- Non-crop sources of beneficial arthropods vary within-season across a prairie agroecosystem
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7 Abstract

- 1. Ground-dwelling arthropods can be important generalist predators in agroecosystems, and can use non-crop features as overwintering habitats. However, it is unclear which types of landscape features constitute useful non-crop habitat, and at what spatial scale organisms gather resources. Additionally, the same landscape feature may act as a source or a destination for arthropods at different times of the year, but this is rarely considered.
- 2. We modeled the abundance of four common species of Canadian prairie arthropods caught in a set of 198 in-field and roadside pitfall traps (June to August of 2017). Functional regression was used in order to simultaneously consider both the habitat preferences and the timing of movement from the land cover classes.
- 3. Pterostichus melanarius (Coleoptera: Carabidae) and Pardosa moesta (Araneae: Lycosidae) were attracted to canola (Brassica napus) during the early summer, then dispersed to grasslands, wetlands, and grassy road margins at the end of the summer. In particular, Pterostichus melanarius aggregated in canola early in the growing season, suggesting that its role in suppressing crop pests may be underestimated. Pardosa moesta (Araneae: Lycosidae) and Phalangium opilio (Opiliones: Opilionidae) showed weak patterns of seasonal migration, and were more influenced by large-scale geographic patterns rather than landscape composition.
- 4. Synthesis and applications: Our results suggest that predatory arthropods migrate into canola crops during the early summer, and that grasslands and wetlands act as seasonal reservoir habitats. Farmers and land managers should consider preserving existing habitat in order to maintain pest-control services across the season.
- 8 Keywords: beetles; spiders; harvestmen; movement; ecosystem services; natural enemy; functional data
- 9 analysis; seminatural land

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1. Introduction

Arthropods are one of the most influential groups of animals in agricultural systems. Pest arthropods 11 can cause large reductions in crop yield (Oerke, 2005), which can result in negative downstream economic effects and food insecurity. However, beneficial arthropods living in non-crop patches can increase crop 13 yields by providing ecosystem services, such as extra pollination or pest suppression (Garibaldi et al., 2013; Klein et al., 2007; Losey & Vaughan, 2006). Seasonal "pulses" of food resources act as bottom-up 15 drivers of arthropod communities in crop fields, meaning that agricultural land can increase arthropod abundance in adjacent non-crop land (Diekötter, Kadoya, Peter, Wolters, & Jauker, 2010; Galpern, Johnson, Retzlaff, Chang, & Swann, 2017). These movements of mobile arthropods in and out of crops are important for managing both crop yield and biodiversity in agroecosystems. In this context, finding "win-win" practices – that is, practices that result in better conservation outcomes while maintaining or 20 improving crop yields – is of particular importance. 21 Uncultivated semi-natural land (SNL) can act as important habitat for beneficial arthropods (Duelli

22 & Obrist, 2003), which may spill over into adjacent cropland. Therefore, landscape management is a strategy to increase beneficial arthropod abundance in crops (Albrecht et al., 2010; Landis, Wratten, & 24 Gurr, 2000; Martin et al., 2020; but see Tscharntke et al., 2016; Karp et al., 2018). For example, wild bee abundance in agricultural landscapes is driven by food and nest availability (Roulston & Goodell, 2011), and the effect of SNL on bee abundance and diversity is generally positive. Pest-suppressing predators, such as beetles or spiders, can be limited by water, food, and egg-laying sites (Gardiner et al., 2010; Lövei & Sunderland, 1996; Purtauf et al., 2005), but the effect of SNL on their abundance 29 and diversity is less consistent (Macfadyen & Muller, 2013; Shackelford et al., 2013) as some taxa may use SNL as a travel corridor or overwintering site, but are not completely dependent on it for 31 reproduction. Some taxa are not dependent on SNL at all, and thrive in cultivated landscapes, while 32 others exclusively inhabit SNL and do not venture beyond the edge of cultivated fields. For example, 33 carabid beetles are thought to reproduce in field margins, while the adults feed in fields (Desender & Alderweireldt, 1988). Habitat preference also depends on the life history of the organisms in question, such as the seasonal timing of emergence, feeding, and reproduction, as well as their traits such as dispersal capability. Non-adjacent cropland can also influence arthropod abundance if the organisms 37 are long-distance dispersers (Öberg, Mayr, & Dauber, 2008). In this way, the spillover of beneficial arthropods into crops depends not only on the amount of adjacent SNL, but the specific habitat type, the time of the year, and the spatial scale at which an organism forages.

