

# 1 **Sharing the burden: Cabbage stem flea beetle pest pressure and crop** 2 **damage are lower in rapeseed fields surrounded by other rapeseed crops**

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## 11 **Abstract**

12 The cabbage stem flea beetle (*Psylliodes chrysocephala*) is a significant pest of rapeseed  
13 (*Brassica napus*). Feeding by adult *P. chrysocephala* can cause severe leaf damage and larval  
14 infestation can reduce stem strength, both of which impact crop growth and development,  
15 causing substantial yield losses and economic damage. The structure of the agricultural  
16 landscape can regulate herbivorous pest populations through top-down and bottom-up  
17 processes. This has shown promise in regulating the populations of other herbivorous pests,  
18 but remains relatively unexplored for *P. chrysocephala*. Here we investigate how the structure  
19 of the agricultural landscape influences *P. chrysocephala* abundance (pest pressure) and  
20 associated crop damage. We also examine the effect of the landscape on natural enemies  
21 and their ability to regulate *P. chrysocephala* populations. We show that *P. chrysocephala*  
22 populations are primarily regulated through bottom-up processes. We identify adjacency to  
23 another rapeseed crop and the total proportion of rapeseed grown in the landscape as key  
24 factors influencing beetle pressure, crop damage, and larval infestation, but find no effect of  
25 host crop proportions grown in the previous year at the examined scales up to 1 km  
26 surrounding focal crops. We also observe positive effects of crop heterogeneity and semi-  
27 natural habitat proportions on natural enemy abundance and diversity; however, these  
28 increases had no direct impact on *P. chrysocephala*. Bottom-up processes appear to  
29 contribute to herbivorous pest regulation by diluting beetles in the landscape, and could  
30 represent an important mechanism for sustainably managing pest populations by adapting the  
31 proportions and neighbourhoods of rapeseed crops at small to large spatial scales.

## 32 **Keywords**

33 Herbivorous pest; Pest suppression; Pest regulation; *Psylliodes chrysocephala*.

## 34 1. Introduction

35 Rapeseed, *Brassica napus*, is one of the most widely cultivated oilseed crops (Carré and  
36 Pouzet, 2014). Rapeseed confers myriad benefits to agricultural systems; for example,  
37 rapeseed acts as an economically attractive break crop in crop rotations and supports a  
38 diverse ecological habitat (Alford, 2000; Forleo et al., 2018; Kirkegaard et al., 1993). Rapeseed  
39 can be attacked by a range of herbivorous pests that inflict crop damage (Alford et al., 2003;  
40 Edde, 2021; Zheng et al., 2020). Of these, the cabbage stem flea beetle, *Psylliodes*  
41 *chrysocephala*, is of significant importance (Edde, 2021; Tixeront et al., 2023), having been  
42 recently ranked as the top biotic threat facing European rapeseed production (Zheng et al.,  
43 2020). *P. chrysocephala* adults damage plants through leaf feeding between July and October,  
44 causing up to 25% yield loss (Conrad et al., 2021; Edde, 2021; Ferguson et al., 2003). The  
45 larvae also cause significant damage between October and February when they burrow into  
46 the plant stem, with high infestation delaying crop maturation with strong impacts on yields  
47 (Conrad et al., 2021; Edde, 2021; Ortega-Ramos et al., 2022b).

48 Until recently, *P. chrysocephala* were primarily controlled using neonicotinoid seed treatments  
49 (Ortega-Ramos et al., 2022a), but restrictions on neonicotinoids have increased *P.*  
50 *chrysocephala* risk for rapeseed growers (Ortega-Ramos et al., 2023). Currently, pyrethroids  
51 are the only alternative chemical control option. However, overuse of pyrethroids has led to  
52 the development of insecticide resistance (Højland et al., 2016; Willis et al., 2020). In turn, this  
53 has increased the need for more sustainable non-chemical management practices (Ortega-  
54 Ramos et al., 2022a). In-field strategies for sustainable management include the integration  
55 of companion crops, sowing into straw mulch (Seimandi-Corda et al., 2023), and intercropping  
56 with legumes (Breitenmoser et al., 2022). However, the effectiveness of these methods can  
57 vary seasonally (Breitenmoser et al., 2022; Seimandi-Corda et al., 2023).

58 Herbivorous pests can also be influenced by the composition and configuration of the  
59 agricultural landscape (Delaune et al., 2021; Haan et al., 2020; Marini et al., 2023; Martin et  
60 al., 2019; Veres et al., 2013; Zhang et al., 2020). Generally, the agricultural landscape can  
61 influence herbivorous pests through two mechanisms: top-down (e.g., by encouraging the  
62 presence, diversity, and activity of natural enemies of pests, including predators and  
63 parasitoids) and bottom-up (e.g., by manipulating the proportion of host crop habitat and  
64 overwintering sites of the pests; Han et al. (2022)). Limiting resource (host crop) availability  
65 can interrupt insect reproduction and migration cycles, which can have significant impacts on  
66 the populations of specialist herbivores.

67 Successful top-down suppression is dependent on the presence and effectiveness of natural  
68 enemies (Dainese et al., 2017), themselves reliant on the availability of habitats and other  
69 resources, including non-cropped or seminatural habitats (Martin et al., 2019). In order to  
70 encourage top-down suppression of herbivorous pests, knowledge of the natural enemy  
71 communities likely to suppress them is needed. Only a few natural enemies of *P.*  
72 *chrysocephala* have been described (Hoarau et al., 2022). The level of successful control  
73 varies greatly (Hoarau et al., 2022), with the Ichneumonid wasp, *Tersilochus tripartitus*,  
74 showing the greatest biocontrol potential (Alford, 2000). For predators, field observations have  
75 identified spatial associations between two Carabid beetles (*Trechus quadristriatus* and  
76 *Pterostichus madidus*) and *P. chrysocephala* larvae (Warner et al., 2003), indicating that some  
77 carabid species might prey on *P. chrysocephala*. It has been proposed that generalist natural  
78 enemies of other flea beetles might also prey on *P. chrysocephala* (Hoarau et al., 2022),

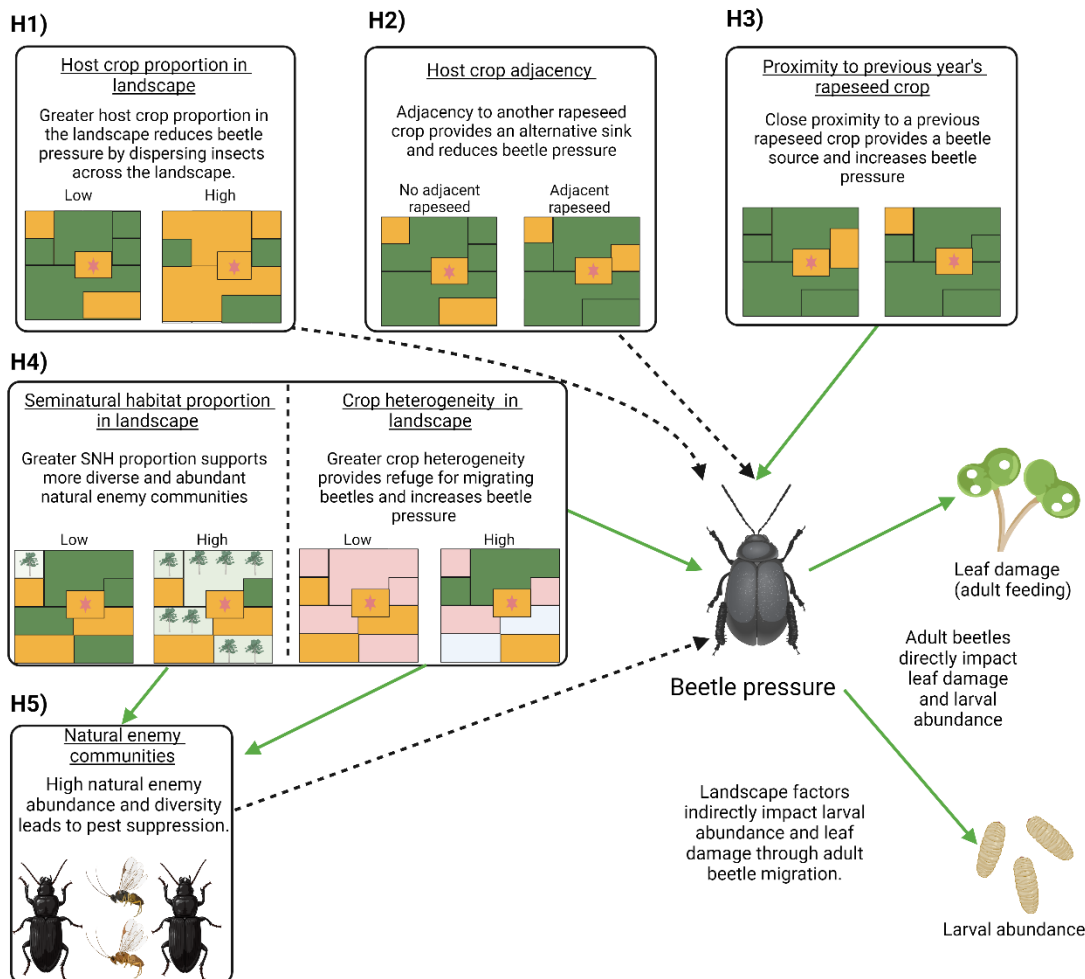
79 although there is no direct evidence of this. Generally, higher natural enemy abundance and  
80 diversity leads to increased pest suppression (Dainese et al., 2019; Dainese et al., 2017), and  
81 supporting natural enemy communities presents an avenue for sustainable management.

