the UK. *Journal of Insect Conservation*, **15**, 55–68.
- 409 Fry, R. & Waring, P. (2001) *A Guide to Moth Traps and Their Use*. Amateur Entomologists' Society, London,  
410 UK.
- 411 Gallo, T. & Waitt, D. (2011) Creating a successful citizen science model to detect and report invasive species.  
412 *BioScience*, **61**, 459–465.

- 413 Gottschling, S. & Meyer, S. (2006) An epidemic airborne disease caused by the oak processionary caterpillar.  
414 *Pediatric Dermatology*, **23**, 64–6.
- 415 Graham, J., Simpson, A., Crall, A., Jarnevich, C., Newman, G. & Stohlgren, T.J. (2008) Vision of a  
416 cyberinfrastructure for nonnative, invasive species management. *BioScience*, **58**, 263–268.
- 417 Groenen, F. & Meurisse, N. (2012) Historical distribution of the oak processionary moth *Thaumetopoea*  
418 *processionaria* in Europe suggests recolonization instead of expansion. *Agricultural and Forest*  
419 *Entomology*, **14**, 147–155.
- 420 Groenendijk, D. & Ellis, W.N. (2010) The state of the Dutch larger moth fauna. *Journal of Insect Conservation*,  
421 **15**, 95–101.
- 422 Hauser, C.E. & McCarthy, M.A. (2009) Streamlining ‘search and destroy’: cost-effective surveillance for  
423 invasive species management. *Ecology Letters*, **12**, 683–92.
- 424 Hesterberg, U., Harris, K., Stroud, D., Guberti, V., Busani, L., Pittman, M., Piazza, V., Cook, A. & Brown, I.  
425 (2009) Avian influenza surveillance in wild birds in the European Union in 2006. *Influenza and Other*  
426 *Respiratory Viruses*, **3**, 1–14.
- 427 Hobbs, R.J. & Humphries, S.E. (1995) An integrated approach to the ecology and management of plant  
428 invasions. *Conservation Biology*, **9**, 761–770.
- 429 Ibáñez, I., Silander Jr, J.A., Allen, J.M., Treanor, S.A. & Wilson, A. (2009) Identifying hotspots for plant  
430 invasions and forecasting focal points of further spread. *Journal of Applied Ecology*, **46**, 1219–1228.
- 431 Isaac, N.J.B., van Strien, A.J., August, T.A., de Zeeuw, M.P. & Roy, D.B. (2014) Statistics for citizen science:  
432 extracting signals of change from noisy ecological data. *Methods in Ecology and Evolution*, **5**, 1052–1060.
- 433 Kulasekera, V.L., Kramer, L., Nasci, R.S., Mostashari, F., Cherry, B., Trock, S.C., Glaser, C. & Miller, J.R.  
434 (2000) West Nile virus infection in mosquitoes, birds, horses, and humans, Staten Island, New York,  
435 2000. *Emerging Infectious Diseases*, **7**, 722–5.
- 436 Maxwell, B.D., Lehnhoff, E. & Rew, L.J. (2009) The rationale for monitoring invasive plant populations as a  
437 crucial step for management. *Invasive Plant Science and Management*, **2**, 1–9.
- 438 Mehta, S. V., Haight, R.G., Homans, F.R., Polasky, S. & Venette, R.C. (2007) Optimal detection and control  
439 strategies for invasive species management. *Ecological Economics*, **61**, 237–245.
- 440 Millennium Ecosystem Assessment. (2005) *Ecosystems and Human Well-Being: Biodiversity Synthesis*. World  
441 Resources Institute, Washington DC.
- 442 Mindlin, M.J., le Polain de Waroux, O., Case, S. & Walsh, B. (2012) The arrival of oak processionary moth, a  
443 novel cause of itchy dermatitis, in the UK: experience, lessons and recommendations. *Public Health*, **126**,  
444 778–81.
- 445 Nov, O., Arazy, O. & Anderson, D. (2014) Scientists@Home: what drives the quantity and quality of online  
446 citizen science participation? *PloS ONE*, **9**, e90375.
- 447 Pejchar, L. & Mooney, H.A. (2009) Invasive species, ecosystem services and human well-being. *Trends in*  
448 *Ecology & Evolution*, **24**, 497–504.
- 449 Pocock, M.J.O., Chapman, D.S., Sheppard, L.J. & Roy, H.E. (2013) *Developing a Strategic Framework to*  
450 *Support Citizen Science Implementation. Final Report to SEPA*. Centre for Ecology & Hydrology,  
451 Wallingford, Oxon.
- 452 Pocock, M.J.O., Roy, H.E., Preston, C.D. & Roy, D.B. (2015) The Biological Records Centre: a pioneer of  
453 citizen science. *Biological Journal of the Linnean Society*, **115**, 475–493.
- 454 Pyšek, P. & Richardson, D.M. (2010) Invasive species, environmental change and management, and health.  
455 *Annual Review of Environment and Resources*, **35**, 25–55.
- 456 Rotman, D., Preece, J., Hammock, J., Procita, K., Hansen, D., Parr, C., Lewis, D. & Jacobs, D. (2012) Dynamic  
457 changes in motivation in collaborative citizen-science projects. *Proceedings of the ACM 2012 Conference*  
458 *on Computer Supported Cooperative Work*, 217.
- 459 Roy, H.E., Bacon, J., Beckmann, B., Harrower, C.A., Hill, M.O., Isaac, N.J.B., Preston, C.D., Rathod, B.,  
460 Rorke, S.L., Marchant, J.H., Musgrove, A., Noble, D., Sewell, J., Seeley, B., Sweet, N., Adams, L.,  
461 Bishop, J., Jukes, A.R., Walker, K.J. & Pearman, D. (2012) *Non-Native Species in Great Britain:*  
462 *Establishment, Detection and Reporting to Inform Effective Decision Making*. Defra, London, UK.
- 463 Roy, H.E., Peyton, J., Aldridge, D.C., Bantock, T., Blackburn, T.M., Britton, R., Clark, P., Cook, E., Dehnen-