Despite what is known about the spatial and temporal aspects of beneficial arthropod spillover, these 41 are rarely considered together. First, some non-crop habitat may act as sources of arthropods during some times of the year, but as destinations during other times, meaning that the timing of spillover from 43 non-crop habitat may not align with important periods in crop development. Arthropods can migrate from managed crops back into adjacent SNL at the end of the season (Desender & Alderweireldt, 1988; Tscharntke, Rand, & Bianchi, 2005), but this is seldom considered. Second, the spatial "grain" of the landscape that is relevant to an organism depends on their traits and life history (Ahrenfeldt 47 et al., 2015). For instance, bumblebee abundance can be influenced by both nearby and far-away SNL abundance (Westphal, Steffan-Dewenter, & Tscharntke, 2006), but this is poorly studied for other groups of beneficial organisms (but see Sander, Purtauf, Wolters, & Dauber, 2006). Finally, the feeding and life-history characteristics of many predatory ground beetles and spiders are often poorly-documented, especially in North America. This makes it difficult to make generalizations of 52 which habitat types harbour beneficial arthropods, making it even more difficult to provide accurate information to farmers about land management practices. 54

In this study, we used pitfall traps to determine how landscape composition affects the seasonal abundance of predatory arthropods in a Canadian prairie agroecosystem. We considered the following hypotheses: 1. Untilled semi-natural land provides egg-laying and feeding areas for predatory arthropods, meaning that it should act as a source of arthropods during the early part of the season, and a destination during the later part of the season. 2. Crops provide food (pest insects) for predatory arthropods during the growing season. This should result in agricultural land becoming a destination for predators in the early part of the season as they migrate into the crop, and a source at the end of the season as they migrate out of the crop. 3. Crops may act as a temporary feeding site for predators. Therefore, crops may act as a local destination for arthropods, but will also be negatively associated at larger (landscape-level) scales. Using a large pitfall trapping dataset, we related seasonal changes in arthropod abundance to landscape composition at multiple spatial scales.

6 2. Methods

67 2.1. Data collection

We used a set of 198 pitfall traps installed in road margins (minimum of 5 m away from the road edge, 85 traps) and in-field locations (113) across southern Alberta, Canada in 2017 (Figure 1). The sites spanned a west-to-east gradient of four natural subregions, including foothills parkland, foothills

fescue, mixedgrass, and dry mixedgrass (Natural Regions Committee, 2006). Traps were placed starting on May 16, and collections ended on August 28, with collection occurring continuously and traps being emptied every 14 days on average (SD: 3). This resulted in 850 unique collection events, taking place 73 across a total of 11614 trapping days. In-field traps were placed in canola crops (68), wetlands (16), grassy field edges (11), and remnant prairie grasslands (18). Traps in canola were installed at 25, 75, 75 and 200 m along a transect heading away from the nearest non-crop feature (wetland, grassy field edges, or remnant prairie), while the trap at 0 m was installed in the non-crop feature itself. We used 77 582 mL Solo® cups buried up to the rim and partially filled with propylene glycol, with 2 cm wire mesh mounted over the rim to prevent vertebrates from falling into the traps. Specimens were identified to species using appropriate taxonomic literature (Dondale & Redner, 1990; Edgar, 1990; Lindroth, 1966; Vogel, 2004; Yigit, Bayram, Corak, & Danisman, 2007). We used counts of organisms in our pitfall traps as proxies of ecosystem services (specifically, pest 82 control). However, increased counts of organisms in pitfall traps can represent higher activity levels

control). However, increased counts of organisms in pitfall traps can represent higher activity levels
(same number of organisms but more mobile) or higher abundances in the vicinity of the trap. This
makes it impossible to disentangle arthropod activity from density using single traps (Lang, 2000; Zhao
et al., 2013), meaning that counts from traps represent the "activity-density" of a given organism,
not absolute density. Despite this drawback, activity-density is generally positively related to pest
consumption, at least in carabids (Boetzl, Krimmer, Krauss, & Steffan-Dewenter, 2018; González, Seidl,
Kadlec, Ferrante, & Knapp, 2020; Trichard, Ricci, Ducourtieux, & Petit, 2014), making it acceptable
for our study.

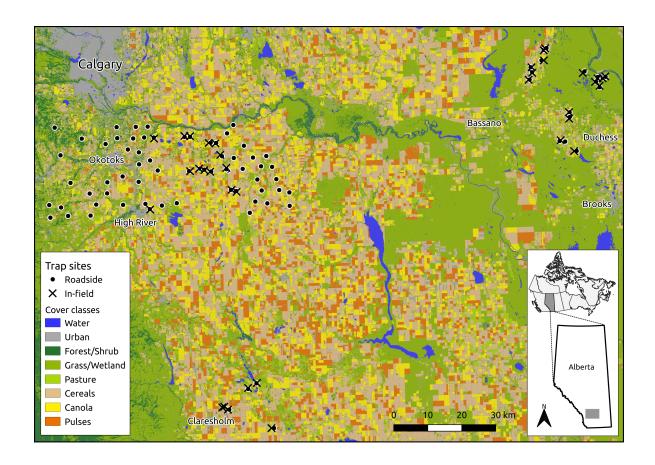


Figure 1: Classified land cover and location of pitfall traps during 2017. Inset map shows study location within Alberta provincial boundaries. Roadside traps were installed in road margins, while in-field traps were installed directly in canola fields, wetlands, or grassy field edges.

