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1 **Dynamic role of grasslands as sources of soil-dwelling insect pests: new insights from *in silico***  
2 **experiments for pest management strategies**

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11 **Keywords:** pest control, population dynamics, landscape dynamics, reservoir, sink, connectivity,  
12 reaction-diffusion-advection model, land-use change, wireworm, click beetle

13

14 **Highlights:**

- 15 • The role of grassland in field colonisation by insect pests is investigated *in silico*
- 16 • Population dynamics of wireworms is modelled, accounting for life cycle and dispersion
- 17 • Spatiotemporal effects of grassland in field history, field neighbourhood or their interaction,  
18 depend on the pest life cycle and its dispersal capacity
- 19 • Grassy areas can act either as pseudo-sinks or as sources on neighbouring crops
- 20 • Control strategies based on grassland arrangements appear trait-specific

21

22 **Abstract:**

23 Sustainable pest control strategies hinge on the knowledge of movement ecology within the agricultural  
24 landscape where contrasted habitat qualities intermingle, thereby influencing arthropod dispersal.  
25 Hence, habitat manipulation in space and time can be a lever for action to control pests with regard to  
26 landscape compositional constraints. In this study, we examined the role of grassland arrangements in  
27 field colonisation by soil-dwelling pests within a dynamic agricultural landscape, and discussed the  
28 implications for pest management with a focus on wireworms. For this purpose, we proposed a  
29 framework combining (i) a spatially explicit and mechanistic model describing the pest population  
30 dynamics in both aerial and soil compartments involved along its life cycle, and (ii) spatiotemporal  
31 representations of various landscape contexts. We addressed the role of grassland in plot history, in plot  
32 neighbourhood, or in both history and neighbourhood. Our results show that species with a short life  
33 cycle are more responsive to changes in land use, and that the neighbourhood effect strongly relies on  
34 dispersal mechanisms (random vs directed movements). We also highlight how the arrangement of  
35 grassy landscape elements in space and time can mitigate crop infestation by soil-dwelling pests, thereby  
36 emphasizing the relevance of managing grassland regimes. Once informed by critical pest life traits, our  
37 approach opens avenue for the exploration of the spatiotemporal land use manipulation meant for pest

38 management. Future research consists in the exploration of suppressive patterns in simplified but  
39 realistic agricultural landscapes, generated under agronomic constraints at the farm or landscape scales.

40

41

## 42 **1. Introduction**

43 Targeting a sustainable crop protection, freed from its reliance on pesticides, requires a better  
44 understanding of field colonisation processes driving pest population dynamics. Looking for  
45 environmental-friendly strategies for pest control reemphasizes the importance of movement ecology.  
46 Indeed, insect pest habitats are fragmented rendering dispersal a key process determining pest population  
47 dynamics at the agricultural landscape scale. This suggests that agricultural landscape is a relevant scale  
48 for designing pest resilient agroecosystems that minimise the use of pesticides (Bourhis et al., 2017,  
49 2015; Tschardt and Brandl, 2004).

50 Such a pest control approach should account for processes operating at different spatial and temporal  
51 scales, thereby requiring an extensive knowledge about pest ecology. The survival of species hinge on  
52 their ability to detect and reach suitable resources when necessary in their life cycle (Dunning et al.,  
53 1992; Vasseur et al., 2013). Interestingly, a few studies have suggested that some insects effectively  
54 move between asynchronous land uses depending on resource availability (Aviron et al., 2018; Bressan  
55 et al., 2010; Men et al., 2004). Habitat connectivity in space and time (Martensen et al., 2017; Moilanen  
56 and Hanski, 2001; Taylor et al., 1993) is thus a pivotal driver of pest dispersal success in dynamic  
57 agricultural landscapes. Consequently, the spatial and temporal arrangement of land uses can be a lever  
58 for action to control species abundances with regard to landscape compositional constraints (Fischer and  
59 Lindenmayer, 2007; Jonsson et al., 2010; Parisey et al., 2016; Polasky et al., 2008; Roques and Hamel,  
60 2007; With and King, 2001)

61 In that, Elateridae can be considered as relevant model organisms (Blackshaw et al., 2017). Wireworms,  
62 the larvae of click beetles (Coleoptera: Elateridae) of which there are more than 10,000 species in about  
63 400 genera known worldwide, are among the most notorious soil-dwelling pests (Balachowsky and  
64 Mesnil, 1935; Miles, 1942; Traugott et al., 2015). Elaterid pest species are capable of exploiting both  
65 cultivated and uncultivated areas in the agricultural mosaic (Milosavljević et al., 2016). Consequently,  
66 understanding their movement from sources that sustain important populations, i.e. uncultivated areas,  
67 to vulnerable crops is central in the development of pest management strategies. The larvae undergo a  
68 number of instars and the complete elaterid life cycle varies between two to five years (Balachowsky  
69 and Mesnil, 1935; Furlan, 1998; Miles, 1942; Sufyan et al., 2013). In Europe, *Agriotes* spp. are the most  
70 harmful species. They attack the roots and tubers of a wide range of crops, thereby inflicting severe  
71 economic damage in potatoes, maize and vegetable production, and may even threaten perennial crops  
72 such as soft fruits or orchards (Traugott et al., 2015).

73 Though scarce and fragmented, knowledge regarding the biology and the ecology of *Agriotes* spp.  
74 indicates that uncropped field margins and areas are the most favourable habitat for wireworms, and act

75 as reservoirs from which they disperse into adjacent crops (Blackshaw et al., 2017). Based on correlative  
76 approaches, some studies have attempted to unravel relationships between larvae or adult abundances  
77 or densities and landscape features (Benefer et al., 2012; Blackshaw and Vernon, 2006; L. Furlan et al.,  
78 2017; Hermann et al., 2013; Kozina et al., 2015; Poggi et al., 2018a; Saussure et al., 2015). Altogether,  
79 these studies shed light on the potential for a dynamic landscape to shape wireworm populations.  
80 However, while providing insights of the main factors responsible for wireworm or click beetle density  
81 levels, such approaches fail in describing the mechanisms driving pest colonisation, and subsequently  
82 in elucidating the source/sink ecological processes that operate at the landscape scale.

83 The main objective of our study was to examine the role of grasslands in field colonisation by soil-  
84 dwelling pests within a dynamic agricultural landscape, and discuss its implications for pest  
85 management with a focus on wireworms. For this purpose, we first developed a spatially explicit and  
86 mechanistic model describing the pest population dynamics in both aerial and soil compartments  
87 involved along its life cycle. Larvae and adult mortality, larval development, oviposition, and adult  
88 movements (diffusion and advection) were the main processes considered. Model parameterisation  
89 stems from an extensive review of the literature dealing with the biology and ecology of wireworms.  
90 Then, landscape elements comprising grasslands and vulnerable habitats (cropped fields) were  
91 characterised by their carrying capacity. Combining these two framework components, we investigated  
92 the interplay of landscape spatiotemporal dynamics, considering grasslands as pest sources, and of  
93 species-specific traits (e.g. life cycle duration) on field infestation levels, as illustrated by the conceptual  
94 diagram (Fig. 1). We also highlighted how the balance between diffusion and advection movements  
95 drives the population flows on both sides of the boundaries between grassland and field crops. Finally,  
96 we demonstrated that spatiotemporal connectivity between grasslands significantly affects the pest  
97 infestation levels within crops.

98 Providing new insights into the role of grasslands in pest infestation in a dynamic agricultural mosaic,  
99 our findings bring an original contribution to the search for innovative landscape-scale strategies for  
100 pest management.

101

## 102 **2. Material and methods**

103

### 104 **2.1. Population dynamics model**

105 Insect pests considered in this study cause crop damage during larval development in soil but disperse  
106 as adults by walk and flight. Both the aboveground and the belowground compartments were explicitly  
107 considered. The population dynamics was derived along two spatial dimensions  $\mathbf{x}=(x,y)$  through a set of  
108 reaction-advection-diffusion equations modelling the following biological and ecological mechanisms:

- 109 • the emergence of adults from mature larvae,
- 110 • the adult random and directed motions,
- 111 • the oviposition,

- 112 • the adult mortality,
- 113 • the larval development,
- 114 • and the larval density-dependent mortality.