464 Schmutz, K., Dines, T., Dobson, M., Edwards, F., Harrower, C., Harvey, M.C., Minchin, D., Noble, D.G.,  
465 Parrott, D., Pocock, M.J.O., Preston, C.D., Roy, S., Salisbury, A., Schönrogge, K., Sewell, J., Shaw, R.H.,  
466 Stebbing, P., Stewart, A.J.A. & Walker, K.J. (2014) Horizon scanning for invasive alien species with the  
467 potential to threaten biodiversity in Great Britain. *Global Change Biology*, **20**, 3859–3871.

468 Roy, H.E., Rorke, S.L., Beckmann, B., Booy, O., Botham, M.S., Brown, P.M.J., Harrower, C., Noble, D.,  
469 Sewell, J. & Walker, K. (2015) The contribution of volunteer recorders to our understanding of biological  
470 invasions. *Biological Journal of the Linnean Society*, **115**, 678–689.

471 Sharov, A.A., Leonard, D., Liebhold, A.M., Roberts, E.A. & Dickerson, W. (2002) ‘Slow The Spread’: A  
472 national program to contain the gypsy moth. *Journal of Forestry*, **100**, 30–36.

473 Snäll, T., Kindvall, O., Nilsson, J. & Pärt, T. (2011) Evaluating citizen-based presence data for bird monitoring.  
474 *Biological Conservation*, **144**, 804–810.

475 Sullivan, B.L., Aycrigg, J.L., Barry, J.H., Bonney, R.E., Bruns, N., Cooper, C.B., Damoulas, T., Dhondt, A.A.,  
476 Dietterich, T., Farnsworth, A., Fink, D., Fitzpatrick, J.W., Fredericks, T., Gerbracht, J., Gomes, C.,  
477 Hochachka, W.M., Iliff, M.J., Lagoze, C., La Sorte, F.A., Merrifield, M., Morris, W., Phillips, T.B.,  
478 Reynolds, M., Rodewald, A.D., Rosenberg, K. V., Trautmann, N.M., Wiggins, A., Winkler, D.W., Wong,  
479 W.-K., Wood, C.L., Yu, J. & Kelling, S. (2014) The eBird enterprise: An integrated approach to  
480 development and application of citizen science. *Biological Conservation*, **169**, 31–40.

481 Szabo, J.K., Vesk, P.A., Baxter, P.W.J. & Possingham, H.P. (2010) Regional avian species declines estimated  
482 from volunteer-collected long-term data using List Length Analysis. *Ecological Applications*, **20**, 2157–  
483 2169.

484 Townsend, M. (2013) Oak Processionary Moth in the United Kingdom. *Outlooks on Pest Management*, **24**, 32–  
485 38.