We used four highly-abundant species of predatory arthropods found in the pitfall traps, as common 91 species are more important for ecosystem service provision than rare species (Winfree, W. Fox, Williams, Reilly, & Cariveau, 2015). Pterostichus melanarius is an introduced predatory ground beetle (Coleoptera: 93 Carabidae) that is a wide-ranging generalist known to prey upon crop pests (Busch, 2016; Cárcamo & 94 Spence, 1994; Larsen, Work, & Purrington, 2003). It is commonly found in untilled grasslands (Purtauf 95 et al., 2005) and may use grassy field edges as larval habitat (Desender & Alderweireldt, 1988). Pardosa are a genus of wolf spiders (Araneae: Lycosidae) that use an active wandering predation strategy 97 (Young & Edwards, 1990), and are common across the Canadian prairies. Both Pardosa distincta and 98 Pardosa moesta are found in a wide variety of habitats, but are common in ungrazed pastures (Cárcamo, Pinzón, Leech, & Spence, 2014; Dondale & Redner, 1990). P. distincta is also commonly found in 100 disturbed environments (Collins, Jennings, & Forsythe, 1996; Wade & Roughley, 2010), while P. moesta 101 appear to prefer less frequent disturbances (Graham, Buddle, & Spence, 2003; Kowal & Cartar, 2011; 102

but see Dondale & Redner, 1990). Other Pardosa are known to use road margins and wooded areas as 103 habitat (Buddle, 2000; Drapela, Frank, Heer, Moser, & Zaller, 2011). Finally, Phalangium opilio is 104 an introduced harvestman (Opiliones: Phalangiidae) that is found in drier disturbed areas (Bragg & 105 Holmberg, 2009). Juveniles hatch from eggs during the spring, becoming adults during the summer, and 106 eggs are laid during the fall (adults do not overwinter; Bragg & Holmberg, 2009). They are commonly 107 found in human-altered landscapes (Muster & Meyer, 2014; Van de Poel, 2015) and are nocturnal 108 generalist hunters and scavengers (Allard & Yeargan, 2005b, 2005a; Halaj & Cady, 2000), mainly eating 109 small soft-bodied invertebrates (Acosta & Machado, 2007). 110

To characterize landscape composition surrounding the traps, we used publicly available classified 111 landscape data (30 m resolution; Agriculture and Agri-Food Canada, 2018). AAFC cropland landscape 112 classifications from 2017 were very accurate for cultivated fields (~90%), but were less accurate for 113 non-crop areas ($\sim 70\%$). We combined functionally similar landscape categories (cereals: Triticum 114 aestivum and Hordeum vulgare, pulses: Pisum sativum and Lens culinaris, forest: coniferous and broadleaf). At each site, we extracted the proportion of each cover class within 30m annuli (rings), with 116 the inner radius of each annulus ranging from 30 to 1470m in increments of 30m (total of 49 annuli, 1.5 km maximum). The ten most-common cover classes surrounding our sites represented 98% of the 118 total land cover in our study region: grassland, cereal, canola, pasture, pulses, wetland, urban (road 119 margins), shrubland, flax, and forests (Figure S2); flax was removed, as only a single site had nearby 120 flax cover. 121

122 2.2. Analysis

Functional regression (Ramsay & Silverman, 2004; Yen, Thomson, Paganin, Keith, & Mac Nally, 123 2014) was used to incorporate landscape information at different distances, allowing assessment of both 124 local and regional landscape composition (Galpern & Gavin, 2020). Scalar-on-function regression is a 125 special type of linear regression model $(y = X\beta + e)$, where the columns of the model matrix X contain 126 some continuous predictor of the scalar y, and the values of the coefficients (β) are modeled as a smooth 127 function f(x) of the predictors. In our case, the proportion of landscape cover (X) within each annulus 128 surrounding the trap is a function of distance away from the trap, meaning that the coefficients are a 129 smoothed function of distance $(\beta_i = f(distance_i))$, and represent the additive effect of a given type 130 of landscape cover at distances away from the trap (i.e. $X\beta$). This allows for the possibility that the 131 size of the landscape "grain" relevant to a given organism may change over the course of its life stages 132 (Addicott et al., 1987; Gardiner et al., 2010; Lima & Zollner, 1996). It also allows for the possibility 133

that certain land cover types may be locally beneficial, but detrimental at large scales, indicating that
the cover type may not constitute a completely usable habitat.

Scalar-on-function regression of activity-density was fit using generalized additive models (mgcv version 1.8.33; Wood, 2017). Count data of arthropods were modeled using a negative binomial distribution with a log-link function and a single dispersion parameter (θ). To account for different lengths of trap exposure, log-days since trap placement were used as a fixed effect with their slope held at 1 ("offset" variable). Trapping location was included as a fixed effect with 5 levels (canola, field edge, grassland, road margin, and wetlands) to account for the effect of local cover independent of the surrounding landscape composition. Day of year was included as a 1-dimensional smooth, and easting and northing (km east and north of the trap extent centre point) were included as 2-dimensional smooths, in order to account for underlying spatial and temporal autocorrelation in the data.