82 Bottom-up approaches involve manipulating habitat availability (proportion of host crops) or  
83 crop heterogeneity (the diversity of crops) in a landscape to decrease the density and  
84 population pressure of herbivorous pests by limiting resources (Almdal and Costamagna,  
85 2023; Boetzl et al., 2023). Bottom-up approaches have been described to contribute towards  
86 herbivorous pest regulation in similar crop-beetle systems (Boetzl et al., 2023; Zavalnitskaya  
87 et al., 2022), and could potentially be used for the sustainable management of *P.*  
88 *chrysocephala*. One variable that is relatively easy to manipulate is the proportion of host crop  
89 in the landscape; indeed, increased host crop proportion within the growing season has been  
90 shown to decrease the abundance of several other herbivorous pests of rapeseed, including  
91 pollen beetle, stem weevil, and brassica pod midge (Rusch et al., 2013; Thies and Tschardtke,  
92 1999; Zaller et al., 2008a; Zaller et al., 2008b). Furthermore, proximity to a host crop grown in  
93 the previous year can influence herbivorous pest pressure and crop damage, due to these  
94 fields acting as a source for insect populations in the following year (Boetzl et al., 2023; Sulg  
95 et al., 2023). Similarly, the heterogeneity of crops planted across the landscape in previous  
96 and current years can also influence herbivorous pest pressure and damage (Akter et al.,  
97 2023; Almdal and Costamagna, 2023; Scheiner and Martin, 2020; Veres et al., 2013).  
98 However, it is not known how these bottom-up processes impact *P. chrysocephala* pressure  
99 and associated crop damage.

100 These landscape effects are in-line with the landscape-moderated concentration and dilution  
101 hypothesis (Tschardtke et al., 2012). Higher host crop proportions in the previous compared  
102 to the current year could provide resources to support large herbivorous pest populations,  
103 which may immigrate in high numbers (concentration) into fields in the following year. This has  
104 recently been reported for *P. chrysocephala*, with fields within 0-2 km of a previous crop having  
105 a higher probability of experiencing damage than fields that are further away (Hausmann et  
106 al., 2023). In contrast, if host crop proportions are higher in the cropping year than the previous  
107 year, this may facilitate the dispersal of insects across the landscape and lead to lower  
108 densities (dilution) in individual fields (Zavalnitskaya et al., 2022). However, these effects tend  
109 to vary considerably between herbivore species (Boetzl *et al.* 2023) and are likely to depend  
110 strongly on the spatial scale of consideration. The concentration and dilution effect is thought  
111 to impact habitat specialists with a smaller host range, such as *P. chrysocephala* (Williams,  
112 2010), more than generalists (Tschardtke et al., 2012). As different aspects of agricultural  
113 landscapes contribute towards bottom-up and top-down insect control, land-use planning at  
114 the landscape scale could be used to develop sustainable insect management strategies  
115 (Lundin et al., 2021).

116 Here, we seek to elucidate how the agricultural landscape affects *P. chrysocephala* pressure  
117 and crop damage in rapeseed. To achieve this, we characterised the landscape surrounding  
118 14 rapeseed fields, monitored the local arthropod community in each field, and recorded *P.*  
119 *chrysocephala* pressure, leaf damage, and larval infestation. We hypothesize that (Fig. 1): 1)  
120 Host crop proportion will have a concentration-and-dilution effect on beetle pressure, with  
121 greater host crop proportions in the cropping year decreasing beetle pressure; 2) Adjacency  
122 to another rapeseed field in the cropping year will reduce beetle pressure by dispersing insects  
123 locally; 3) Fields in close proximity to previous year rapeseed fields will suffer from higher  
124 beetle pressure; 4) Greater proportions of seminatural habitat and crop heterogeneity provide

125 increased refugia and resources to support higher natural enemy abundance and diversity,  
 126 whereas greater crop heterogeneity provides more refuge habitat for migrating beetles; 5)  
 127 Higher natural enemy abundance and diversity will negatively affect beetle pressure and  
 128 associated crop damage via the provision of increased pest suppression services. Our work  
 129 provides insight into the top-down and bottom-up effects of the landscape on herbivorous  
 130 pests and highlights the potential applications for landscape-driven suppression of an  
 131 important herbivorous pest in an agriculturally relevant cropping system.



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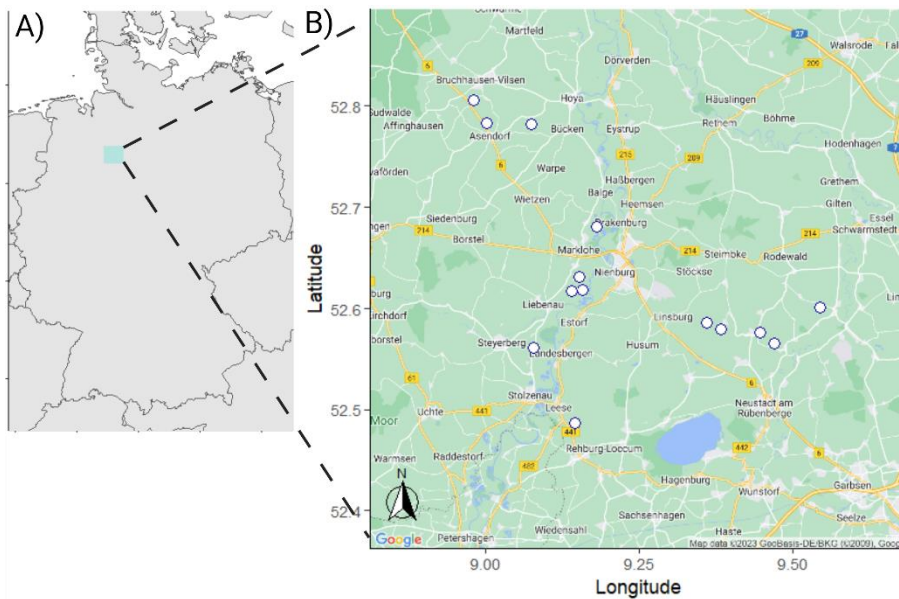
133 **Fig. 1:** Graphical representation of proposed interactions between *P. chrysocephala* and the landscape. Green  
 134 arrows (solid) indicate positive effects and black arrows (dashed) indicate negative effects. In the landscape  
 135 diagrams: Orange squares denote rapeseed fields; pale blue, pink, and green squares represent other crop types;  
 136 semi-natural habitats (SNH) are represented by pale green squares containing tress; red star indicates the  
 137 location of the focal field. Image was compiled in bioRender.

## 138 2. Materials & Methods

### 139 2.1 Site selection and landscape characterisation

140 Fourteen winter rapeseed fields were selected across a landscape gradient (proportion of  
 141 agricultural land in the landscape). Crops were sown in late August or early September 2021,  
 142 fields were drilled and managed by the host farmer (see Table S1 for agronomic information  
 143 for each field) and located in a key rapeseed production region in Lower Saxony, Germany  
 144 (Fig. 2). The landscape around each field was characterised at three spatial scales: 500 m,

145 750 m and 1 km radii. To ensure independence of the landscapes surrounding the sites, we  
 146 selected the sites to be a minimum distance of 1 km apart (the two closest sites were 1.1 km  
 147 apart). We also noted whether each field neighboured another winter rapeseed field.



148

149 **Fig. 2:** A) Location of the study region in Germany (blue shaded area) and B) location of the 14 rapeseed fields  
 150 (white circles). Maps were created in ggmap (v.3.0.2) with the base map used in B) obtained from Google Map  
 151 Services. Image was compiled in bioRender.

152 We characterised the agricultural landscapes using open-access digital crop-cover maps.  
 153 Detailed crop maps were obtained from the Lower Saxony federal database on agricultural  
 154 development (Servicezentrum Landentwicklung und Agrarförderung). These geodata contain  
 155 information on crop species grown in each field across Lower Saxony and are updated  
 156 annually. We calculated the proportion of rapeseed crops in each radius. We assessed the  
 157 compositional heterogeneity of crops (hereafter crop heterogeneity) by extracting the total  
 158 number of fields for each crop species present in the landscape, and calculating Shannon's  
 159 Diversity Index. Table S2 details the number of crop types included at each spatial scale for  
 160 each year. We retrieved data on semi-natural habitat from the Lower Saxony ATKIS database  
 161 (ATKIS-Objektartenkatalog), this database comprises a digital land-use cover map and we  
 162 calculated seminatural habitat by summing the total proportion for forest, woodland, heath,  
 163 moor, swamp, and uncultivated land.