115 Briefly, adults oviposit at a constant rate  $\pi$ , thereby increasing the belowground population that ages at  
 116 a constant velocity  $c$ , undergoing a density-dependent mortality. Wireworms reaching the critical age  
 117  $m_c$  emerge as adults that disperse across the landscape to find resources and suitable habitat. The model  
 118 can be formalised as a system of partial differential equations: for  $t > 0$  and  $0 < m < m_c$

$$119 \quad \begin{cases} \partial_t A(\mathbf{x}, t) = \tau(t)B(\mathbf{x}, t, m_c) + D\Delta_{\mathbf{x}}A(\mathbf{x}, t) - \vec{u}(\mathbf{x}, t) \cdot \nabla_{\mathbf{x}}A(\mathbf{x}, t) - \mu_A A(\mathbf{x}, t) & (1) \\ \partial_t B(\mathbf{x}, t, m) = -c\partial_m B(\mathbf{x}, t, m) - \mu_B \left(\frac{\bar{B}(\mathbf{x}, t)}{K(\mathbf{x}, t)}\right)^\beta B(\mathbf{x}, t, m) & (2) \end{cases}$$

120 with the dynamical boundary conditions w.r.t the maturation variable

$$121 \quad \begin{cases} \partial_t B(\mathbf{x}, t, 0) = \pi A(\mathbf{x}, t) - \mu_B \left(\frac{\bar{B}(\mathbf{x}, t)}{K(\mathbf{x}, t)}\right)^\beta B(\mathbf{x}, t, 0) & (3) \\ \partial_t B(\mathbf{x}, t, m_c) = -c\partial_m B(\mathbf{x}, t, m_c) - \tau(t)B(\mathbf{x}, t, m_c) - \mu_B \left(\frac{\bar{B}(\mathbf{x}, t)}{K(\mathbf{x}, t)}\right)^\beta B(\mathbf{x}, t, m_c) & (4) \end{cases}$$

122 Here,  $\bar{B}(\mathbf{x}, t) = \int_0^{m_c} B(\mathbf{x}, t, m)dm$  represents the total belowground population,  $\Delta_{\mathbf{x}} = \partial_x^2 + \partial_y^2$  is the  
 123 Laplace operator (diffusion over  $\mathbf{x}$ ),  $\nabla_{\mathbf{x}} = \left(\frac{\partial}{\partial x}\right)$  and  $\partial_m$  are the advection operators over  $\mathbf{x}$  and  $m$  (the  
 124 maturity dimension), where  $\partial_x, \partial_y, \partial_m$  denote the partial derivatives with respect to  $x, y$  and  $m$ .

125 Model parameters and their meaning are described in Table 1. Model interpretation is fully presented in  
 126 the following paragraphs. The state variables A and B respectively denote the density of click beetles  
 127 aboveground, and the density of wireworms in soil. Actually, B encapsulates all belowground  
 128 development stages (larval instars, pupae) and is referred to as wireworms thereafter for the sake of  
 129 simplification. The dynamical condition at  $m = 0$  (Eq. 3) reflects both the oviposition and the growth  
 130 of individuals who did not start to mature, while the condition in  $m = m_c$  (Eq. 4) takes into account the  
 131 accumulation of individuals in the last class of maturation before emergence.. As initial conditions, the  
 132 aboveground population was set to zero and the age distribution of larvae was obtained from a  
 133 preliminary model simulation run over 15 years on a homogeneous landscape. The model was solved  
 134 using operator splitting (Press et al., 2007, Chapter 20.3.3). Alternating-direction implicit method solves  
 135 the diffusion (Press et al., 2007, Chapter 20.3.2), while a forward Euler scheme is used for the reaction  
 136 terms, and an upwind scheme for the advection terms (Press et al., 2007, Chapter 20.1). We considered  
 137 the spatial domain under study as a closed system (i.e. equal inflowing and outflowing pest  
 138 populations), thereby applying Neumann (reflecting) boundary conditions at the edges of the spatial  
 139 domain. All developments were made with the R programming language (R Core Team, 2018).

140

141 *Adult emergence and motions*

142 Click beetles emerge during spring. In accordance, we defined their emergence rate as a time-dependent  
143 function  $\tau(t)$  (cf. §2.2 and Suppl. Material 1) in Equation 1.

144 As recent findings suggested that wireworms do not move from field to field (Schallhart et al., 2011),  
145 we considered that only adults contribute to the spatial redistribution of populations. We assumed that  
146 adult motions included two components: a random component expressed through the diffusion term in  
147 Equation 1, involving  $D$  the homogeneous diffusion coefficient; and a directed component expressed  
148 through the advection term in Equation 1 and involving  $\vec{u}$  the advection velocity. Regarding the latter  
149 component, we assumed that click beetles move across the landscape based on the habitat quality they  
150 perceive in their neighbourhood. We defined the local habitat quality  $h(\mathbf{x}, t)$  as a function of local  
151 carrying capacities  $K(\mathbf{x}, t)$ :

$$h(\mathbf{x}, t) = \frac{K(\mathbf{x}, t) - K_{min}}{K_{max} - K_{min}} \quad (5)$$

152 where  $K_{max}$  (resp.  $K_{min}$ ) refers to the most favourable (resp. unfavourable) habitat for wireworms. As did  
153 Bourhis et al. (2017), we derived the perception of the landscape habitat quality, noted  $h_p(\mathbf{x}, t)$ , from a  
154 convolution of a perception kernel (linear inverse distance weighting filter) with the mapping of local  
155 habitat quality:

$$h_p(\mathbf{x}, t) = \frac{\int_{\substack{x_i \in \Omega \\ |x-x_i| < \gamma}} \left(1 - \frac{|x-x_i|}{\gamma}\right) h(x_i, t) dx_i}{\int_{\substack{x_i \in \Omega \\ |x-x_i| < \gamma}} \left(1 - \frac{|x-x_i|}{\gamma}\right) dx_i} \quad (6)$$

156 with  $\Omega$  the whole spatial domain, and  $\gamma$  the distance beyond which habitat quality is no longer perceived  
157 by click beetles. Finally, the advection velocity

$$\vec{u}(\mathbf{x}, t) = V_{max} \frac{\nabla_x h_p(\mathbf{x}, t)}{\max|\nabla_x h_p(\mathbf{x}, t)|} \quad (7)$$

158 reaches its maximum speed  $V_{max}$  where the gradient of the potential function  $h_p(\mathbf{x}, t)$  is maximum.

159

160 *Oviposition*

161 Although the oviposition rate may vary among habitats (Brian, 1947), it was assumed here independent  
162 of space (i.e. land use and associated carrying capacity) and constant over the whole duration of the  
163 adult life. New-borns fall within the belowground population with age 0 (Eq. 3).

164

165 *Adult mortality*

166 A constant mortality rate  $\mu_A$  was applied to the aboveground compartment of the model.

167

168 *Larval development*

169 The belowground population distributes along an ageing axis. The population ages according to a  
170 constant coefficient  $c$ . When reaching the critical maturity  $m_c$ , larvae can emerge as click beetles.

171

172 *Larval mortality*

173 Uncultivated areas, such as grasslands or grassy strips, are acknowledged as favourable habitats for both  
174 adults and larvae of click beetles (Brian, 1947; Miles, 1942), providing sufficient food and humid  
175 conditions (Roebuck et al., 1947), therefore wireworm populations thrive in these areas. On the contrary,  
176 wireworm populations reach much lower infestation levels in cropped fields, where they nevertheless  
177 damage vulnerable crops (Miles, 1942). In our model, we assumed that the carrying capacity  $K(x,t)$  is a  
178 proxy for habitat quality, and that larval mortality relies on a constant component  $\mu_B$  weighted by a  
179 density-dependent factor. While the constant component provides a stationary annual dynamics within  
180 the landscape, the density-dependent term ensures that the population dynamics is centred on the  
181 carrying capacity. The exponent  $\beta$  controls the dynamics inertia, i.e. the speed at which the local  
182 population converges towards the local carrying capacity. For the sake of simplicity, larval mortality  
183 was independent of the age. Note that the reaction term departs from the canonical logistic growth model  
184 as the density dependence applies on larval mortality only rather than on larval population growth rate  
185 (which also depends on adult oviposition rate), emphasising that population is limited by larval  
186 competition for food rather than by adult competition for oviposition sites.

187

188 **2.2. Model parameterisation**

189

190 *Land-use dependent carrying capacities ( $K_G$  and  $K_C$ )*

191 We considered two types of land use: grasslands, known as favourable habitats (Furlan, 2004; Parker  
192 and Howard, 2001), and cropped fields.