For each landscape cover class, we used three scalar-on-function terms. First, we used the proportion cover in the annuli surrounding each trap location (spatial effect of cover class). Second, we used the average proportion cover across days of the year (temporal effect of cover class). Finally, we used a tensor-product interaction of landscape cover and day of year (spatio-temporal interaction of cover class). We used 10 basis dimensions for the spatial and temporal landscape smoothers, and 16 basis dimensions for the spatio-temporal interaction. The effective degrees of freedom for all smoothing terms were far below the number of basis dimensions, indicating that no additional basis dimensions were needed (see Tables S2, S4, S6, S8). To remove unimportant terms from the model, we used thin-plate regression splines with shrinkage, a continuous analog to stepwise model selection where weaker terms are completely removed from the model rather than reduced to a line (Marra & Wood, 2011). The deviance residuals from each model were visually inspected for normality and equal variance (Hilbe, 2011; Wood, 2017). Finally, we calculated the proportion of explained variance (R²) for each set of terms (Nakagawa, Johnson, & Schielzeth, 2017; Nakagawa, Schielzeth, & O'Hara, 2013).

Preliminary model fits revealed that some of the landscape terms in our model were strongly concurved with each other, so we removed them from the model or combined them. Concurvity is a nonlinear analogue of multicollinearity which can bias estimates of standard errors (Buja, Hastie, & Tibshirani, 1989). There is no agreed-upon threshold of "unacceptable" concurvity, but 0.5 is commonly used (Dominici, 2002; Ramsay, Burnett, & Krewski, 2003). Forests and shrubland were combined into a single "woodland" category, and grassland and wetland were combined into a single "grassland" category, as they were strongly positively correlated at all distances, indicating similar feature classes.

Cereal was removed from the model, as it was negatively correlated with canola cover at distances less than 200m and was positively correlated at distances over 500m, reflecting the most common crop rotations (canola \rightarrow wheat \rightarrow barley) and the commonly-used 800 m block structure of farmland in our study region (quarter-sections). This resulted in a reduced set of landscape terms, none of which were strongly concurved with each other (cropland: canola, pulses; SNL: grassland, pasture, woodland, roadside; see Figure S1).

Box 1: Interpreting Functional Regression Plots

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Functional linear regression is a special case of linear regression, where both the independent variable is predicted by a functional dependent variable, meaning that the slope is a function (vector) rather a fixed (scalar) value (Ramsay & Silverman, 2004; Yen et al., 2014). Therefore, the interpretation of a functional regression plot is very different from that of a simple linear regression plot. Figure 2a shows a simple linear regression plot of a hypothetical relationship between trap activity-density (y) and a single predictor, proportion cover (x), which is measured in a circle of a fixed radius surrounding each trap. If we suppose that the relationship between x and y is similar at all radii at which x is measured, then Figure 2b shows the resulting functional linear regression plot, where the slope (β) is the same at all distances (we avoid the problem of overlapping concentric circles by using concentric annuli). In reality, it is more likely that nearby cover will be more influential on activity-density than far-away cover, which could result in a relationship similar to Figure 2c. As functional regression takes any continuous predictor of y, it is also possible to use proportion cover (measured within an annulus of a fixed radius) as a predictor of measurements taken across time, allowing the effect of proportion cover to vary over the course of the season (Figure 2d). Finally, the effect of proportion cover over distance and time may be modeled as a smoothed surface, representing the joint spatio-temporal influence of landscape cover on activity-density.

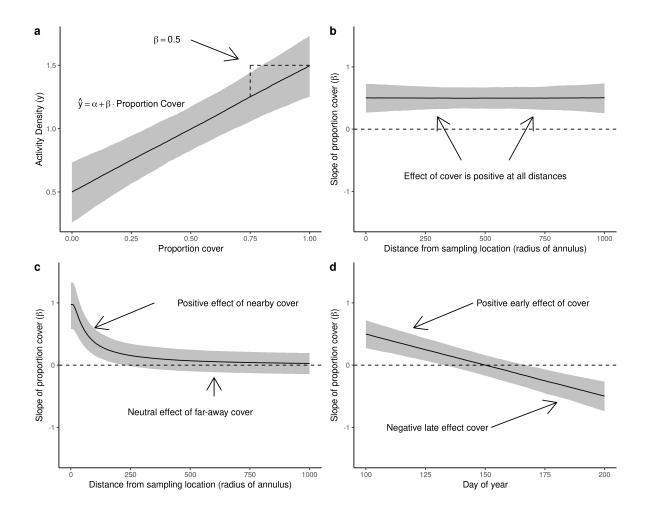


Figure 2: Examples of (functional) linear regression plots. a) shows a simple linear regression plot, with a single slope value for a single regressor. b) shows the equivalent plot in functional regression form, where the the slope is a smoothed function that maps onto a function-valued regressor. c) and d) show functional linear regression plots where the slope varies with the location of the regressor.