164 We considered the following landscape variables at each spatial scale surrounding focal  
 165 rapeseed fields: The proportion of rapeseed fields planted in the previous cropping season,  
 166 the proportion of rapeseed fields planted in the cropping season, crop heterogeneity in the  
 167 previous cropping season, crop heterogeneity in the cropping season, the change in rapeseed  
 168 growing area between years, and the proportion of seminatural habitat in the cropping year.  
 169 We carried out landscape characterisation using QGIS v.3.24.3 and tested for correlations  
 170 between each landscape variable (Fig. S1-S3). Change in rapeseed proportion and rapeseed  
 171 proportion in the previous year were colinear at all spatial scales (Fig. S1-3; cor 0.794 – 0.863)  
 172 so we discarded change in rapeseed proportion from downstream analysis. We used Moran's  
 173 I test to assess the spatial autocorrelation of the remaining landscape variables. Spatial  
 174 autocorrelation analysis indicates that only one variable (previous year crop heterogeneity at

175 the 500 m spatial scale) was spatially autocorrelated, no autocorrelation was detected for any  
176 other landscape variable (Table S3; Fig. S4-S6).

## 177 **2.2 Experimental design and field assessments**

178 Each field comprised two 100 m long transects with five 2 m<sup>2</sup> quadrats spaced equidistantly  
179 along the transect. Both transects ran parallel to the field edge, the first transect was 5 m from  
180 the field boundary and the second transect was 25 m from the field boundary. Insect traps  
181 were placed at the central quadrat along each transect.

### 182 2.2.1 Assessments and sampling

183 Fields were visited weekly for five weeks from w/c 27.09.2021 until w/c 25.10.2021, coinciding  
184 with the main period of *P. chrysocephala* migration (Conrad et al., 2021), and were used to  
185 determine pest pressure. Pest pressure was measured by installing a yellow pan trap at  
186 vegetation height in the central quadrat along each transect. Traps were exposed for seven  
187 days and the number of beetles caught was used to determine weekly pest pressure.

188 Three rounds of in-crop assessments were carried out in September (w/c 27.09.2021;  
189 calendar week 39), October (w/c 25.10.2021; calendar week 43), and November (w/c  
190 22.11.2021; calendar week 47). In-crop assessments comprised two adult feeding damage  
191 assessments (September and October), a larval abundance assessment (November), and  
192 invertebrate trapping (September, October, and November). Timing of these assessments  
193 corresponded with peak periods of *P. chrysocephala* migration (September), egg laying  
194 (October), and the early periods of larval infestation (October/November) (Conrad et al., 2021).

195 For the leaf damage assessments, the proportion of leaf-area eaten was scored for 20 random  
196 plants per quadrat in September and October. *P. chrysocephala* damage can be distinguished  
197 from other damage (e.g., slug damage) by the characteristic shot-holes produced during  
198 feeding. The larvae abundance assessment consisted of randomly selecting five plants per  
199 quadrat, the diameter of the stem of each plant was recorded and the plant was dissected.  
200 Any *P. chrysocephala* larvae present in each stem were grouped by instar stage and the total  
201 larvae number was recorded.

202 Invertebrate trapping involved installing a pair of invertebrate traps (a yellow pan trap and a  
203 pitfall trap) in the central quadrat of each transect. Each trap was 1/3 full of water with a few  
204 drops of detergent. Pan traps were exposed for 48 h and pitfall traps for one-week. Pan traps  
205 were uncovered and placed within the crop canopy, pitfall traps were installed flush with the  
206 soil surface and covered with a rain roof. Trap contents were collected in 70% ethanol, and  
207 stored in glass jars until analysis. The total abundance of invertebrates was recorded and  
208 individuals were grouped into Order. Family-level identification was carried out for the  
209 Coleoptera, Hymenoptera, and Aranea. We used Schaefer (2018) to support invertebrate  
210 identification. As relatively little is known about the natural enemies of *P. chrysocephala* we  
211 used a proxy measurement for natural enemies that comprised arthropod groups previously  
212 described to contain natural enemies of *P. chrysocephala* as well as more generalist predator  
213 groups: Braconidae, Ichneumonidae (Hymenoptera); Caribidae (Coleoptera); Dysderidea,  
214 Theridiidae, Linyphiidae, Thomisidae, Lycosidae, Opiliones (Aranea). We calculated natural  
215 enemy  $\alpha$ -diversity using Shannon's diversity metric.

216 We were not able to obtain yield data from the study sites as a high number of fields suffered  
217 from crop failure in the spring following infestation with cabbage root fly (*Delia radicum*).

218 However, we believe that our study provides important insight into the landscape drivers  
219 behind *P. chrysocephala* risk and subsequent crop damage. Leaf damage at the cotyledon  
220 stage can be a useful proxy for potential yield-reducing damage as correlations between flea-  
221 beetle damage and yield loss have been described in spring rapeseed (Lundin, 2020).

## 222 **2.3 Statistical analysis**

223 Data were analysed in R Studio v.2022.02.3 running R (Ihaka and Gentleman, 1996) v.4.20.  
224 The following additional packages were used for data pre-processing: tidyverse (v.1.3.1;  
225 Wickham et al., 2019); data analysis: ape (5.7-1; Paradis et al., 2019), car (v.3.1-0; Fox and  
226 Weisberg, 2018), DHARMA (v.0.4.6; Hartig and Lohse, 2017), glmmTMB (v.1.1.7; Brooks et  
227 al., 2017), vegan (v.2.6-2; Dixon, 2003), lme4 (v.1.1-34; Bates et al., 2014), MuMIn (v.1.47.1;  
228 Barton, 2009), piecewiseSEM (v.2.3.0; Lefcheck, 2016); data visualisation: ggplot2 (v.3.3.6;  
229 Wickham, 2016), ggpubr (v.0.4.0; Kassambara, 2020), GGally (v.2.1.2; Schloerke et al.,  
230 2021), ggmap (v.3.0.2; Kahle et al., 2013).

231 We analysed the response variables beetle pressure, larvae abundance, and natural enemy  
232 abundance using generalised linear mixed models with negative binomial distribution in R  
233 package glmmTMB (Brooks et al., 2017). The response variables leaf damage and Shannon  
234 diversity of natural enemies were modelled using linear mixed effects models in R package  
235 lme4 (Bates et al., 2014). In all models, we included field as a random factor in order to account  
236 for variation in agronomic practices (e.g., planting scheme, crop management etc.) and for  
237 multiple sampling within each field. We used a Variance Inflation Factor (VIF) cut-off value of  
238 five to define collinear variables (Zuur et al., 2009) and removed any explanatory variables  
239 that breached this threshold. All other non-collinear explanatory variables were retained in the  
240 final model and final models were tested for significance using analysis of deviance tests (Type  
241 II Wald  $X^2$  tests). The fitted-residual plots of the final models were assessed to check model  
242 suitability and conformance to model assumptions.

### 243 2.3.1 Data pooling and transformations

244 To avoid zero-inflation, leaf damage data were averaged at the quadrat level and the total  
245 number of larvae was summed at the quadrat level. Natural enemy abundances were summed  
246 and natural enemy family richness was pooled across the two trap types. We calculated two  
247 beetle pressure metrics: Beetle pressure at week 39 and cumulative beetle pressure (sum of  
248 all beetle numbers from week 39-43). This produced 56 observations for beetle pressure, 280  
249 for feeding damage, 140 for larvae abundance, and 84 for natural enemies. Leaf damage data  
250 were logit transformed to account for bound proportional data (Warton and Hui, 2011). To aid  
251 model convergence the proximity to previous year rapeseed was square-root transformed.  
252 These transformed data were used in all analyses described below.

### 253 2.3.2 Data analysis

254 To determine how beetle pressure influenced leaf damage and larvae abundance we modelled  
255 leaf damage observed during the first assessment round against beetle pressure in week 39,  
256 and leaf damage observed during the second assessment round and larvae abundance  
257 against cumulative beetle pressure.

258 To test the influence of neighbouring rapeseed crops, we modelled cumulative beetle  
259 pressure, leaf damage, and larvae abundance against adjacent rapeseed field, proximity to a  
260 rapeseed crop in the previous year, the total abundance of natural enemies, and  $\alpha$ -diversity

261 of natural enemies. Natural enemy abundance and diversity were not correlated ( $r = 0.136$ ;  $t$   
262  $= 1.24$ ;  $df = 82$ ;  $p = 0.217$ ).

263 We examined the influence of the landscape on beetle pressure, leaf damage, larvae  
264 abundance, natural enemy abundance, and natural enemy  $\alpha$ -diversity in three separate  
265 models, one for each spatial scale (500, 750, and 1,500 m radii). In all models, explanatory  
266 variables included the calculated landscape variables (detailed in Section 2.1 above). In  
267 models explaining beetle pressure, leaf damage, and natural enemy abundance and diversity,  
268 assessment round was included as an explanatory variable. Where multiple spatial scales  
269 were found to influence the response variable, we used Akaike Information Criterion values  
270 corrected for small sample sizes (AICc) to identify the spatial scale that best explained the  
271 observed variation. To achieve this, we considered models with lower AICc values as better  
272 predictors of the response variable (Burnham and Anderson, 2002; Moraga et al., 2019).  
273 When comparing AICc values, we used  $\Delta AICc > 2$  to indicate difference in model fit.