486 Tree Health and Plant Biosecurity Expert Taskforce. (2012) *Interim Report*.

487 Wagenhoff, E. & Veit, H. (2011) Five Years of Continuous *Thaumetopoea processionea* Monitoring: Tracing  
488 Population Dynamics in an Arable Landscape of South-Western Germany. *Gesunde Pflanzen*, **63**, 51–61.

489 Williams, D.T., Straw, N., Townsend, M., Wilkinson, A.S. & Mullins, A. (2013) Monitoring oak processionary  
490 moth *Thaumetopoea processionea* L. using pheromone traps: the influence of pheromone lure source, trap  
491 design and height above the ground on capture rates. *Agricultural and Forest Entomology*, **15**, 126–134.

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<br>probability of<br>recording ( $\bar{S}$ )          | Percentage of 10km grid squares |                |                  |                | Percentage of 1km grid squares |                |                  |                |
|---|---------------------------------|----------------|------------------|----------------|--------------------------------|----------------|------------------|----------------|
|   | Very low<br>(0.05%)             | Low<br>(0.39%) | Medium<br>(1.4%) | High<br>(2.4%) | Very low<br>(0.05%)            | Low<br>(0.39%) | Medium<br>(1.4%) | High<br>(2.4%) |
| Threshold for<br>predicted detection<br>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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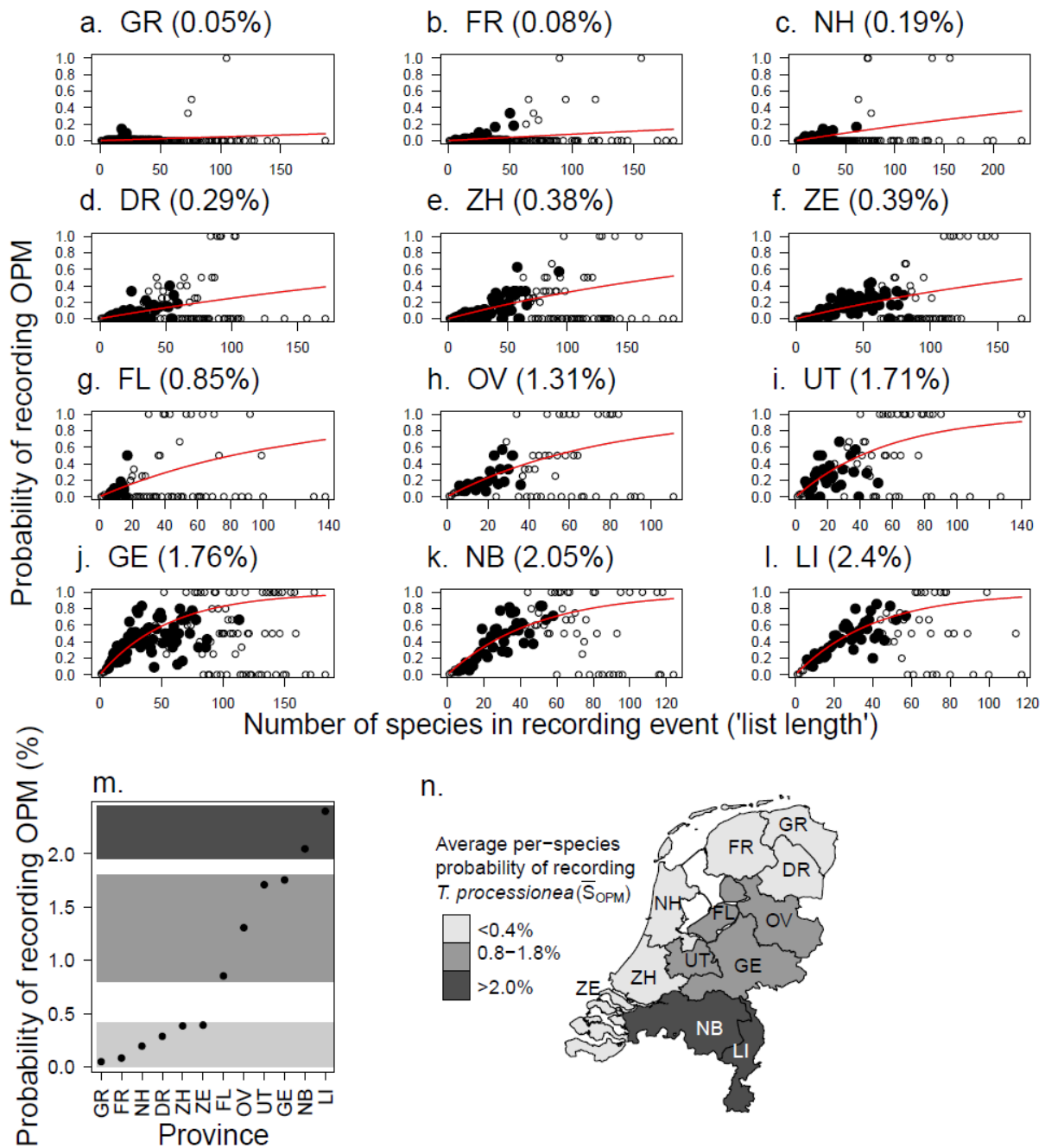
504 **Table 2.** A framework for considering the role of citizen science in the detection of rare events, such  
 505 as invasive or rare species