Functional regression plots of landscape composition reveal which cover classes are acting as sources or destinations of a given arthropod, and at what spatial scale. Positive effects of nearby landscape cover classes indicate that arthropods are spilling over *from* it, therefore acting as a source. However, negative responses to nearby landscape cover classes can indicate one of two things: 1) It may indicate that arthropods are spilling over *into* it (therefore acting as a destination) or 2) it may indicate that no arthropods are available to spill over *from* it. We consider 1) to be the more likely scenario, as it is unlikely that any one cover class is *completely* unoccupied. For example, a cover class with low-quality habitat could act as an ecological trap (Galpern et al., 2017) if it acts as a destination but never as

a source. Alternatively, a cover class could act as a destination by providing high-quality habitat at certain times of the year (e.g. egg laying/feeding), but act as a source at other times (e.g. during emergence).

In the Results section, we refer to local- and landscape-level effects, where local indicates the effects
of landscape composition at scales of <500 m, whereas landscape indicates scales >500 m. Similarly,
we refer to early- and late-season effects, where landscape composition affected activity-density on June
204 20th or August 20th, respectively.

3. Results

3.1. Landscape composition and trap catches

Grassland, cereal, and canola were the three most-abundant landcover classes surrounding our 207 traps, accounting for 77% of land cover (Figure S2). Several landscape "fingerprints" were evident in the landscape annuli, with cereal cover increasing with distance away, along with a corresponding 209 decrease in canola cover. Study sites were originally chosen for sampling arthropods surrounding canola 210 agroecosystems, so canola was a dominant signal in the landscape, but there was also a large amount 211 of variation in grassland and pasture surrounding each site (Figure S2). Grassland cover was largely 212 constant with distance, but a cluster of sites had uniformly high or low cover of grassland with distance. The pitfall traps caught a total of 18968 Pterostichus melanarius, 5397 Pardosa distincta, 2350 Pardosa 214 moesta, and 34090 Phalangium opilio (mean trapping rates per day: 1.58, 0.49, 0.22, 2.92, SD: 5.69, 215 1.08, 1.09, 5.48, respectively). 216

217 3.2. Ground beetle: Pterostichus melanarius

P. melanarius activity-density was strongly influenced by trap location and landscape composition (Tables S1, S2). Canola crops had a marginally higher activity-density of P. melanarius than other trap locations (Figure 3a). Grassland cover was the most important land cover type in explaining activity-density (R²: 0.104, Table 1), and local grassland had a negative effect late in the summer (Figure 3b), suggesting that P. melanarius may move into grasslands later in the year. Local canola had a positive effect in the early season, but landscape-level canola had a negative effect, indicating that areas with widespread canola coverage had lower P. melanarius activity-density (Figure 3c). Pulses acted as a late-season source, indicating that P. melanarius may migrate out of the crop after (or during) harvest (Figure 3d). Finally, P. melanarius activity-density had a strong temporal and spatial component (both

p <0.001), indicating that phenology and local geographic factors were important drivers of ground beetle activity-density (Figure S3). Landscape composition and trap location explained 22% of the variance in activity-density, while the spatial and temporal smoothers accounted for 41% (Table 1).

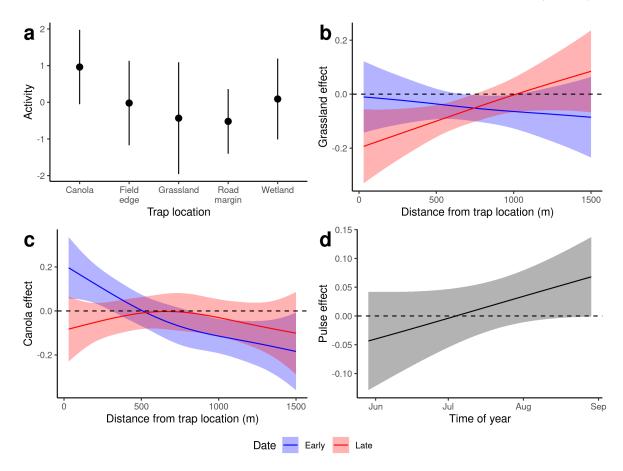


Figure 3: Landscape influence on *Pterostichus melanarius* activity-density. a) shows the effect of local trap location, and b-c) are functional regression plots of the proportion grassland and canola at distances away from each trap. d) is a functional regression plot over time, showing the effect of pulses at different times of the year. Lines and dots represent means, and bars and shaded regions represent 95 % confidence regions (1.96 x SE). Coloured regions represent early- and late-season effects (blue = June, red = August).

3.3. Wolf spiders: Pardosa distincta and Pardosa moesta

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Activity-density of P. distincta and P. moesta was influenced by trap location and landscape composition (Tables S3, S4, S5, S6), but landscape composition had relatively weak effects on both species (R^2 : 0.0446, 0.0963). Unlike P. melanarius, both P ardosa species had far lower activity-density in canola than any other cover type (Figures 4a, 5a). Activity-density of both P ardosa species had a strong temporal and spatial component, although the temporal component was dominant for P. distincta, whereas the spatial component was dominant for P. moesta (Figures S4 and S5; both p<0.001).