### 274 2.3.3 Piecewise structural equation modelling

275 We used piecewise structural equation modelling (Lefcheck, 2016) to further explore the direct  
276 and indirect effects of the landscape. For this, we built models that only contained variables  
277 identified as significant in our prior analyses; each model included field as a random effect.  
278 We extracted standardised coefficients and evaluated model fit using Fisher's C statistic.

## 279 **3. Results**

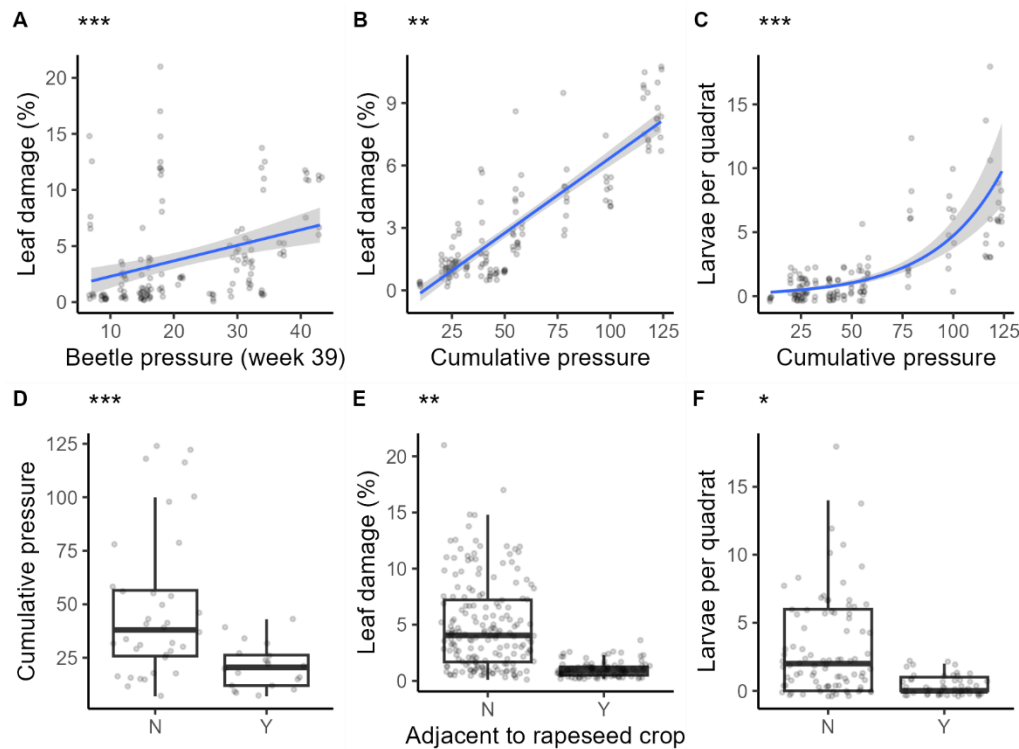
### 280 **3.1. Do *P. chrysocephala* pressure, rapeseed adjacency, and natural enemies** 281 **influence leaf damage and larval load?**

282 Adult beetle pressure observed at the start of the monitoring period was positively related to  
283 the leaf damage observed in the same week (calendar week 39;  $X^2_1 = 37.43$ ;  $p = <0.001$ ; Fig.  
284 3A; Table S4). We observed a similar positive relationship between cumulative beetle  
285 pressure and leaf damage during the second assessment round in late October (calendar  
286 week 43;  $X^2_1 = 7.79$ ;  $p = 0.005$ ; Fig. 3B; Table S4). Cumulative beetle pressure also positively  
287 influenced larvae abundance in November ( $X^2_1 = 52.81$ ;  $p = <0.001$ ; Fig. 3C; Table S4).  
288 Furthermore, the presence of a rapeseed field adjacent to the focal rapeseed crop significantly  
289 decreased beetle pressure ( $X^2_1 = 15.89$ ;  $p = <0.001$ ; Fig. 3D), leaf damage ( $X^2_1 = 7.65$ ;  $p =$   
290  $0.006$ ; Fig. 3E), and larvae abundance in the focal field ( $X^2_1 = 6.52$ ;  $p = 0.011$ ; Fig. 3F).  
291 Proximity to the nearest rapeseed crop grown in the previous year did not affect beetle  
292 pressure, leaf damage, or larvae abundance (Table S5).

293 In contrast, we did not detect any significant influence of the abundance and diversity of natural  
294 enemy communities on beetle pressure, leaf damage, or larvae abundance (Table S6).  
295 However, because little is known about antagonists of *P. chrysocephala* the natural enemy  
296 groups examined here (Braconid and Ichneumonid wasps, Carabid beetles, and spiders) can  
297 only be considered as potential enemies of *P. chrysocephala*.

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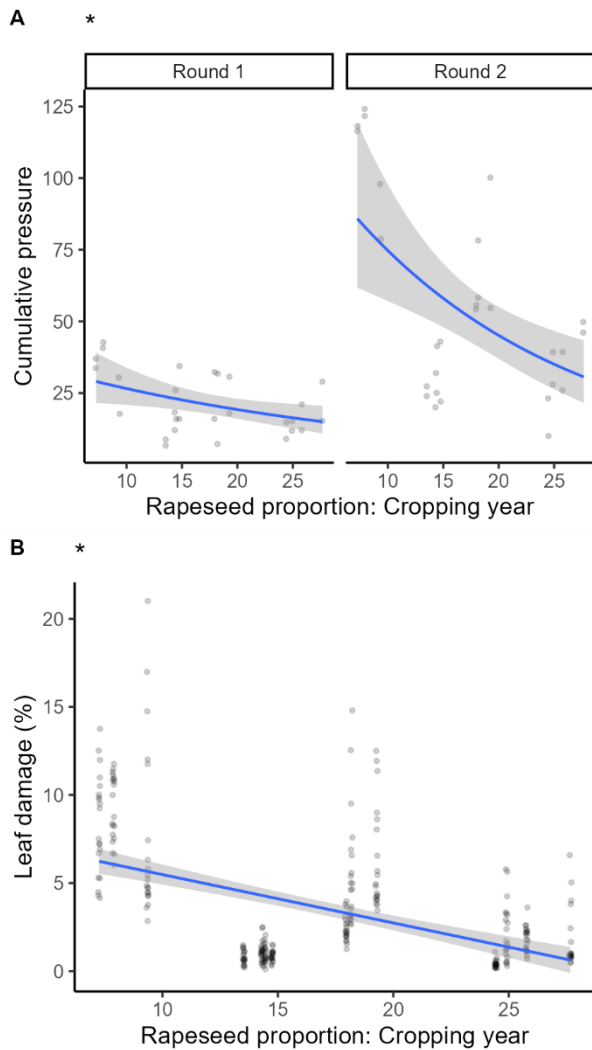


299

300 **Fig. 3:** A) The relationship between the beetle pressure observed in week 39 and the mean percent of leaf damage  
 301 observed during assessment round one; B) The relationship between cumulative beetle pressure over all five  
 302 weeks (week 39-43) of monitoring and the mean percent of leaf-area damage observed during assessment  
 303 round two; C) The relationship between cumulative beetle pressure and mean larvae abundance during assessment  
 304 round three. Blue lines show predicted values of the models. Shaded areas represent the confidence intervals. D-  
 305 F) The influence of an adjacent rapeseed crop on cumulative beetle pressure (D), mean percent of leaf damage  
 306 (E), and total larvae abundance (F). Underlying data are displayed as grey points. Asterisk denotes level of  
 307 significance for associated statistical tests: \* p < 0.05; \*\* p < 0.01; \*\*\* p < 0.001

### 308 **3.2 P. chrysocephala pressure and crop damage are affected by the proportion of winter** 309 **rapeseed**

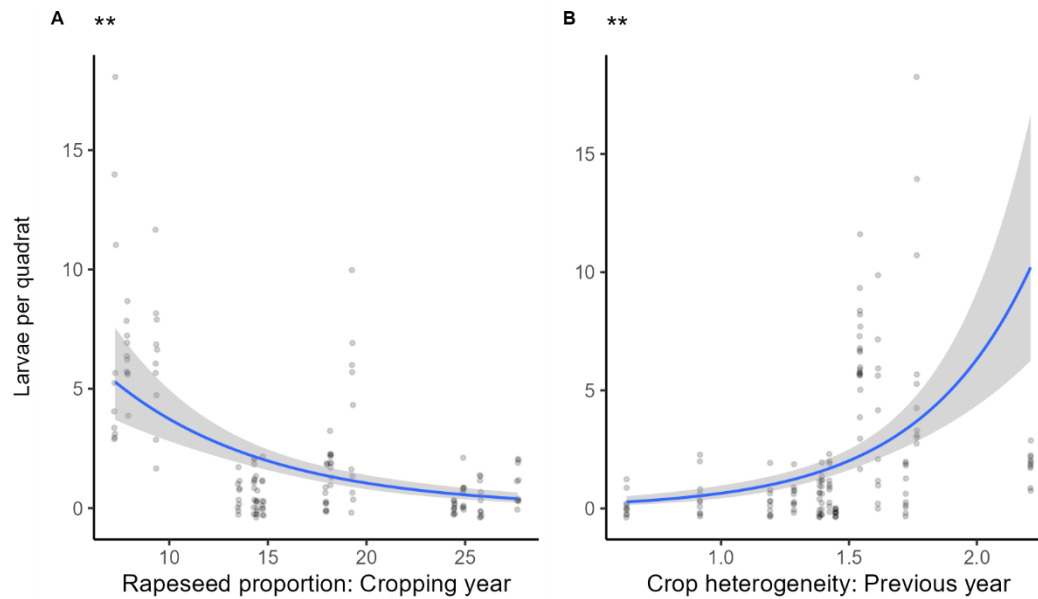
310 We examined how the agricultural landscape affects beetle pressure, leaf damage, and larvae  
 311 abundance at three spatial scales (500 m, 750 m, and 1 km radii). We observed significant  
 312 effects of the proportion of rapeseed in the cropping year on beetle pressure and leaf damage  
 313 at the 500 m radius (Fig. 4; Table S7; Table S8): Beetle abundance ( $X^2_1 = 4.66$ ;  $p = 0.031$ ;  
 314 Fig. 4A) and leaf damage ( $X^2_1 = 4.99$ ;  $p = 0.026$  Fig. 4B) were lower at sites with a higher  
 315 rapeseed proportion. No other tested landscape variable or spatial scale affected beetle  
 316 pressure or leaf damage.