193 Regarding grasslands, we set the carrying capacity  $K_G$  to 2000 individuals per square meter (thereafter  
194 ind/m<sup>2</sup>) based on field studies carried out before the common access to chemical pesticides (Roebuck et  
195 al., 1947; Ross et al., 1947; Salt and Hollick, 1944). Based on a thorough study of two pastures at least  
196 10-year old and regularly cut for hay, Salt and Hollick (1944) estimated densities about 850 and 2000  
197 ind/m<sup>2</sup>. Locally, they even observed densities reaching 6000 ind/m<sup>2</sup> in both pastures. Within other  
198 pastures less intensively sampled, they found about 2500 ind/m<sup>2</sup> on average. Roebuck et al. (1947)  
199 studied intensively the infestation level of an old meadow, and determined a density of about 300 ind/m<sup>2</sup>  
200 by the hand-sorting method. Following Salt and Hollick (1944) who consider that this method  
201 underestimates densities by two thirds, the actual infestation could be of about 1000 ind/m<sup>2</sup>. Ross et al.  
202 (1947) studied six permanent pastures grazed and regularly hayed, and assessed a density of about 1000  
203 ind/m<sup>2</sup>.

204 In cropped fields, we set the carrying capacity  $K_C$  to 120 ind/m<sup>2</sup>, i.e. about one order of magnitude  
205 smaller than  $K_G$ , and fairly close to values indicated by Roebuck (1924). Indeed, Roebuck (1924) studied  
206 the decline of wireworm populations in newly set up crop fields with repeated ploughing, and found

207 average population density of 102, 80, 56 and 32 ind/m<sup>2</sup> the 1st, 2nd, 3rd and 4th year after ploughing  
208 from grass, respectively.

209

#### 210 *Belowground population maturation (c)*

211 Wireworm species differ greatly from one another in their biology and ecology (Furlan, 2005; Ritter and  
212 Richter, 2013), in particular as regards the duration of their biological cycle. In France, four species  
213 from the genus *Agriotes* are responsible for most economical crop damage. Among them, *A. lineatus*,  
214 *A. obscurus* and *A. sputator*, predominantly present in the northern regions, exhibit a long life cycle  
215 (duration of larval stage of about 4 years), whereas *A. sordidus*, mainly found in south-western regions,  
216 exhibits a short life cycle (duration of larval stage of about 2 years). In our study, we set the duration of  
217 the larval stage ( $c^{-1}$ ) at 2 years for short life-cycle species and at 4 years for long life-cycle species.

218

#### 219 *Emergence rate*

220 Studies relative to the date of emergence of *Agriotes* click beetles are fairly consistent. For instance, in  
221 Austria, Landl et al. (2010) observed that the emergence of *A. lineatus* began at the end of April, which  
222 is concordant with the observations of other authors from Eastern Europe and from western Canada  
223 (Vernon et al., 2001). Regarding *A. ustulatus*, *A. lineatus* and *A. obscurus*, Parker and Howard (2001)  
224 in United Kingdom reported emergences starting at mid-April and adults being captured until mid-July  
225 (hence, they probably emerged until mid-June), while in Germany, Sufyan et al. (2007) observed that  
226 the swarming period lasted from late April to the end of August (for *A. lineatus* and *A. obscurus*).  
227 Although most authors reported two periods of adult activity, one main in May and one weaker in June-  
228 July (Landl et al., 2010; Sufyan et al., 2007), we simply modelled adult emergence as a continuous  
229 process spanning over a period governed by the cumulative degree days (see Supplementary Material  
230 1). In our study, the emergence of click beetles spans from late April to the beginning of June, with a  
231 peak of emergence located towards the middle of the period of emergence (i.e. mid-May). Over this  
232 period, all wireworms at the critical age have emerged (see Supplementary Material 1, Fig. A1B).

233

#### 234 *Belowground mortality ( $\mu_B, \beta$ )*

235 Two main model parameters govern the larval mortality, namely the mortality rate  $\mu_B$  and the inertia  
236 exponent  $\beta$ . The mortality rate was determined by means of a preliminary optimisation procedure before  
237 each simulation, given all other parameter values, such that larval density oscillated yearly around the  
238 carrying capacity in any homogeneous landscape.  $\beta$  was estimated empirically to 0.2 from infestation  
239 data collected in arable lands by Roebuck (1924) over 8 years. As shown in a sensitivity analysis (see  
240 Supplementary Material 2), despite scarce knowledge on this parameter  $\beta$ , it has a significant effect on  
241 the model output and its estimation would probably deserve further attention.

242



243 *Adult dispersion ( $D, V_{max}$ )*

244 Literature provides a limited amount of information regarding click-beetle dispersal. In particular,  
245 diffusion and advection processes can seldom be disentangled. Brian (1947) observed that *A. obscurus*  
246 and *A. lineatus* (but not *A. sputator*) frequently flew in a glass cage experiment. Crozier et al. (2003)  
247 reported that these species can be strong fliers, able to perform mass flights, travelling approximately 5  
248 to 10 km/h for distances ranging from less than 1 m to 100 m on one occasion and with an average flight  
249 covering 2 to 3 m. More recently, Blackshaw and Hicks (2013) concluded that *Agriotes spp.* disperse  
250 widely and may originate from adjacent fields. Many studies have attempted to provide quantitative  
251 estimates of click-beetle dispersal capacity, generally using mark-recapture experiments with  
252 pheromone traps. Blackshaw et al. (2017) recaptured individuals (*A. lineatus*, *A. obscurus*, *A. sputator*)  
253 at 1m from release point after 1h, and at 30m after 19h. Schallhart et al. (2009) studied the isotopic  
254 signature of *A. obscurus* beetles captured in pheromone traps located in a grassland adjacent to a maize  
255 field: 13% of individuals captured 80m inside the grassland plot had a signature characteristic of their  
256 maize field origin. Based on this knowledge, we set the diffusion coefficient to 72 m<sup>2</sup>/day, hence  
257 producing a mean dispersion radius of 15m in one day, 21m in two days, etc. (Shigesada and Kawasaki  
258 (1997), page 38). The advection speed was set to 10 m/day.

259

260 *Perceptual range ( $\gamma$ )*

261 We assumed that adult click beetles could perform directed motions as responses to distant stimuli (e.g.  
262 olfactory stimuli). Advection processes may result from foraging for food or laying sites, from searching  
263 for a sexual partner, etc. In our model, we described advection at the population level as a tendency to  
264 move along a gradient of increasing habitat quality expressed in terms of carrying capacities. In  
265 particular, as wireworms thrive in grasslands, we expect adult click beetles to disperse preferentially  
266 towards nearby grassy areas. We set the perceptual range (i.e. the distance beyond which habitat quality  
267 is no longer perceived by click beetles) to 10 m. This estimated value mainly stems from studies dealing  
268 with pheromone attractiveness for click beetles, since it is the only advection process for which literature  
269 provides some guidance. Sufyan et al. (2007) assessed the range of attractiveness of pheromone traps  
270 (for both species *A. lineatus* and *A. obscurus*) through a mark – recapture experiment. They recaptured  
271 > 60% and < 10% of beetles in pheromone traps placed at 2 and 60 m from their release point,  
272 respectively. In a later experiment, Sufyan et al. (2011) estimated the maximum distance of perception  
273 of a pheromone trap to be around 15-20 m. Blackshaw et al. (2018) studied the attractive range of traps  
274 baited with *A. obscurus* pheromone to male beetles in both still air and wind conditions. They found that  
275 the attraction distance was less than 5m in still air but increased with airflow.

276

277 *Aboveground population mortality ( $\mu_A$ )*

278 Adult life expectancy varies among studies. Brian (1947) observed that longevity of *A. obscurus* on  
279 various crops ranged from 69 to 84 days for females and from 45 to 75 days for males. Regarding *A.*

280 *ustulatus*, Hinkin (1983) reported an average longevity of 28 days for females and 19 days for males,  
281 while Furlan (1996) measured an adult lifespan of 24 (SD=5) days for females. More recently, Hicks  
282 and Blackshaw (2008) as well as Blackshaw et al. (2017) performed mark–release experiments on click  
283 beetles ; they showed respectively that adults (males and females combined) can survive for at least 45  
284 and 25 days. Accordingly, in our model, based on these elements and in accordance with our own field  
285 observations, we set the adult average lifespan to 25 days for both sexes.