 $P.\ distincta$ activity-density was negatively affected by landscape-level pasture (Figure 4b), while woodlands had a negative effect late in the season (Figure 4c). This suggests that large amounts of pasture are unsuitable habitat for $P.\ distincta$, and that they migrate into woodlands later in the season. However, trap location and the temporal random effect (Figure S4a) explained roughly 10 times more variance than landscape composition (R^2 : 0.45 versus 0.045), meaning that while $P.\ distincta$ may be locally abundant in certain cover types, it is highly general in its overall habitat preferences.

P. moesta activity-density responded to grassland, canola, and road margins. Grasslands had a 243 local negative effect, but a positive landscape-level effect, indicating that while these cover types act 244 as a local destination, the general amount of grassland in the area had a positive effect (Figure 5b). 245 Local canola cover had a positive early effect on activity-density, but a negative local effect later in the 246 season (Figure 5c). Local road margins also had a similar effect, acting as an early source and a late destination (Figure 5d). Pulses also had a positive effect at the landscape level (Figure 5e), but this 248 explained very little variance in P. moesta activity-density (Table 1, R^2 : 0.004). Similar to P. distincta, trap location and an East-West spatial random effect (Figure S5b) explained roughly 5 times more 250 variance in P. moesta activity-density than landscape composition (R^2 : 0.5 versus 0.096), meaning that P. moesta is also fairly general in its overall habitat preferences, and is influenced more heavily by 252 larger-scale abiotic factors. 253

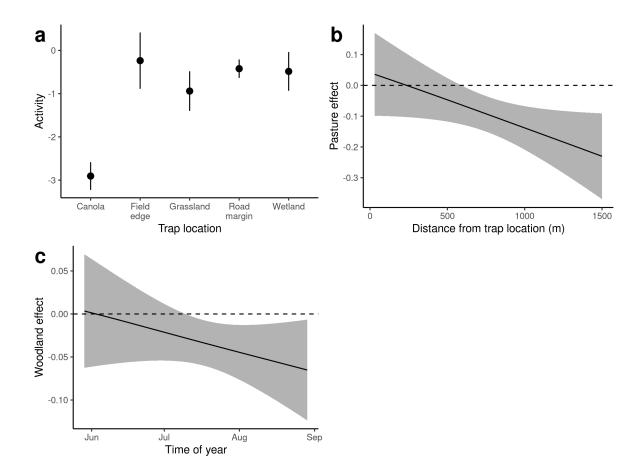


Figure 4: Landscape influence on $Pardosa\ distincta$ activity-density. a) shows the effect of local trap location, b) is a functional regression plot of the proportion pasture at distances away from each trap, and c) is a functional regression plot over time, showing the effect of woodland at different times of the year. Lines and dots represent means, and bars and shaded regions represent 95 % confidence regions (1.96 x SE).

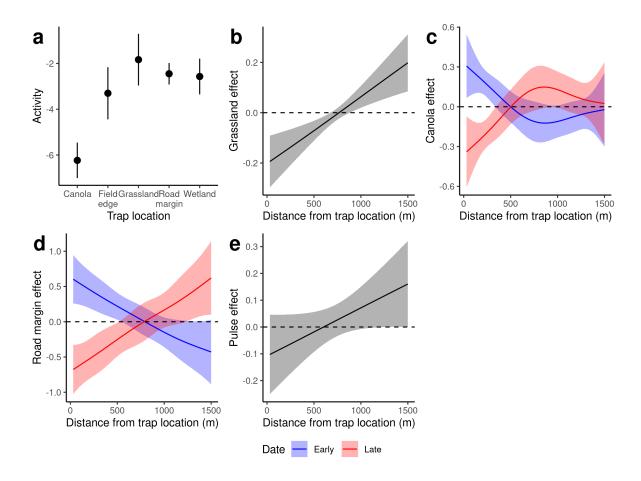


Figure 5: Landscape influence on $Pardosa\ moesta$ activity-density. a) shows the effect of local trap location, and b-e) are functional regression plots of the proportion grassland, canola, road margins, and pulses at distances away from each trap. Lines and dots represent means, and bars and shaded regions represent 95 % confidence regions (1.96 x SE). Coloured regions represent early- and late-season effects (blue = June, red = August).

3.4. Harvestman: Phalangium opilio

 $P.\ opilio$ activity-density was somewhat influenced by trap location, with field edges and wetlands having the highest activity-density (Figure 6a), but this effect was swamped by the effect of landscape and the spatiotemporal random effects (Table 1), meaning that $P.\ opilio$ activity-density is only mildly influenced by local cover type. Nearby grassland and woodland both had negative effects on Phalangium opilio activity-density, but only early in the season (Figures 6b,c), suggesting that $P.\ opilio$ may preferentially migrate to these cover classes. The temporal random effect was overwhelmingly the most important predictor of $P.\ opilio$ activity-density (R^2 : 0.44), showing that activity-density increased during the spring, and showed very little decline during the rest of the season (Figure S6a). Similar to $P.\ moesta,\ P.\ opilio$ also showed a distinct East-West spatial random effect (Figure S6b), indicating

that their activity-density is also influenced by larger-scale abiotic factors.