317

318 **Fig. 4:** Relationship between cumulative beetle pressure, leaf damage, and the agricultural landscape at the 500  
 319 m spatial scale. Influence of rapeseed proportion in the cropping year on: A) Cumulative beetle pressure; B) Leaf  
 320 damage (%). Lines in panel A show the general linear regression and lines in panels B show the linear regression;  
 321 shaded areas represent the confidence intervals. Underlying data are displayed as grey points. Panel A shows  
 322 data separately for each assessment round. Asterisk denotes level of significance for associated statistical tests: \*  
 323  $p < 0.05$ ; \*\*  $p < 0.01$ ; \*\*\*  $p < 0.001$

324 Larvae abundance was influenced by host crop proportion in the cropping year and previous  
 325 year crop heterogeneity (Table S9). Fewer larvae were observed as the proportion of winter  
 326 rapeseed grown in the cropping year increased at a 500 m spatial scale ( $X^2_1 = 9.62$ ;  $p = 0.002$ ;  
 327 Fig. 5A) and higher larvae abundance was observed at sites with a greater previous year crop  
 328 heterogeneity ( $X^2_1 = 7.24$ ;  $p = 0.007$ ; Fig. 5B). No other landscape variable or spatial scale  
 329 influenced larvae abundance (Table S9).



330

331 **Fig. 5:** Relationship between larvae abundance and the agricultural landscape at the 500 m spatial scale. A)  
 332 Influence of rapeseed proportion in the cropping year; B) Crop heterogeneity in the previous year. Lines show the  
 333 general linear regression; shaded area represents the standard error. Underlying data are displayed as grey points.  
 334 Asterisk denotes level of significance for associated statistical tests: \*  $p < 0.05$ ; \*\*  $p < 0.01$ ; \*\*\*  $p < 0.001$

### 335 **3.3 Landscape effects on natural enemy communities**

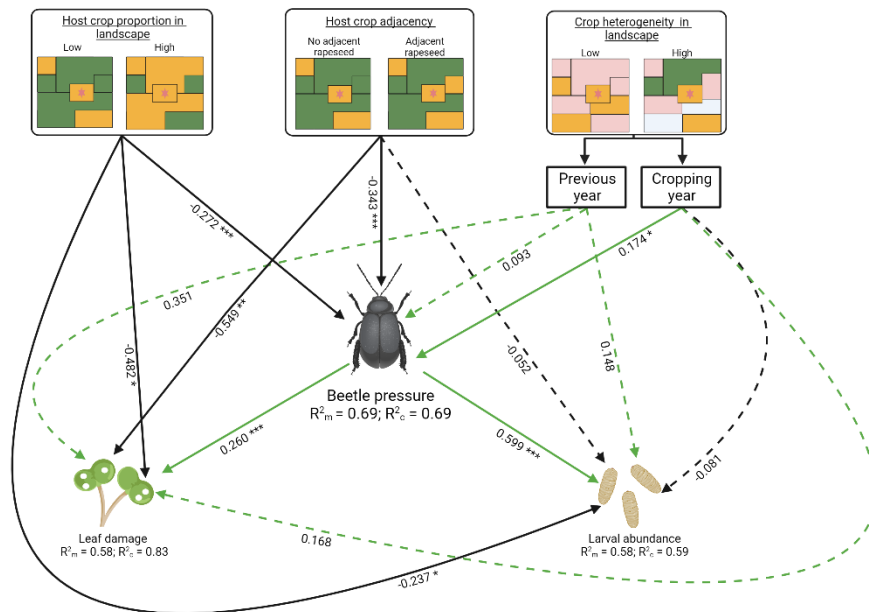
336 Natural enemy abundance was influenced by several landscape variables at all spatial scales.  
 337 Natural enemy abundance was higher in fields with a greater crop heterogeneity in the  
 338 previous ( $X^2_1 = 6.42$ ;  $p = 0.011$ ; Table S10; Fig. S7A) and cropping year ( $X^2_1 = 6.35$ ;  $p = 0.012$ ;  
 339 Table S10; Fig. S7B) at the 500 m spatial scale. Natural enemy abundance was also affected  
 340 by cropping year crop heterogeneity and the proportion of seminatural habitat in the landscape  
 341 at the 750 and 1,000m spatial scales (Table S10; Fig. S7). The  $\alpha$ -diversity of the natural enemy  
 342 communities increased as rapeseed proportion increased in the previous year at the 1,000 m  
 343 spatial scale (Table S11; Fig. S8).

### 344 **3.4 Structural equation modelling**

345 Beetle pressure influenced leaf damage and larvae abundance, and all three decreased when  
 346 the experimental field was adjacent to another rapeseed field (Fig. 3). Two landscape  
 347 parameters were identified as important drivers of beetle pressure, leaf damage, and larvae  
 348 abundance: An increasing proportion of rapeseed in the landscape in the cropping year  
 349 reduced beetle pressure, leaf damage, and larvae abundance (Fig. 4; 5); larvae abundance  
 350 was higher in fields with a greater crop heterogeneity in the previous year (Fig. 5). We  
 351 constructed a piecewise structural equation model to identify the potential cascading effects  
 352 between these bottom-up variables, focussing on the 500 m spatial scale (Fig. 6). Natural  
 353 enemy communities were also affected by several bottom-up landscape parameters (Fig. S7,  
 354 S8) but had no direct influence on beetle pressure or associated crop damage. For clarity,  
 355 natural enemies were not included in Fig. 6.

356 The piecewise structural equation model indicated that beetle pressure was strongly  
 357 influenced by the proportion of rapeseed grown in the cropping year and adjacency to another  
 358 rapeseed crop (Fig. 6), with plant damage and larvae abundance also influenced by these  
 359 landscape factors. A key determinant of leaf damage and larvae abundance was the direct

360 effect of beetle pressure. Crop heterogeneity influenced larval abundance; however, our  
 361 structural equation model suggests this is an indirect effect and that larval abundance is  
 362 primarily driven by beetle pressure. Fisher's C statistic provides an overall estimation of the fit  
 363 of our piecewise structural equation model ( $C = 9.14$ ;  $p = 0.691$ ).



364

365 **Fig. 6:** Piecewise structural equation model. Solid lines denote significant effects. Dashed lines indicate non-  
 366 significant relationships. Green lines show positive effects and black lines denote negative effects. The numbers  
 367 along the arrows are standardised path coefficients, and stars mark the significance level (\*  $< 0.05$ , \*\*  $< 0.01$ , \*\*\*  $<$   
 368  $0.001$ ). In the landscape diagrams: Orange squares denote repassed fields; pale blue, pink, and green squares  
 369 represent other crop types; red star indicates the focal field. Image was compiled in bioRender.

## 370 4. Discussion

371 In this paper we examined whether the landscape surrounding rapeseed fields influences in-  
 372 field abundance and crop damage caused by *P. chrysocephala*, a key herbivorous pest of  
 373 rapeseed. We find that regulation of *P. chrysocephala* populations at the landscape-level is  
 374 primarily driven through bottom-up processes, with little evidence of top-down suppression.  
 375 We show that host crop proportion in the landscape, particularly adjacency to another  
 376 rapeseed crop, is a significant bottom-up regulator of herbivorous pest pressure, crop damage,  
 377 and larval infestation. These bottom-up processes potentially contribute to herbivorous pest  
 378 regulation by diluting beetles in the landscape, thereby reducing herbivorous pest pressure  
 379 and limiting crop damage. We also find that crop heterogeneity in the landscape can influence  
 380 larval infestation in rapeseed plants, but this is potentially an indirect effect that is mediated  
 381 by overall beetle pressure. Although we find no direct link between natural enemy populations  
 382 and beetle pressure (i.e., no direct evidence of top-down regulation of herbivorous pests) we  
 383 observe bottom-up regulation of natural enemy communities by crop heterogeneity and  
 384 seminatural habitat at the landscape scale.