286

### 287 *Oviposition rate ( $\pi$ )*

288 Most authors have studied oviposition under laboratory conditions. Brian (1947) found an average  
289 fecundity ranging from 30 to 150 (*A. obscurus*) depending on the crop (flax, potato and wheat were  
290 associated with low fecundity while grasslands of *Lolium*, *Festuca* and *Agrostis* were associated with  
291 high fecundity). Miles and Cohen (1941), Cohen (1942), and Furlan (1996), reported an average number  
292 of eggs laid per female of 109 in *A. obscurus* (min=78, max=186), 103 in *A. obscurus* and 78 in *A.*  
293 *ustulatus* (SD=27.9, min=52, max=140). Accordingly, we considered fecundity to be of 80 eggs per  
294 female over its lifespan (25 days as mentioned in the previous paragraph). Considering that sex ratio  
295 tends towards 1:1 (Blackshaw et al., 2017; Furlan, 1996), the oviposition rate was set to  $(80/2)/25=1.6$   
296 eggs per capita per day.

297

## 298 **2.3. Landscape contexts**

299 We defined four landscape contexts (Fig. 2 and Fig. 3) to investigate the processes driving the  
300 colonisation of vulnerable habitats (arable crops) characterised by a habitat quality lower than the one  
301 associated with grasslands (favourable habitats). The first three contexts are shown in Fig.2: a  
302 homogeneous cropped field cultivated over 2 years (Fig. 2A), a grassland in the field history (Fig. 2B)  
303 or in the neighbourhood of the cropped field (Fig. 2C). Finally, Fig. 3 describes the fourth context where  
304 we compared two dynamic landscapes  $\Omega_1$  and  $\Omega_2$  exhibiting the same duration of land use over time  
305 (landscape composition) but through contrasted spatial configurations.

306

## 307 **3. Results**

308

### 309 **3.1. Population dynamics in a homogeneous landscape context**

310 Fig. 4 shows the dynamics of above- and belowground populations in the simple case of a homogeneous  
311 landscape (Fig. 2A) comprising vulnerable cropped fields only. Simulation starts in winter with a null  
312 aboveground population and a larval density at equilibrium, hence 120 individuals per square meter.

313 Until the emergence of adults, the only process at stake is larval mortality, which reduces the  
314 belowground population to about 90 individuals per square meter. Adults emerge in early spring  
315 producing an increase in the aboveground population density with a peak value determined by (i) the

316 density of mature larvae at the onset of the emergence period, (ii) climate and (iii) adult life expectancy.  
317 It is followed by a decrease resulting from the constant mortality rate  $\mu_A$ . About 65 days after the last  
318 emergence, the aboveground population becomes extinct. This pattern repeats over the years.  
319 After adults have emerged, they disperse and lay eggs, thus refilling the belowground population  
320 compartment that increases, reaches a maximum at the end of the oviposition period, and then decreases  
321 due to larval mortality. After a few years a stationary pattern occurs.

322

### 323 **3.2. Wireworm dynamics with grassland in field history**

324 The effect of a grassland in field history (Fig. 2B) on the wireworm population density is illustrated on  
325 Fig. 5 for two pest life cycle durations: 2 years (short life cycle) and 4 years (long life cycle). Both  
326 dynamics exhibit the periodic behaviour described in the previous section (§3.1). At time 0, the field is  
327 converted to grassland, which translates into an important rise in habitat quality resulting in a subsequent  
328 wireworm population increase.

329 In Fig. 5, we compare the dynamics of a short life cycle population and a long life cycle population. The  
330 amplitudes of intra-annual oscillations are more important for the former. On one hand, the proportion  
331 of the belowground population that emerges as adults is larger in the case of a short life cycle (about  
332 half the population emerges each year) than in that of a long life cycle (about a quarter emerges each  
333 year). Given that oviposition rate is independent from life-cycle duration in our model, the increase in  
334 wireworm density is greater for short cycle species. On the other hand, once the aboveground population  
335 has vanished, high larval density results in a strong mortality until the next laying period.

336 In summary, short life cycle wireworms react more promptly to changes in habitat quality, thereby  
337 leading to higher infestation levels in temporary grasslands. Conversely, after grassland conversion into  
338 crop, wireworm density reaches more quickly the new current carrying capacity  $K_C$  in the case of short  
339 cycle species. Note that these conclusions stem from our model hypotheses, where a difference in life  
340 cycle duration results in different mortality rates (land-use carrying capacities and oviposition rates  
341 being unchanged).

342

### 343 **3.3. Wireworm dynamics in a field with a newly set-up grassland in its neighbourhood**

344 We simulated the effect of the conversion of half of a crop field into permanent grassland (Fig. 2C). Fig.  
345 6 shows the wireworm density along a direction orthogonal to the grassland/crop field border (40 meters  
346 on both sides), over a period of 10 years, and for a range of values of the diffusion coefficient  $D$ , the  
347 maximal advection speed  $V_{max}$  and the perceptual range  $\gamma$ . Diffusion tends to homogenise wireworm  
348 densities in space, while advection entails population movements from low towards high carrying  
349 capacity places.

350 In Fig. 6A, only diffusion applies. Wireworm densities increase from  $K_C$  (120 ind/m<sup>2</sup>) towards  $K_G$  (2000  
351 ind/m<sup>2</sup>) along the 10 years on the side covered with grassland. Adult click beetle diffusion results in an

352 increase in wireworm density in the adjacent cropped field with an intensity that vanishes away from  
353 the ecotone.

354 In the next six panels (Figs. 6B-G), diffusion and advection have antagonistic effects. Advection fosters  
355 population displacements from low towards high quality habitat, whereas diffusion promotes a  
356 homogenisation of the densities.

357 During the transient period before reaching equilibrium density in the grassland, the set-up of a grassland  
358 can have opposite effects on wireworm density in crop, depending on the advection strength (Figs. 6B-  
359 D). Indeed, a high maximal advection speed  $V_{max}$  results in a reduction of wireworm density inside the  
360 cropped field due to advection effect exceeding diffusion effect. For low maximal advection speed, the  
361 opposite occurs. Thus, 5 years after grassland set-up, wireworm density can be greater (Fig. 6B), similar  
362 (Fig. 6C) or lower (Fig. 6D) than the initial density in the crop. Changes in perceptual range have similar  
363 effects than changes in maximal advection speed (Figs. 6B-G).

364 A mechanistic effect of the population redistribution must be noted here. Unlike advection, diffusion is  
365 an inherently density-dependent process. As population increases within the grassland, so does the  
366 diffusion intensity, mechanically. The neighbourhood effect of grassland is therefore changing  
367 throughout the season, as diffusion takes more control over the redistribution.

368

### 369 **3.4. Effect of spatiotemporal distribution of grasslands on wireworm densities**

370 We considered two dynamic landscape contexts  $\Omega_1$  and  $\Omega_2$  spanning over four years (Fig. 3). Landscape  
371 composition over time (i.e. the proportions of cropped field and grassland over the duration of the virtual  
372 experiment) remains unchanged, while spatial configurations differ. The land cover within plot A is  
373 identical over the years in  $\Omega_1$  and  $\Omega_2$ , however the mean wireworm densities differ all along the four-  
374 year period. Final densities deviate by 10%. Indeed, in  $\Omega_1$ , wireworm density increases in plot B the first  
375 two years due to a favourable habitat (grassland). Subsequently, when plot A turns into grassland  
376 whereas plot B switches to cropped field, both advection and diffusion induce aboveground population  
377 movements from plot B to plot A, resulting in a meaningful wireworm increase once oviposition has  
378 been completed, that combines with the expansion of wireworm density due to the high carrying capacity  
379 associated with grassland. In  $\Omega_2$ , simultaneous evolution of densities in plots A and B do not favour any  
380 population flows.

381

## 382 **4. Discussion**

383

### 384 **4.1. Model relevance and novelty**

385 Insect pests inflict severe damage to agricultural crops, in spite of the intense use of agrochemical inputs.  
386 In fragmented and changing environment, as most agricultural landscapes are, the movement of insect  
387 pests has a strong influence on their abundance and spatiotemporal distribution, hence the extent of the

388 inflicted crop damage (Mazzi and Dorn, 2012). Thus, understanding dispersal mechanisms may help  
389 design effective pest management strategies. Here, we developed a spatially explicit mechanistic model  
390 to describe the population dynamics of insect pests inflicting crop damage while developing in soil, and  
391 spreading across the landscape as adults. We accounted for both aboveground and belowground  
392 population dynamics. To date, few studies have considered the link between both compartments though  
393 many species of ecological and economic importance have both above- and belowground life cycle  
394 stages (Benefer et al., 2012). Therefore, our developments could serve in investigations dealing with a  
395 wide variety of soil-dwelling insect pests.