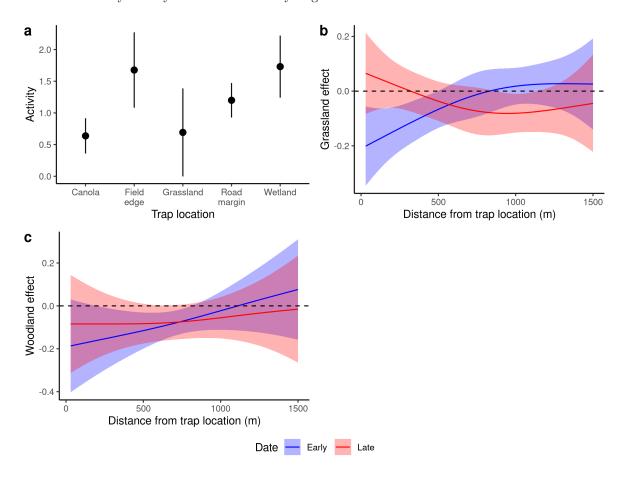


Figure 6: Landscape influence on *Phalangium opilio* activity-density. a) shows the effect of local trap location, and b-c) are functional regression plots of the proportion grassland and woodland at distances away from each trap. Lines and dots represent means, and bars and shaded regions represent 95 % confidence regions (1.96 x SE). Coloured regions represent early- and late-season effects (blue = June, red = August).

4. Discussion

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Our prediction of SNL having late-season effects on beneficial arthropods was supported (Table 2). Specifically, grassland had a negative effect on *P. melanarius* activity-density, woodlands on *P. distincta* activity-density, and road margins on *P. moesta* activity-density, all during late summer (August - September). This suggests that these arthropods migrate to these habitats at the end of the summer to prepare for oviposition or overwintering. However, *P. opilio* showed no evidence of this, instead showing a negative effect of grassland *early* in the season, suggesting that a) they aggregate in grasslands or b) grasslands do not constitute suitable habitat for *P. opilio*, resulting in low spillover.

Table 1: ${\bf R}^2_{GLMM}$ for model components, representing the proportion of total variance explained by each term. ${\bf R}^2$ was calculated using methods from Nakagawa et al. 2013 and Nakagawa et al. 2017.

Term	Pterostichus	Pardosa	Pardosa	Phalangium
161111	melanarius	distincta	moesta	opilio
Trap Location	0.080	0.294	0.379	0.026
Day of Year	0.040	0.160	0.000	0.436
Spatial Location	0.373	0.000	0.125	0.085
Grassland	0.104	0.000	0.017	0.086
Canola	0.030	0.002	0.037	0.000
Pasture	0.000	0.032	0.002	0.000
Woodland	0.000	0.009	0.001	0.017
Pulses	0.005	0.000	0.004	0.001
Road margins	0.000	0.001	0.034	0.002
Residual	0.369	0.502	0.400	0.348

Table 2: Synthesis of model results. Pest control potential is based on likelihood of early-season aggregation to canola crops.

Species	Summary	Pest control potential
Pterostichus melanarius	Moves to canola during early season Disperses to grassland during late season	High
Pardosa distincta	 Negative effect of pasture May move locally to field edges	Low
Pardosa moesta	 Moves to canola during early season Disperses to road margins and grassland during late season Moves to grasslands across the season. 	Medium
Phalangium opilio	· Moves to grassland early in the season	Low

There was little evidence of SNL acting as early-season sources, aside from road margins having a 273 positive early effect on P. moesta activity-density. However, some cover classes had a positive effect 274 at the landscape level (grassland and road margins on P. moesta), suggesting earlier or larger-scale 275 dispersal away from these cover classes. Canola had a positive effect on P. melanarius and P. moesta in the early season, suggesting that these organisms aggregate to canola crops and disperse at the end 277 of the season (mass effects sensu Shmida & Wilson, 1985). Canola also had negative landscape-level effects on P. melanarius while pasture had negative effects on P. distincta, meaning that these cover 279 types likely represent only partial habitat for these arthropods. Finally, the influence of landscape composition tended to be limited to a radius of about 500m, indicating that large-scale and local 281 land cover can be important to beneficial arthropod abundance. There was also a strong temporal 282 component to most of the landscape effects, meaning that it is important to consider how destinations and sources may change over the season. 284