### 385 4.1 Greater host crop proportion and adjacency to another rapeseed crop reduces 386 herbivorous pest pressure and crop damage

387 A key finding of our study was that field sites with a higher proportion of rapeseed grown in  
 388 the same cropping season had lower *P. chrysocephala* pressure, reduced leaf damage, and  
 389 decreased larval abundance. Lower *P. chrysocephala* pressure suggests a dilution of beetles

390 across the host crops in the surrounding landscape. Similar effects on other herbivorous pests  
391 of rapeseed have been observed, including for pollen beetle, stem weevil, brassica pod midge,  
392 and other flea beetle species. This includes lower abundance of the asparagus beetle  
393 (*Crioceris asparagi*) in fields surrounded by a greater proportion of host crop over a 1 km  
394 spatial scale (Zavalnitskaya et al., 2022); Zaller et al., 2008b reported a similar relationship  
395 between host crop proportion and pollen beetle insect abundance across several spatial  
396 scales. Scheiner and Martin (2020) also observed a reduction in the abundance of leaf-  
397 chewing herbivorous pests (including other flea beetle species) in cabbage fields where a  
398 greater proportion of host crop was grown in the surrounding landscape, and Josso et al.  
399 (2013) described a decrease in the number of cabbage root fly eggs in fields surrounded by a  
400 higher proportion of Brassica crops over a 1 km circumference. The observed decrease in the  
401 abundance of *P. chrysocephala* (this study), pollen beetle and stem weevil (Zaller et al.,  
402 2008b), asparagus beetle (Zavalnitskaya et al., 2022), cabbage root fly (Josso et al., 2013),  
403 and general leaf-chewing insects (Scheiner and Martin, 2020) as the proportion of host crop  
404 grown in the surrounding landscape increases indicates a dilution effect of foliar-feeding  
405 insects across the landscape. This broadly follows the landscape-moderated dilution  
406 hypothesis (Tscharntke et al., 2012).

407 With regards to the relationship between increasing host crop proportion and lower leaf  
408 damage: Zaller et al. (2008a) observed a decrease in damage caused by pollen beetles (% of  
409 podless peduncles) and pod midge (% of premature pods) as the proportion of rapeseed in  
410 the surrounding landscape increased. This is in-line with our observations of reduced *P.*  
411 *chrysocephala* damage and lower larvae abundance at sites surrounded with a greater  
412 rapeseed proportion. Our structural equation modelling suggests that the observation of  
413 reduced leaf damage and lower larvae abundance is both an indirect bottom-up effect of the  
414 landscape and a direct effect of beetle pressure. We also observed an effect of crop  
415 heterogeneity on larval abundance, with larval infestation increasing at sites that had a higher  
416 crop heterogeneity in the previous season. The structural equation model suggests that larval  
417 abundance is primarily mediated by *P. chrysocephala* pressure, with limited direct effect of  
418 crop heterogeneity. Crop heterogeneity in the previous year has been reported to influence  
419 population densities of similar flea beetle species (*Phyllotreta undulata*) in spring rapeseed  
420 fields, with increasing crop heterogeneity over a 2,000 m radii influencing beetle abundance  
421 (Boetzel et al., 2023).

422 These observations indicate that the proportion of host crop habitat in the surrounding  
423 landscape is a key driver of herbivorous pest pressure and resulting crop damage in  
424 agricultural ecosystems. We also found that presence/absence of a neighbouring rapeseed  
425 crop was a key factor determining *P. chrysocephala* pressure and damage, with lower beetle  
426 abundance and less damage observed in rapeseed fields that were adjacent to another  
427 rapeseed crop. This is in-line with recent research indicating that adjacent habitat influences  
428 population dynamics of herbivorous pests, including the herbivorous pests of Brassica crops  
429 (Akter et al., 2023).

#### 430 **4.2 Proximity to previous year's rapeseed crop and influence of natural enemies**

431 We did not detect any influence of previous season host crop proportion on any *P.*  
432 *chrysocephala* parameter measured, indicating that a habitat sink is a more important factor  
433 determining *P. chrysocephala* dispersal across a landscape at the examined scales than a  
434 habitat source. However, a recent study examining the effects of landscape (previous crop

435 and seminatural habitat) on *P. chrysocephala* at larger spatial scales (10 km) indicated that  
436 increased rapeseed proportion in the previous year can influence the probability of larval  
437 infestation (Hausmann et al., 2023).

438 Proximity to previous year's host crop can also affect herbivorous pest pressure (Hausmann  
439 et al., 2023; Weisz et al., 1994; Zaller et al., 2008a) and crop damage, including for other flea  
440 beetle species (Boetzi et al., 2023) and for *P. chrysocephala* (Hausmann et al., 2023). We did  
441 not detect any influence of decreasing proximity to previous year's rapeseed crop. However,  
442 the majority of our fields were in relatively close proximity to a previous rapeseed field: 42%  
443 were directly adjacent to a previous rapeseed crop and 50% had a previous rapeseed field  
444 within 1 km, only one field was more than 1 km away from a previous rapeseed crop. Recent  
445 observations of lower *P. chrysocephala* abundance as proximity to previous year's rapeseed  
446 crop increased (Hausmann et al., 2023) examined this over a larger spatial scale (10 km).

447 We did not observe any direct impact of natural enemies on *P. chrysocephala* abundance or  
448 any indirect effect on leaf damage and larvae abundance. However, as only a few natural  
449 enemies of *P. chrysocephala* have been described (Hoarau et al., 2022) we used proxy  
450 measurements for natural enemy groups and may have overestimated natural enemy  
451 communities. Nonetheless, the abundance of these arthropod communities was influenced by  
452 bottom-up processes at several spatial scales, with influencing landscape factors including  
453 crop heterogeneity and the proportion of seminatural habitat. Increased natural enemy  
454 abundance at sites surrounded with more seminatural habitat follows previous observations  
455 and is likely due to the reliance of these insects on the availability of habitats, refugia, and  
456 resources (Martin et al., 2019). Landscapes with greater crop heterogeneity can provide  
457 greater resource availability for natural enemies and can thereby support more abundant  
458 insect populations, including populations of other herbivorous pests (Boetzi et al., 2023;  
459 Redlich et al., 2018). Therefore, the bottom-up processes that regulate natural enemy  
460 communities are likely associated with an increased presence and abundance of alternative  
461 prey species and increased non-crop habitats that support a greater abundance of non-  
462 agricultural insects.

## 463 **5. Conclusion**

464 Our results confirm emerging research on the importance of year-to-year patterns in host crop  
465 amounts in determining the abundance and damage of herbivorous pests in crops. For a key  
466 herbivorous pest responsible for major losses in winter rapeseed systems, we show that a  
467 high proportion of rapeseed (host crop) in the landscape can decrease herbivorous pest  
468 pressure and crop damage, and that adjacency to another rapeseed field can further  
469 contribute to diluting pest impacts. Moreover, we show that herbivorous pest pressure is  
470 mainly driven by bottom-up processes with no evidence of regulation by natural enemies.  
471 Manipulating the agricultural landscape to deliver sustainable herbivorous pest suppression is  
472 an approach that has been suggested for similar herbivorous pests. However, implementing  
473 these approaches is often difficult as planning at the landscape scale goes beyond the  
474 individual farm level and requires integration with multiple stakeholders in order to deliver the  
475 benefits (Lundin et al., 2021). Given strong evidence suggesting that *P. chrysocephala*  
476 populations, alongside other major herbivorous pests of rapeseed, are regulated by host crop  
477 proportions in the agricultural landscape, we recommend the development of collective,  
478 landscape-wide crop rotation strategies that enable sustainable management in these major  
479 crop systems while minimising the use of synthetic pesticides. Furthermore, given apparent

480 similarities of bottom-up regulation processes for a range of herbivorous pests in rapeseed  
481 and other crops, we suggest that exploring the interactive effects between multiple herbivorous  
482 pests in the same crop and landscapes would highlight important regulatory landscape-scale  
483 processes that could be exploited to achieve synergistic suppression of multiple pest species.  
484 Examining the top-down and bottom-up processes that influence herbivorous pest pressure  
485 across scales in other understudied organisms will help identify commonalities that can be  
486 used to regulate herbivorous pest populations more broadly.

487

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## 492 Ethical statement

493 No specific permissions applied to this study according to the national legislation.

## 494 CRediT authorship

495 Conceptualisation: DJL, EAM, AMCP. Methodology: DJL, EAM, AMCP. Formal analysis: DJL.  
496 Investigation: AMCP, PM, DJL. Data curation: AMCP, DJL. Writing – original draft: DJL.  
497 Writing – review and editing: EAM. Visualisation: DJL. Supervision: DJL, EAM. Funding  
498 acquisition: DJL, EAM.

## 499 Declaration of Competing Interests

500 The authors declare that they have no known competing financial interests or personal  
501 relationships that could have appeared to influence the work reported in this paper.

## 502 Data availability

503 Data and code are available open access via the University of Liverpool Research Data  
504 Catalogue (Leybourne et al., 2023): <https://datacat.liverpool.ac.uk/2511/> To protect privacy of  
505 the participating farmers we are unable to publicly share geodata associated with the project.  
506 This information can be requested by contacting the corresponding author.