396 Combined with spatial representations of dynamic landscape contexts, our population dynamics model  
397 enables the investigation of the spatial redistribution of populations over time given the spatial  
398 distribution of habitat quality and species life-history traits (dispersal ability, life cycle duration, etc.).  
399 We thus provide an appropriate framework to study how the arrangement of grassy landscape elements  
400 in space and time can mitigate crop infestation by soil-dwelling pests.

401

#### 402 **4.2. The dynamics of grassland as reservoirs or sources of pests**

403 In our results, we highlight (§3.2) the influence of the legacy of previous land uses on the current  
404 wireworm infestation level. This is well documented in plant sciences, e.g. regarding soil  
405 suppressiveness (Bailey et al., 2009; Postma et al., 2010), effects of biofumigation (Matthiessen and  
406 Kirkegaard, 2006), etc. Here we show that given its habitat favourability, the presence of a grassland in  
407 plot history entails the build-up of wireworm populations, and acts as a reservoir (Fig. 5). Interestingly,  
408 species life-history traits mitigate the changes in population density in response to land use change.  
409 Indeed, short life cycle species may be more responsive to land use change (Fig. 5).

410 At a broader spatial scale (landscape context, §3.3), the wireworm population response to the presence  
411 of a grassland in an adjacent field relies on the pest life-traits (dispersal ability, life cycle duration),  
412 habitat quality, and habitat patch dynamics. Local habitat quality governs the wireworm density in the  
413 field core, whereas density in borders results from the diffusion and advection of click beetles between  
414 habitats over time (Fig. 6). When advection is high, the inner edge of a low-quality habitat is temporary  
415 depleted at the expense of the adjacent higher quality habitat. However, diffusion (which tends to  
416 homogenise density) usually plays the opposite effect. Hence, wireworm density is ruled by the trade-  
417 off between these two antagonistic dispersal mechanisms. In a grassland where wireworm density has  
418 reached the local carrying capacity, diffusion at the ecotones will be high and sustain the colonisation  
419 of neighbouring crops. Therefore, the presence of a temporary versus permanent grassland in the  
420 neighbourhood may significantly change the outcome in terms of spatial distribution of wireworm  
421 populations. If advection is strong and grasslands highly favourable, temporary grasslands may act as  
422 pseudo-sinks, while permanent grasslands act as sources for click beetle populations. This clearly  
423 emphasizes the relevance of managing grassland regimes.

424

### 435 **4.3. Implications in terms of pest management strategies**

436 Our study outlines the complex link between pest colonisation patterns within the agricultural mosaic  
437 and the entanglement of habitats. The results presented in §3.4 (also Fig. 3 and Fig. 7) clearly support  
438 that the spatiotemporal arrangement of grasslands opens avenues for the management of soil-dwelling  
439 insect pests. As stated in previous studies, habitat manipulation can be successfully implemented to  
440 enhance the effectiveness of natural enemies (Landis et al., 2000), and mitigate the impacts of invasive  
441 arthropod pests (Jonsson et al., 2010). Yet here we emphasize the importance of the dynamics of land  
442 use spatial configurations. For example, click beetle spillover between a grassland and a neighbouring  
443 cropped field depends on the population replenishment in the favourable habitat: a recently installed  
444 grassland may act as a pseudo-sink, whereas a permanent grassland may act as a source. This finding  
445 evidences that depending on the spatiotemporal dynamics of habitat distribution, the role of a particular  
446 landscape element can switch from favourable to detrimental to pest population in fields. Consequently,  
447 pest management strategies relying on habitat spatial connectivity also call for the consideration of  
448 spatiotemporal continuity (or discontinuity).

439 Our modelling framework helps apprehending the effects of agroecological infrastructures (green veins,  
440 grassy strips, etc.) on pest communities. Grassy landscape elements are well known reservoirs for  
441 beneficial organisms (Blackshaw et al., 2017) but can also increase wireworm pressure. Our modelling  
442 approach is precious to identify the appropriate grassland management regime that provides the best  
443 trade-off to balance wireworm damage and a bundle of ecosystem services such as sustaining natural  
444 pest control, prevention against soil erosion, etc. (Hermann et al., 2013). Regarding the broad spatial  
445 and temporal scales related to the functioning of agroecosystems, frameworks enabling virtual  
446 experiments provide a relevant, if not unique, way to anticipate the complex trade-off between beneficial  
447 and detrimental effects associated with the adoption of agri-environmental schemes.

448

### 449 **4.4. Limitations and perspectives**

450 The reaction-diffusion-advection model we conceived to explore the main mechanisms at stake in the  
451 colonisation of vulnerable crops at the landscape scale is parsimonious. Model parameters were set  
452 based on a comprehensive review of literature dealing with the biology and the ecology of click beetles  
453 and their larvae. Nevertheless, knowledge gaps complicated the parameterisation step. For instance,  
454 knowledge on click beetle dispersal is still rudimentary, and though it has been evidenced that they can  
455 move across adjacent plots (Schallhart et al., 2009), some unpublished research suggest they may  
456 disperse on a much wider scale than currently mentioned (Lorenzo Furlan, personal communication).  
457 Dispersal patterns between uncultivated areas (grasslands) and cropped fields remain insufficiently  
458 documented. Moreover, in our approach we used the carrying capacity as a proxy for habitat quality.  
459 However, most experimental data we found date back to the first half of the twentieth century (Roebuck,  
460 1924; Roebuck et al., 1947; Ross et al., 1947; Salt and Hollick, 1949, 1944) and limit to grasslands.  
461 Their current accuracy can be questioned given changes in habitat properties and possibly in pest life-

462 history traits that have occurred since then. Further experiments to assess the carrying capacities  
463 associated with the main land uses occurring in the current agricultural landscapes are definitely  
464 required.

465 Furthermore, we did not explicitly consider crop management practices within our modelling  
466 framework, but rather treated them as an implicit component of the habitat quality associated with each  
467 land use. However, there is clear evidence that farming practices influence insect abundance (see for  
468 example Puech et al., 2014 and Kladivko, 2001). Soil cultivation directly affects the number and size of  
469 wireworms (Salt and Hollick, 1949). Interestingly, Furlan et al. (2020) demonstrated that an accurate  
470 ploughing timing of meadows, when rotation includes meadows, provides an efficient management  
471 tactic to prevent soil-pest damage to maize crops. Thus, combining virtuous cropping practices and a  
472 relevant strategy for the spatiotemporal arrangement of grasslands derived from our modelling  
473 framework may provide a holistic integrated pest management approach, as well as a credible alternative  
474 to chemical pesticides, in accordance with the principles of the European Directive 2009/128/EC on the  
475 sustainable use of pesticides (Barzman et al., 2015; Lorenzo Furlan et al., 2017).

476 Finally, in our study we examined various situations dealing with the presence of grassland in plot  
477 history (legacy effect) or neighbourhood (neighbourhood effect), as well as in dynamic landscape  
478 contexts (spatiotemporal effect). Such analytic landscape contexts allowed to address the role of  
479 grassland as sources of soil-dwelling pests, yet clearly do not reflect the real complexity of the  
480 agricultural mosaic, that stems from stakeholder decisions under agronomical, economical, socio-  
481 technical constraints. The representation of agroecosystems is still a research issue (Poggi et al., 2018b),  
482 and some model representations encompass more or less explicitly these constraints (Bareille et al.,  
483 2020; Martel et al., 2017; Ricci et al., 2018). Hence, a next research avenue consists in the exploration  
484 of suppressive patterns in simplified but realistic agricultural landscapes, generated under agronomic  
485 constraints at the farm or landscape scales.