| 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).

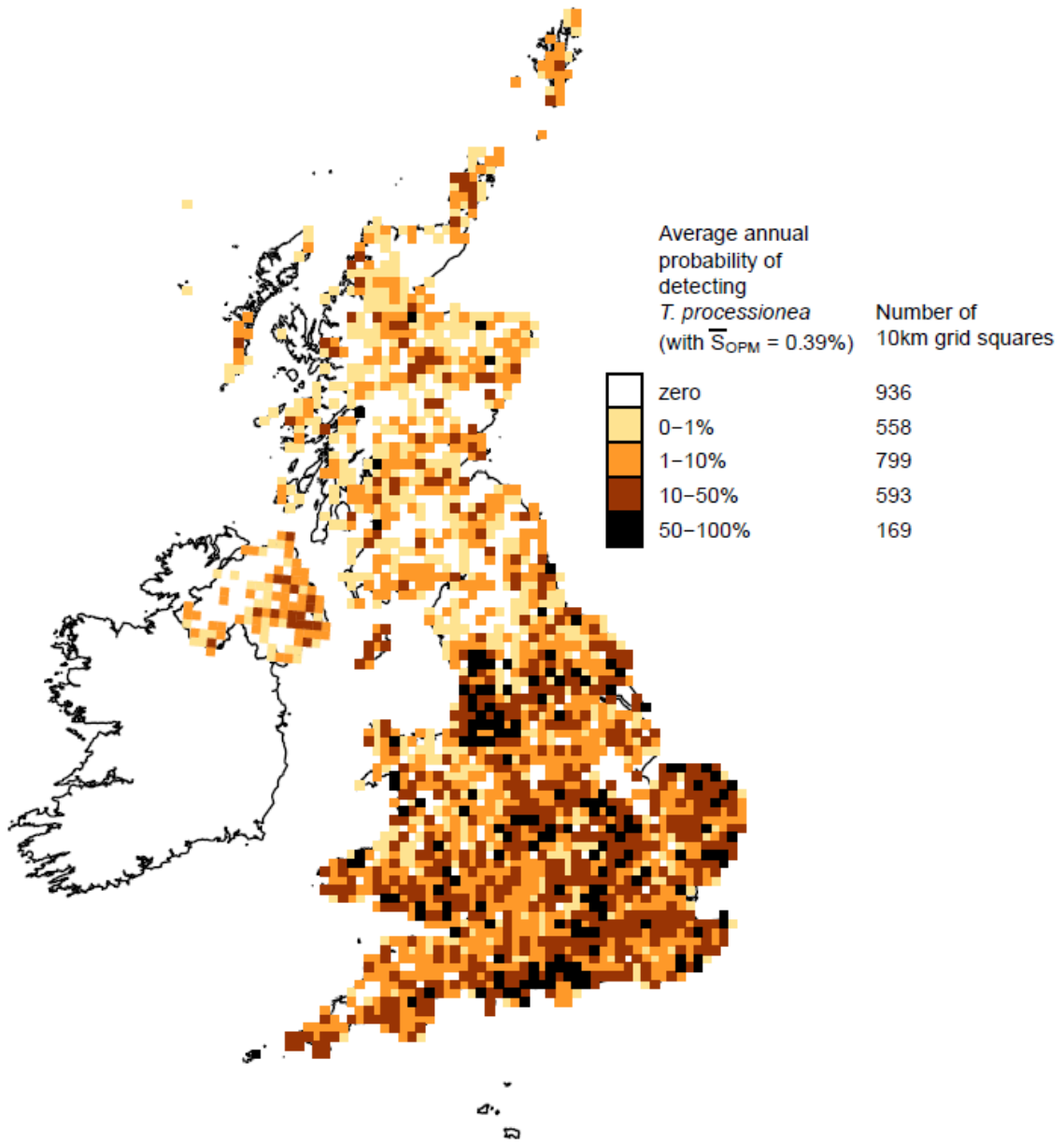
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518 **Fig. 2.** The average annual probability of detecting *T. processionea* ( $\hat{D}$ ), if it were present, in 10km  
519 grid squares in the UK based on the observed recording effort during 25 July-30 August in 2000-  
520 2009. The results are 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).