## 507 References

- 508 **Akter, S., Rizvi, S. Z. M., Haque, A., Reynolds, O. L., Furlong, M. J., Melo, M. C., Osborne, T.,**  
509 **Mo, J., McDonald, S., Johnson, A. C. et al.** (2023). Continent-wide evidence that landscape context  
510 can mediate the effects of local habitats on in-field abundance of pests and natural enemies. *Ecology*  
511 *and Evolution* **13**, e9737.
- 512 **Alford, D. V.** (2000). Biological control of insect pests on oilseed rape in Europe. *Pesticide*  
513 *Outlook* **11**, 200-202.
- 514 **Alford, D. V., Nilsson, C. and Ulber, B.** (2003). Insect pests of oilseed rape crops. *Biocontrol*  
515 *of oilseed rape pests* **1**.
- 516 **Almdal, C. D. and Costamagna, A. C.** (2023). Crop diversity and edge density benefit pest  
517 suppression through bottom-up and top-down processes, respectively. *Agriculture, Ecosystems &*  
518 *Environment* **349**, 108447.
- 519 **Barton, K.** (2009). MuMIn: multi-model inference. [http://r-forge.r-](http://r-forge.r-project.org/projects/mumin/)  
520 [project.org/projects/mumin/](http://r-forge.r-project.org/projects/mumin/).
- 521 **Bates, D., Mächler, M., Bolker, B. and Walker, S.** (2014). Fitting linear mixed-effects models  
522 using lme4. *arXiv*, 1406.5823.
- 523 **Boetzel, F. A., Bommarco, R., Aguilera, G. and Lundin, O.** (2023). Spatiotemporal isolation of  
524 oilseed rape fields reduces insect pest pressure and crop damage. *Journal of Applied Ecology* **60**,  
525 1388-1398.
- 526 **Breitenmoser, S., Steinger, T., Baux, A. and Hiltbold, I.** (2022). Intercropping winter oilseed  
527 rape (*Brassica napus* L.) has the potential to lessen the impact of the insect pest complex. *Agronomy*  
528 **12**, 723.
- 529 **Brooks, M. E., Kristensen, K., Van Benthem, K. J., Magnusson, A., Berg, C. W., Nielsen, A.,**  
530 **Skaug, H. J., Machler, M. and Bolker, B. M.** (2017). glmmTMB balances speed and flexibility among  
531 packages for zero-inflated generalized linear mixed modeling. *The R Journal* **9**, 378-400.



- 532 **Burnham, K. P. and Anderson, D. R.** (2002). Model selection and multimodel inference: a  
 533 practical information-theoretic approach. USA: Springer New York.
- 534 **Carré, P. and Pouzet, A.** (2014). Rapeseed market, worldwide and in Europe. *OCL* **21**, D102.
- 535 **Conrad, N., Brandes, M., Ulber, B. and Heimbach, U.** (2021). Effect of immigration time and  
 536 beetle density on development of the cabbage stem flea beetle, (*Psylliodes chrysocephala* L.) and  
 537 damage potential in winter oilseed rape. *Journal of Plant Diseases and Protection* **128**, 1081-1090.
- 538 **Dainese, M., Martin, E. A., Aizen, M. A., Albrecht, M., Bartomeus, I., Bommarco, R.,  
 539 Carvalho, L. G., Chaplin-Kramer, R., Gagic, V., Garibaldi, L. A. et al.** (2019). A global synthesis  
 540 reveals biodiversity-mediated benefits for crop production. *Science Advances* **5**, eaax0121.
- 541 **Dainese, M., Schneider, G., Krauss, J. and Steffan-Dewenter, I.** (2017). Complementarity  
 542 among natural enemies enhances pest suppression. *Scientific Reports* **7**, 8172.
- 543 **Delaune, T., Ouattara, M. S., Ballot, R., Sausse, C., Felix, I., Maupas, F., Chen, M., Morison,  
 544 M., Makowski, D. and Barbu, C.** (2021). Landscape drivers of pests and pathogens abundance in  
 545 arable crops. *Ecography* **44**, 1429-1442.
- 546 **Dixon, P.** (2003). VEGAN, a package of R functions for community ecology. *Journal of*  
 547 *Vegetation Science* **14**, 927-930.
- 548 **Edde, P. A.** (2021). 3 - Arthropod pests of rapeseed (canola) (*Brassica napus* L.). In *Field Crop*  
 549 *Arthropod Pests of Economic Importance*, (ed. P. A. Edde), pp. 140-207. USA: Academic Press.
- 550 **Ferguson, A. W., Klukowski, Z., Walczak, B., Clark, S. J., Mugglestone, M. A., Perry, J. N. and**  
 551 **Williams, I. H.** (2003). Spatial distribution of pest insects in oilseed rape: implications for integrated  
 552 pest management. *Agriculture, Ecosystems & Environment* **95**, 509-521.
- 553 **Forleo, M. B., Palmieri, N., Suardi, A., Coaloa, D. and Pari, L.** (2018). The eco-efficiency of  
 554 rapeseed and sunflower cultivation in Italy. Joining environmental and economic assessment.  
 555 *Journal of Cleaner Production* **172**, 3138-3153.
- 556 **Fox, J. and Weisberg, S.** (2018). An R companion to applied regression. USA: SAGE  
 557 publications.
- 558 **Haan, N. L., Zhang, Y. and Landis, D. A.** (2020). Predicting landscape configuration effects on  
 559 agricultural pest suppression. *Trends in ecology & evolution* **35**, 175-186.
- 560 **Han, P., Lavoit, A.-V., Rodriguez-Saona, C. and Desneux, N.** (2022). Bottom-up forces in  
 561 agroecosystems and their potential impact on arthropod pest management. *Annual Review of*  
 562 *Entomology* **67**, 239-259.
- 563 **Hartig, F. and Lohse, L.** (2017). R package 'DHARMA'. [https://cran.r-](https://cran.r-project.org/web/packages/DHARMA/index.html)  
 564 [project.org/web/packages/DHARMA/index.html](https://cran.r-project.org/web/packages/DHARMA/index.html).
- 565 **Hausmann, J., Heimbach, U., Gabriel, D. and Brandes, M.** (2023). Effects of regional crop  
 566 rotations on autumn insect pests in winter oilseed rape. *Pest Management Science* **Early View**.
- 567 **Hoarau, C., Campbell, H., Prince, G., Chandler, D. and Pope, T.** (2022). Biological control  
 568 agents against the cabbage stem flea beetle in oilseed rape crops. *Biological Control* **167**, 104844.
- 569 **Højland, D. H., Nauen, R., Foster, S. P., Williamson, M. S. and Kristensen, M.** (2016).  
 570 Incidence, spread and mechanisms of pyrethroid resistance in European populations of the cabbage  
 571 stem flea beetle, *Psylliodes chrysocephala* L. (Coleoptera: Chrysomelidae). *PLOS ONE* **10**, e0146045.
- 572 **Ihaka, R. and Gentleman, R.** (1996). R: a language for data analysis and graphics. *Journal of*  
 573 *computational and graphical statistics* **5**, 299-314.
- 574 **Josso, C., Le Ralec, A., Raymond, L., Saulais, J., Baudry, J., Poinot, D. and Cortesero, A. M.**  
 575 (2013). Effects of field and landscape variables on crop colonization and biological control of the  
 576 cabbage root fly *Delia radicum*. *Landscape Ecology* **28**, 1697-1715.
- 577 **Kahle, D. J. and Wickham, H.** (2013). ggmap: spatial visualization with ggplot2. *The R Journal*  
 578 **5**, 144.
- 579 **Kassambara, A.** (2020). ggpubr: "ggplot2" based publication ready plots. *R package*.
- 580 **Kirkegaard, J. A., Angus, J. F., Gardner, P. A. and Cresswell, H. P.** (1993). Benefits of brassica  
 581 break crops in the Southeast wheatbelt. *Proc. 7th Aust. Agron. Cons. Adelaide*, 19-24.