486

## 487 **5. Conclusion**

488

489 Using mechanistic modelling, we illustrated how species life-history traits interact with spatiotemporal  
490 arrangement of habitats in agricultural landscape to shape pest abundances in vulnerable crops. Through  
491 neighbouring effect, legacy effect or their interaction, the distribution of grassy elements across the  
492 landscape has discernible effects on wireworm abundance in crops. We showed here that such endeavour  
493 is knowledge-hungry as it must be informed by some critical pest traits, but also a measurable alternative  
494 to a chemically driven crop protection. Our modelling framework allowed us to explore the ecosystem  
495 service of crop pest reduction that grasslands could provide through their ecological role on legacy effect  
496 and movements of pest populations (pseudo-sink vs source). Obviously, other ecosystem services  
497 provided by grasslands, such as biological control conservation, could be integrated in a common

498 framework offering exciting perspectives for the management of the trade-offs in a bundle of services  
499 and to identify situations or practices that allow switching from antagonism to synergy.

500

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509

510

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723 **Table 1:** List of variables and parameters.

Name	Description	Unit <sup>#</sup>	Nominal value	Range (see Suppl. Mat. 2)
$\tau$	Emergence rate of adults	d <sup>-1</sup>	†	-
D	Diffusion coefficient	m <sup>2</sup> /d	72	-
$\pi$	Oviposition rate	d <sup>-1</sup>	1.6	[1.2, 3.5]
c	Advection coefficient along the maturity dimension	y <sup>-1</sup>	0.25	[1/6, 1/2]
$\mu_A$	Aboveground mortality rate	d <sup>-1</sup>	0.04	[1/35, 1/15]
$\mu_B$	Belowground mortality rate	d <sup>-1</sup>	*	
$K_G$	Grassland carrying capacity	m <sup>-2</sup>	2000	[800, 3000]
$K_C$	Crop carrying capacity	m <sup>-2</sup>	120	
$\beta$	Exponent associated with belowground density-dependent mortality	-	0.2	[0.05, 0.4]
$\gamma$	Perception radius	m	10	-
$V_{\max}$	Maximum advection speed	m/d	10	-

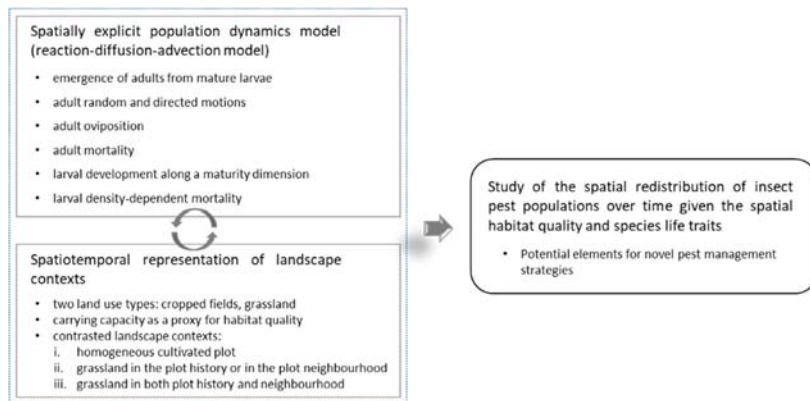
724 <sup>#</sup>d: day and y: year.

725 † See Supplementary Material 1.

726 \* Assessed using an optimisation procedure (see §2.2).

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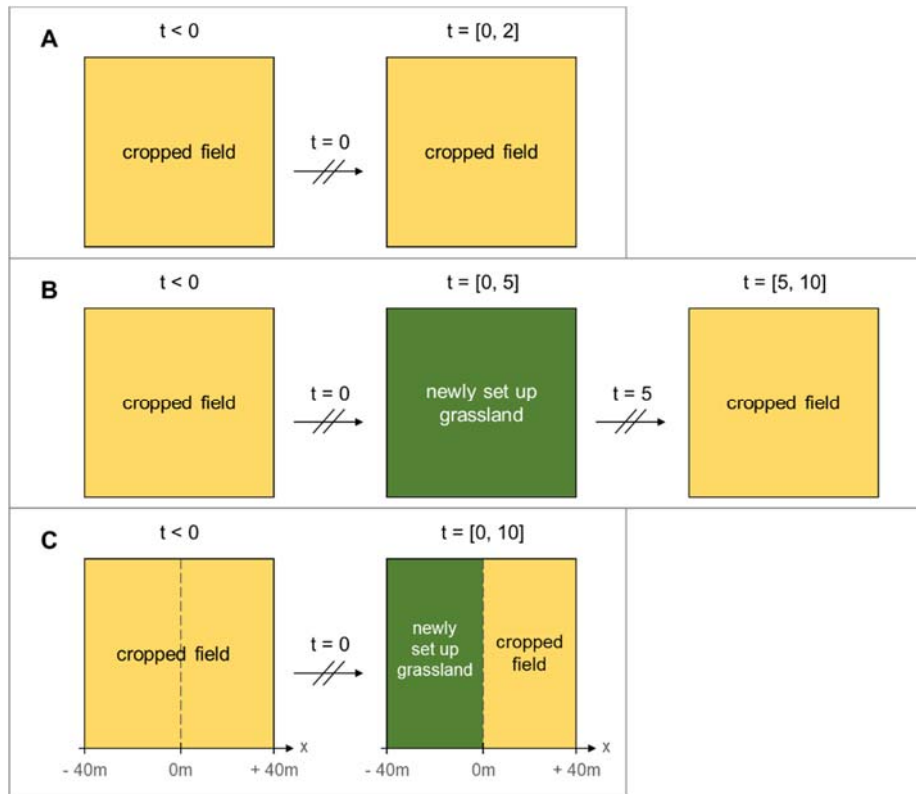
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730 **Fig. 1:** Conceptual diagram of the modelling framework.





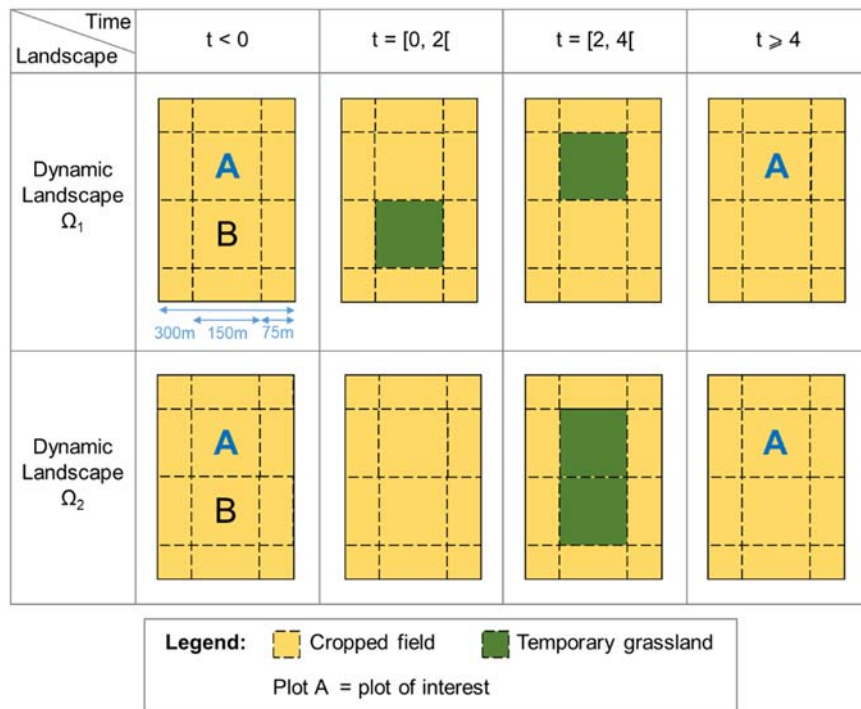
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733 **Fig. 2:** Three landscape contexts investigated along the study. Initial conditions correspond to stationary  
 734 populations within a long-term homogeneous cropped field; initial wireworm densities correspond to  
 735 the crop carrying capacity. (A) Homogeneous area studied over 2 years. (B) Five-year unmanaged area  
 736 (grassland) followed by a five-year cultivated area (cropped field). (C) Heterogeneous area resulting  
 737 from the conversion of a cropped field into two neighbouring plots, one cultivated and the other  
 738 unmanaged.