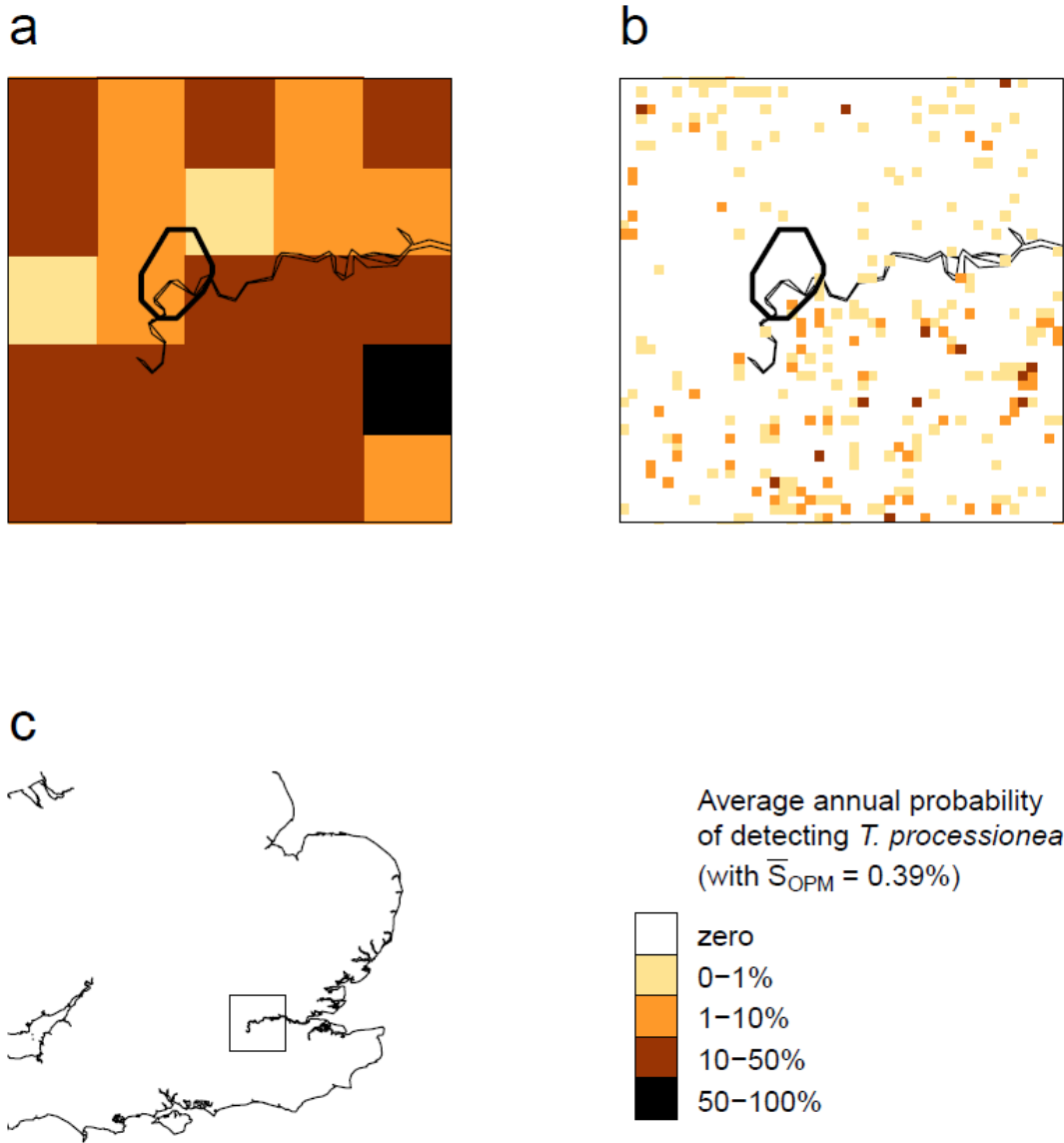
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525 **Figure 3.** The probability of detecting *T. processionea*, if it was present, in (a) 10km and (b) 1km grid  
 526 squares in a 50km square containing the current range of *T. processionea* in west London (thick black  
 527 outline is the minimum 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.