- 582 **Lefcheck, J. S.** (2016). piecewiseSEM: Piecewise structural equation modelling in r for  
583 ecology, evolution, and systematics. *Methods in Ecology and Evolution* **7**, 573-579.
- 584 **Leybourne, D. J., Pahl, A., Melloh, P. and Martin, E. A.** (2023). Observations of cabbage  
585 stem flea beetle abundance, associated crop damage and levels of larval infestation, and arthropod  
586 trapping data for oilseed rape fields in Lower Saxony, Germany. DataCat: The Research Data  
587 Catalogue. <https://datacat.liverpool.ac.uk/2511/>.
- 588 **Lundin, O.** (2020). Economic injury levels for flea beetles (*Phyllotreta spp.*; Coleoptera:  
589 Chrysomelidae) in spring oilseed rape (*Brassica napus*; Brassicales: Brassicaceae). *Journal of*  
590 *Economic Entomology* **113**, 808-813.
- 591 **Lundin, O., Rundlöf, M., Jonsson, M., Bommarco, R. and Williams, N. M.** (2021). Integrated  
592 pest and pollinator management – expanding the concept. *Frontiers in Ecology and the Environment*  
593 **19**, 283-291.
- 594 **Marini, L., Batáry, P. and Tscharrntke, T.** (2023). Testing the potential benefits of small fields  
595 for biocontrol needs a landscape perspective. *Proceedings of the National Academy of Sciences* **120**,  
596 e2218447120.
- 597 **Martin, E. A., Dainese, M., Clough, Y., Báldi, A., Bommarco, R., Gagic, V., Garratt, M. P. D.,**  
598 **Holzschuh, A., Kleijn, D., Kovács-Hostyánszki, A. et al.** (2019). The interplay of landscape  
599 composition and configuration: new pathways to manage functional biodiversity and agroecosystem  
600 services across Europe. *Ecology Letters* **22**, 1083-1094.
- 601 **Moraga, A. D., Martin, A. E. and Fahrig, L.** (2019). The scale of effect of landscape context  
602 varies with the species' response variable measured. *Landscape Ecology* **34**, 703-715.
- 603 **Ortega-Ramos, P. A., Cook, S. M. and Mauchline, A. L.** (2022a). How contradictory EU  
604 policies led to the development of a pest: The story of oilseed rape and the cabbage stem flea  
605 beetle. *GCB Bioenergy* **14**, 258-266.
- 606 **Ortega-Ramos, P. A., Coston, D. J., Seimandi-Corda, G., Mauchline, A. L. and Cook, S. M.**  
607 (2022b). Integrated pest management strategies for cabbage stem flea beetle (*Psylliodes*  
608 *chrysocephala*) in oilseed rape. *GCB Bioenergy* **14**, 267-286.
- 609 **Ortega-Ramos, P. A., Mauchline, A. L., Metcalfe, H., Cook, S. M., Girling, R. D. and Collins,**  
610 **L.** (2023). Modelling the factors affecting the spatiotemporal distribution of cabbage stem flea beetle  
611 (*Psylliodes chrysocephala*) larvae in winter oilseed rape (*Brassica napus*) in the UK. *Pest Management*  
612 *Science* **Early View**.
- 613 **Paradis, E. and Schliep, K.** (2019). ape 5.0: an environment for modern phylogenetics and  
614 evolutionary analyses in R. *Bioinformatics* **35**, 526-528.
- 615 **Redlich, S., Martin, E. A. and Steffan-Dewenter, I.** (2018). Landscape-level crop diversity  
616 benefits biological pest control. *Journal of Applied Ecology* **55**, 2419-2428.
- 617 **Rusch, A., Valantin-Morison, M., Sarthou, J. P. and Roger-Estrade, J.** (2013). Effect of crop  
618 management and landscape context on insect pest populations and crop damage. *Agriculture,*  
619 *Ecosystems & Environment* **166**, 118-125.
- 620 **Schaefer, M.** (2018). Brohmer-Fauna von Deutschland: Ein Bestimmungsbuch unserer  
621 heimischen Tierwelt. Germany: Quelle & Meyer Verlag GmbH & Co.
- 622 **Scheiner, C. and Martin, E. A.** (2020). Spatiotemporal changes in landscape crop  
623 composition differently affect density and seasonal variability of pests, parasitoids and biological  
624 pest control in cabbage. *Agriculture, Ecosystems & Environment* **301**, 107051.
- 625 **Schloerke, B., Cook, D., Larmarange, J., Briatte, F., Marbach, M., Thoen, E., Elberg, A.,**  
626 **Toomet, O., Crowley, J., Hofmann, H. et al.** (2021). GGally: Extension to 'ggplot2'. [https://cran.r-](https://cran.r-project.org/web/packages/GGally/index.html)  
627 [project.org/web/packages/GGally/index.html](https://cran.r-project.org/web/packages/GGally/index.html).
- 628 **Seimandi-Corda, G., Winkler, J., Jenkins, T., Kirchner, S. M. and Cook, S. M.** (2023).  
629 Companion plants and straw mulch reduce cabbage stem flea beetle (*Psylliodes chrysocephala*)  
630 damage on oilseed rape. *Pest Management Science* **Early View**.

- 631 **Sulg, S., Kovács, G., Willow, J., Kaasik, R., Smagghe, G., Lövei, G. L. and Veromann, E.**  
 632 (2023). Spatiotemporal distancing of crops reduces pest pressure while maintaining conservation  
 633 biocontrol in oilseed rape. *Pest Management Science* **Early View**.
- 634 **Thies, C. and Tscharrntke, T.** (1999). Landscape Structure and Biological Control in  
 635 Agroecosystems. *Science* **285**, 893-895.
- 636 **Tixeront, M., Dupuy, F., Cortesero, A. M. and Hervé, M. R.** (2023). Understanding crop  
 637 colonization of oilseed rape crops by the cabbage stem flea beetle (*Psylliodes chrysocephala* L.  
 638 (Coleoptera: Chrysomelidae)). *Pest Management Science* **Early View**.
- 639 **Tscharrntke, T., Tylianakis, J. M., Rand, T. A., Didham, R. K., Fahrig, L., Batáry, P., Bengtsson,  
 640 J., Clough, Y., Crist, T. O., Dormann, C. F. et al.** (2012). Landscape moderation of biodiversity  
 641 patterns and processes - eight hypotheses. *Biological Reviews* **87**, 661-685.
- 642 **Veres, A., Petit, S., Conord, C. and Lavigne, C.** (2013). Does landscape composition affect  
 643 pest abundance and their control by natural enemies? A review. *Agriculture, Ecosystems &  
 644 Environment* **166**, 110-117.
- 645 **Warner, D. J., Allen-Williams, L. J., Warrington, S., Ferguson, A. W. and Williams, I. H.**  
 646 (2003). Mapping, characterisation, and comparison of the spatio-temporal distributions of cabbage  
 647 stem flea beetle (*Psylliodes chrysocephala*), carabids, and Collembola in a crop of winter oilseed rape  
 648 (*Brassica napus*). *Entomologia Experimentalis et Applicata* **109**, 225-234.
- 649 **Warton, D. I. and Hui, F. K. C.** (2011). The arcsine is asinine: the analysis of proportions in  
 650 ecology. *Ecology* **92**, 3-10.
- 651 **Weisz, R., Smilowitz, Z. and Christ, B.** (1994). Distance, rotation, and border crops affect  
 652 colorado potato beetle (Coleoptera: Chrysomelidae) colonization and population density and early  
 653 blight (*Alternaria Solani*) severity in rotated potato fields. *Journal of Economic Entomology* **87**, 723-  
 654 729.
- 655 **Wickham, H.** (2016). *ggplot2: Elegant graphics for data analysis*. USA: Springer New York.
- 656 **Wickham, H., Averick, M., Bryan, J., Chang, W., McGowan, L. D. A., François, R.,  
 657 Grolemond, G., Hayes, A., Henry, L. and Hester, J.** (2019). Welcome to the Tidyverse. *Journal of  
 658 open source software* **4**, 1686.
- 659 **Williams, I. H.** (2010). The major insect pests of oilseed rape in Europe and their  
 660 management: an overview. In *Biocontrol-based integrated management of oilseed rape pests*, (ed. I.  
 661 H. Williams), pp. 1-43. The Netherlands: Springer Dordrecht.
- 662 **Willis, C. E., Foster, S. P., Zimmer, C. T., Elias, J., Chang, X., Field, L. M., Williamson, M. S.  
 663 and Davies, T. G. E.** (2020). Investigating the status of pyrethroid resistance in UK populations of the  
 664 cabbage stem flea beetle (*Psylliodes chrysocephala*). *Crop Protection* **138**, 105316.
- 665 **Zaller, J. G., Moser, D., Drapela, T., Schmöger, C. and Frank, T.** (2008a). Effect of within-field  
 666 and landscape factors on insect damage in winter oilseed rape. *Agriculture, Ecosystems &  
 667 Environment* **123**, 233-238.
- 668 **Zaller, J. G., Moser, D., Drapela, T., Schmöger, C. and Frank, T.** (2008b). Insect pests in  
 669 winter oilseed rape affected by field and landscape characteristics. *Basic and Applied Ecology* **9**, 682-  
 670 690.
- 671 **Zavalnitskaya, J., Bloom, E. H. and Szendrei, Z.** (2022). Non-host habitat at local and  
 672 landscape scales promotes overwintering success of a specialist insect pest. *Agriculture, Ecosystems  
 673 & Environment* **338**, 108114.
- 674 **Zhang, Y., Haan, N. L. and Landis, D. A.** (2020). Landscape composition and configuration  
 675 have scale-dependent effects on agricultural pest suppression. *Agriculture, Ecosystems &  
 676 Environment* **302**, 107085.
- 677 **Zheng, X., Koopmann, B., Ulber, B. and von Tiedemann, A.** (2020). A global survey on  
 678 diseases and pests in oilseed rape—current challenges and innovative strategies of control. *Frontiers  
 679 in Agronomy* **2**.
- 680 **Zuur, A. F., Ieno, E. N., Walker, N. J., Saveliev, A. A. and Smith, G. M.** (2009). Mixed effects  
 681 models and extensions in ecology with R. USA: Springer New York.