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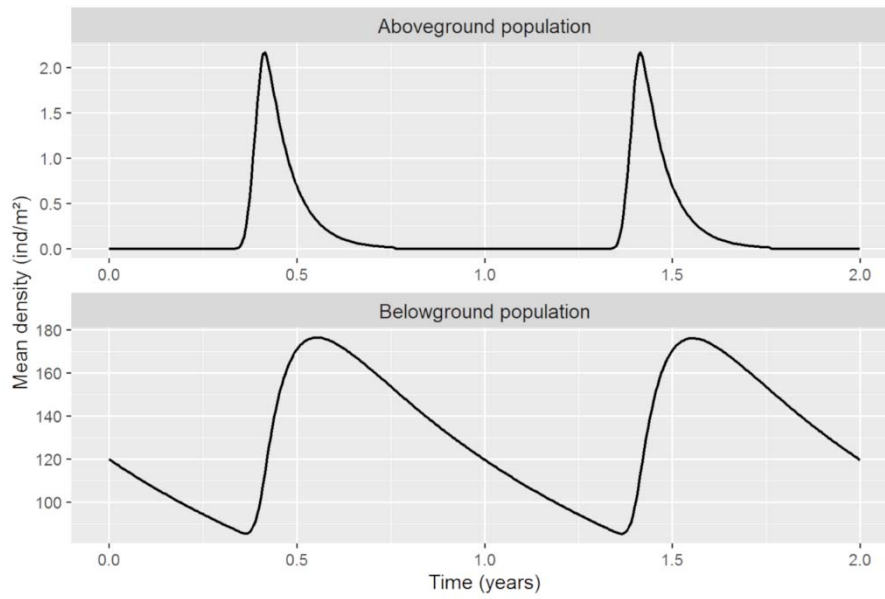
743 **Fig. 3:** Dynamic landscapes  $\Omega_1$  and  $\Omega_2$  exhibiting the same composition in terms of landscape features

744 over the period under study but contrasted spatial configurations.

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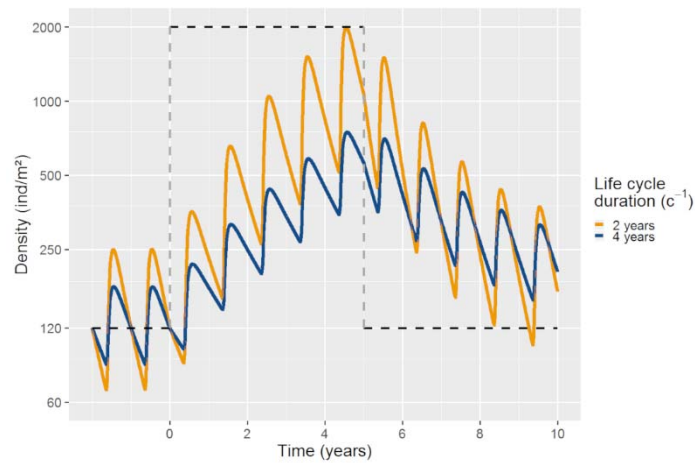
749 **Fig. 4:** Above- and belowground population dynamics in a homogeneous landscape context (as depicted  
750 in Fig. 2A). Simulations obtained by setting model parameters to their nominal values reported in Table  
751 1.

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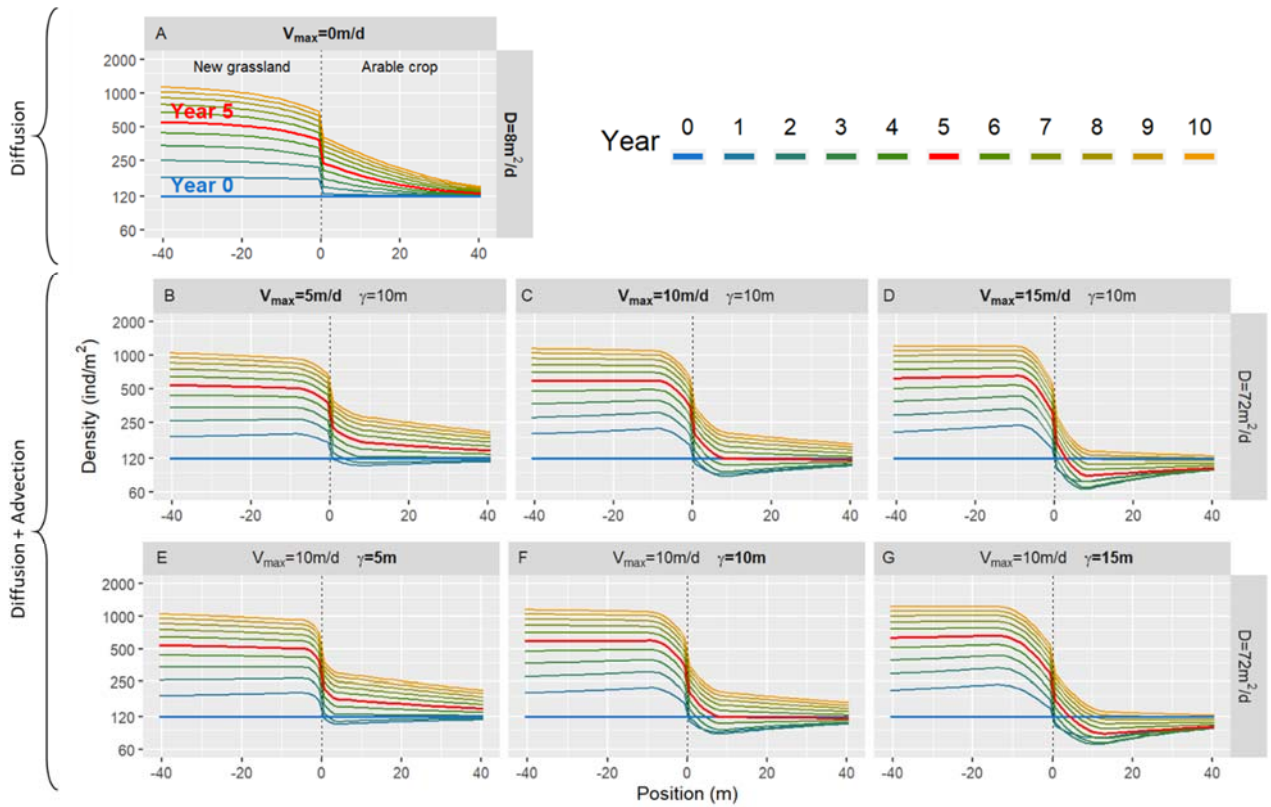
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757 **Fig. 5:** Wireworm population dynamics in a landscape with grassland in the plot history (as depicted in  
758 Fig. 2B). The orange and blue lines are associated with two different values of the life cycle duration (2  
759 and 4 years respectively); other model parameters set to their nominal values. The dashed step curve  
760 indicates current values of the carrying capacity ( $K_G=2000$  ind/m<sup>2</sup> and  $K_C=120$  ind/m<sup>2</sup>).

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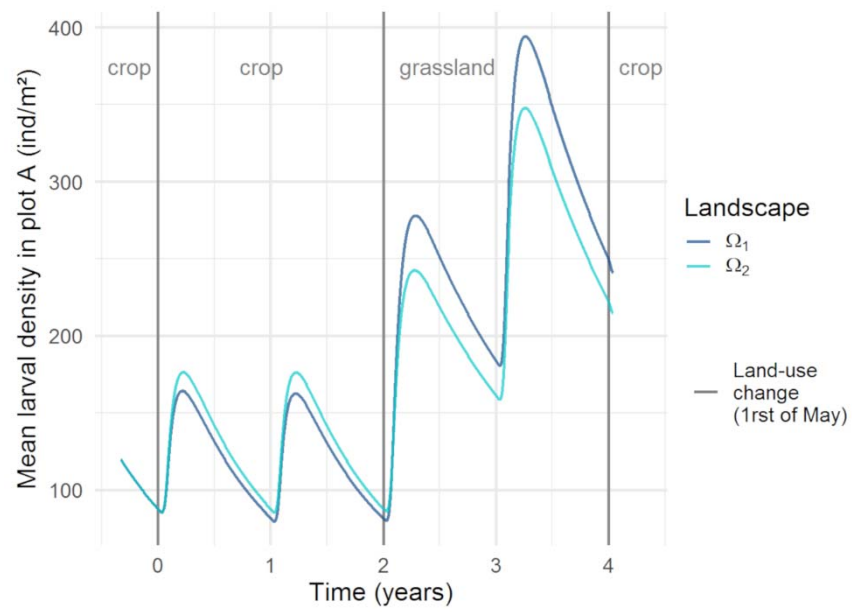
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**Fig. 6:** Wireworm density along the spatial dimension orthogonal to the boundary between the cropped field (right side) and the newly set up grassland (left side) (40 meters on both sides of the ecotone), as depicted in Fig. 2C), over a period of 10 years, and for different values of the diffusion coefficient  $D$ , the maximal advection speed  $V_{max}$  and the perception radius  $\gamma$ . Values assigned to carrying capacities are  $K_G=2000$  ind/m<sup>2</sup> and  $K_C=120$  ind/m<sup>2</sup> for the grassland and the cropped field respectively.