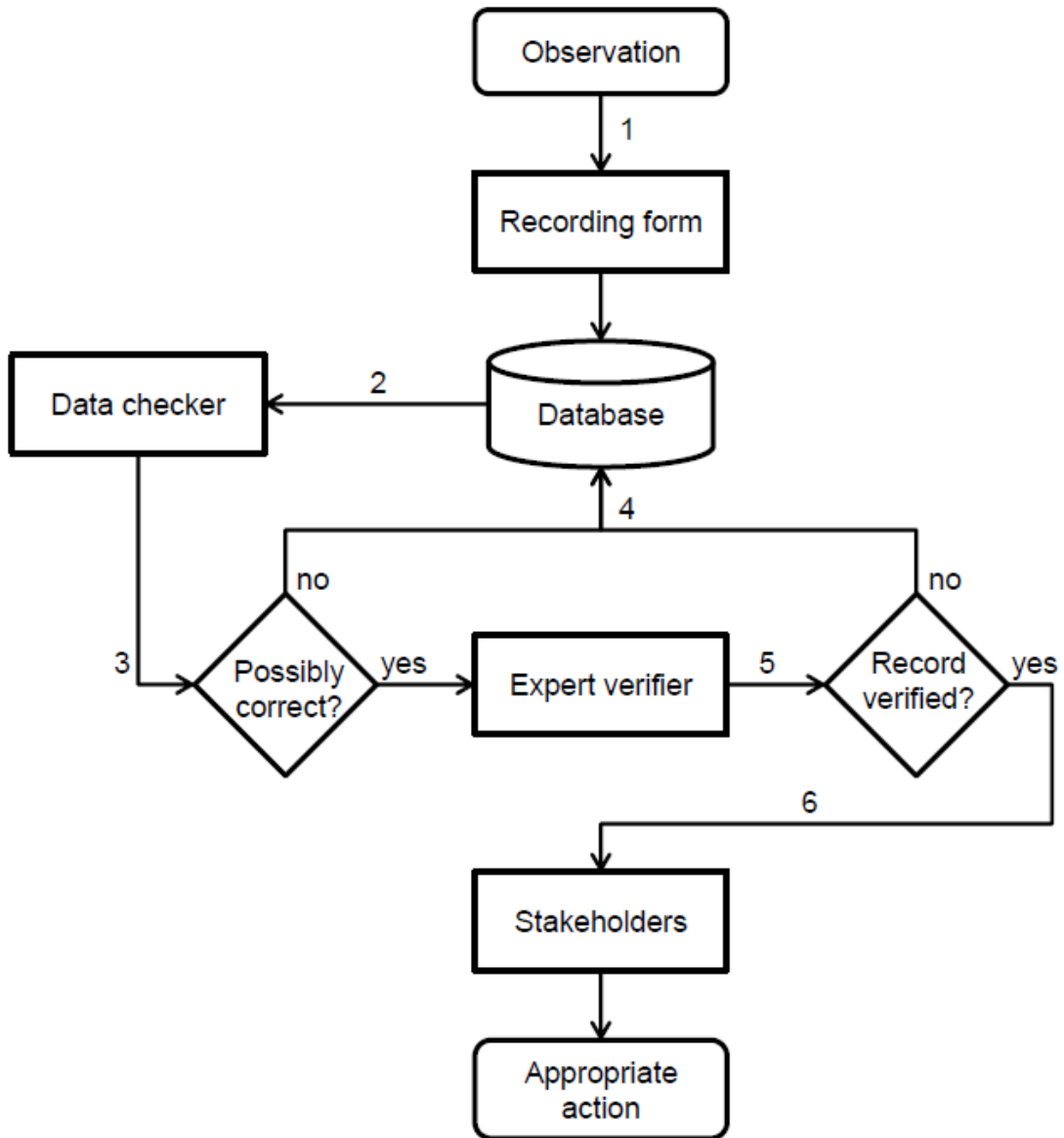


Fig. 4.

538