773

774 **Fig. 7:** Mean wireworm densities in plot A of dynamic landscape contexts  $\Omega_1$  and  $\Omega_2$  (as depicted in  
 775 Fig. 3) over four years. Simulations obtained by setting model parameters to their nominal values  
 776 reported in Table 1.

777 **Appendix 1: Emergence rate of click beetles**

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779 In our study, we modelled adult emergence as a continuous process spanning over a period governed by  
 780 the cumulative degree-days. We formalised the emergence rate as a temperature-dependent functional  
 781 response. Let's note

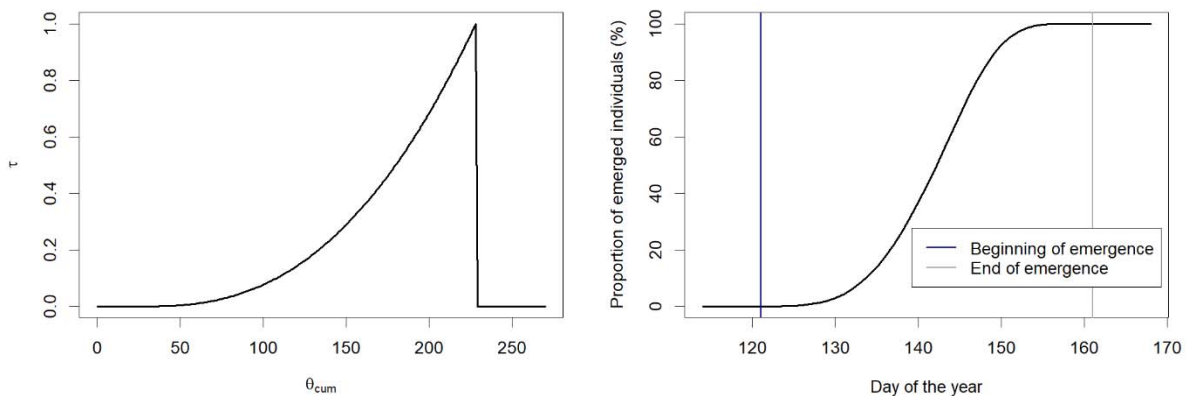
$$782 \quad \theta_{cum}(t) = \sum_{\substack{j=0 \\ \theta(j) > \theta_{base}}}^t (\theta(j) - \theta_{base}) \quad (A1.1)$$

783 where  $\theta_{cum}$  stands for the cumulated number of degree-days starting arbitrarily in January,  $\theta_{base}$  is the  
 784 temperature below which degree-days are not accounted for, set to 10°C in accordance with findings  
 785 from Furlan (2004, 1998) who showed that no larval development happened below this threshold value.  
 786 Denoting  $\theta_{start}$  (resp.  $\theta_{end}$ ) the minimal (resp. maximal) number of cumulated degree-days above (resp.  
 787 below) which emergence starts (resp. ends with the entire pool of mature larvae having emerged), the  
 788 emergence rate writes

$$789 \quad \tau(t) = \begin{cases} 0 & \text{if } \theta_{cum}(t) < \theta_{start} \\ \left( \frac{\theta_{cum}(t) - \theta_{start}}{\theta_{end} - \theta_{start}} \right)^\alpha & \text{if } \theta_{start} \leq \theta_{cum}(t) \leq \theta_{end} \\ 0 & \text{if } \theta_{cum}(t) > \theta_{end} \end{cases} \quad (A1.2)$$

790 with  $\alpha$  set empirically to 2.5 to ensure that the peak of emergence occurs in the middle of the emergence  
 791 window. As the effect of temperature was not in the scope of this study, we averaged the daily  
 792 temperatures in Le Rheu (Brittany, France) over a 10-year period (2008 to 2017), and we chose the  
 793 values of  $\theta_{start}$  and  $\theta_{end}$  that corresponded to late April and beginning of June (April 23<sup>rd</sup> and June 8<sup>th</sup>),  
 794 in accordance with published studies (Landl, 2010; Parker & Howard, 2001; Sufyan, 2007; Villeneuve  
 795 & Latour, 2011). The functional response for the emergence rate and the proportion of emerged adults  
 796 along time (applied to our temperature dataset) are illustrated on Fig. A1.

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798

799 **Fig. A1:** (A) Functional response for the emergence rate: emergence rate vs cumulative degree-days.  
800 (B) Proportion of emerged adults along time.

801

## 802 **References**

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## 813 Appendix 2: Sensitivity analysis

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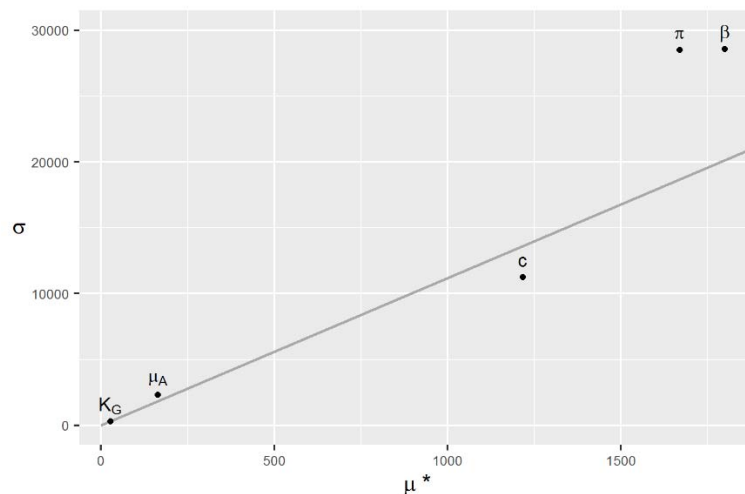
815 We carried out a Morris sensitivity analysis (Morris, 1991) to study the effect of model parameters on  
816 the speed at which the wireworm population density converges toward the carrying capacity of a given  
817 land cover. For this purpose, we considered the situation where a cultivated crop (with carrying capacity  
818  $K_C$ ) is converted to grassland (with carrying capacity  $K_G$ ). Initially, the wireworm density equals the  
819 crop carrying capacity  $K_C$ . We applied our population dynamics model (§2.1) and defined as an output  
820 metric the time required for the wireworm density to reach the mean density between  $K_C$  and  $K_G$ .

821 Among the model parameters, five were suspected of influencing the process of interest: the oviposition  
822 rate ( $\pi$ ), the advection coefficient along the maturity dimension ( $c$ ), the aboveground mortality rate ( $\mu_A$ ),  
823 the grassland carrying capacity ( $K_G$ ), and the exponent associated with the belowground density-  
824 dependent mortality ( $\beta$ ). Ranges of value assigned to these parameters are reported in Table 1 (last  
825 column). Other parameters were set to their nominal value (Table 1), except  $\mu_B$  that was assessed using  
826 a preliminary optimisation procedure as mentioned in §2.2.

827 Figure A2 shows the standard deviation ( $\sigma$ ) versus the absolute mean ( $\mu^*$ ) of the elementary effects  
828 associated with each of the five parameters investigated. Parameters  $K_G$  and  $\mu_A$  have a very low overall  
829 influence on the output metric. On the contrary, the parameter  $c$  has a significant linear influence, while  
830 parameters  $\pi$  and  $\beta$  exhibit a non-linear influence, possibly with interactions, on the output metric.

831 Thus, despite scarce knowledge on this parameter  $\beta$  (driving the strength of the legacy effect), it has a  
832 significant effect on the output metric we investigated, and its estimation would probably deserve further  
833 attention.

834



835

836 **Fig. A2:** Standard deviation ( $\sigma$ ) vs. absolute means ( $\mu^*$ ) of elementary effects obtained from the Morris  
837 method. Elementary effects of five parameters on the time required to reach the mean density between  
838  $K_C$  and  $K_G$ . The oblique grey line corresponds to  $\mu^* = 2SEM$  (Standard Error of the Mean  $SEM = \sigma/\sqrt{r}$ ,  
839 number of trajectories  $r = 200$ ).



840

841 **Reference**

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