The predatory ground beetle *Pterostichus melanarius* responded to landscape composition, primarily to canola, pulses, and grass and wetland cover. Grassland and wetland cover had a negative late-season 286 effect, suggesting that they act as a destination for P. melanarius at the end of the summer. Other studies have found similar results, suggesting that untilled land is important off-field habitat for carabids (Desender & Alderweireldt, 1988; Fournier & Loreau, 2002; but see Hatten, Bosque-Pérez, Labonte, 289 Guy, & Eigenbrode, 2007). In particular, Purtauf et al. (2005) found that carabid density increased with proportion of nearby grasslands, so our negative late-season effect of grasslands may represent P. 291 melanarius moving to overwintering habitat at the end of the season. However, P. melanarius is also 292 found in a wide variety of habitats, and tends to prefer agricultural fields (Larsen et al., 2003). Our 293 models confirmed this, as they showed that canola and pulse crops had positive effects on P. melanarius activity-density early in the season. Early in the season, canola crops had a local positive effect, but a 295 landscape-level negative effect, which may indicate that they provide only temporary feeding habitat. The effect of pulse crops on carabid abundance is not well-studied, but our results suggest that pulses 297 may represent a potential food resource for P. melanarius, albeit a weak one (Table 1). Carabids are predators of pea-leaf weevils (Sitona lineatus) on other Fabaceae (Hamon, Bardner, Allen-Williams, & Lee, 1990), and P. melanarius can prey upon other carabids that are predators of S. lineatus, such as Bembidion quadrimaculatum (Vankosky, Cárcamo, & Dosdall, 2011), so this effect may be direct 301 or indirect. We did not consider the influence of previous years' crops, although carabids tend to be 302 more abundant in areas of high crop rotation (Bertrand, Burel, & Baudry, 2016; Busch, 2016). Finally, P. melanarius has long- and short-winged morphs with very different dispersal abilities (Bourassa,
 Spence, Hartley, & Lee, 2011; Niemelä & Spence, 1999), which may influence the degree to which
 landscape-level proportion cover impacts activity-density.

The wolf spiders Pardosa distincta and Pardosa moesta both responded to landscape composition, but it explained only a small fraction of their activity-density (R^2 : 0.045, 0.096). This means that both 308 species of Pardosa appear to have very general habitat preferences, but may be somewhat influenced by certain cover types. Pasture had a negative effect at the landscape scale, and woodland had a 310 late-season negative effect, suggesting that P. distincta move to woodlands for overwintering or winter 311 foraging (Aitchison, 1984; Buddle, 2000). Pardosa are found in areas close to disturbed roadside strips 312 (Drapela et al., 2011; Drapela, Moser, Zaller, & Frank, 2008; Kowal & Cartar, 2011), but P. distincta 313 appears to be less sensitive to disturbances than P. moesta (Collins et al., 1996; Wade & Roughley, 2010). Moring & Stewart (1994) showed that P. distincta were active in grassy habitats, which we 315 found partial evidence of, as P. distincta tended to have higher activity-density in trap locations with higher grass cover (Figure 4a). Canola had an early-season positive effect on P. moesta, indicating 317 that P. moesta and P. melanarius may both use canola as foraging grounds. We also found that road margins had a positive early-season and a negative late-season effect (similar to Drapela et al., 2008, 319 and @drapela2011). Grassland and pulses also had a local negative effect but a positive landscape-level 320 effect; this suggests that they may constitute suitable habitat for P. moesta and attract them away 321 from other cover types. Wolf spiders (Lycosidae) employ a wandering-active predation strategy (Young 322 & Edwards, 1990), meaning that nearby landscape composition may be more influential to Pardosa 323 than large-scale composition (Öberg, Ekbom, & Bommarco, 2007; Öberg et al., 2008). While trapping 324 location explained a large proportion of variance for both species (0.29 and 0.38), several landscape 325 features were also important at large spatial scales, including grass and wetland, pasture, pulses, and 326 road margins. Therefore, Pardosa dispersal distances may be further than previously thought, either through ballooning as juveniles (Greenstone, 1982; Richter, 1970) or through other long-distance travel. 328 Phalanqium opilio activity-density responded to grass and wetland cover, as well as wooded landscapes. Other studies have found that P. opilio is mostly found in human-altered landscapes with 330 a large proportion of farming (Muster & Meyer, 2014; Van de Poel, 2015). Since most of our study 331 area was in or near farmed land, the lack of variation is understandable, but there were a few notable 332 landscape-level patterns that emerged. P. opilio activity-density was higher in grassy field edges and 333 wetlands, but grasslands and wetlands had a local negative effect on P. opilio until late in the season. These indicate that while grass and wetlands could act as a reservoir, spillover into other land cover types may be limited, as P. opilio avoids heavily grazed areas, possibly due to low humidity (Šajna, Kušar, Novak, & Novak, 2011). P. opilio is also found in large numbers at the edges of forests, and may migrate from forests into farmlands (Van de Poel, 2015). We found the opposite pattern: forests had a weak negative effect on P. opilio, at least early in the year. Unlike the other arthropods, most of the variation in activity-density for P. opilio was explained by day of year ($R^2 = 0.44$), indicating strong seasonal emergence cues. However, there was also a strong East-West spatial random effect in both P. opilio and P. moesta, meaning that larger-scale geographic patterns (possibly related to rainfall or temperature) also influence their activity-density.

Many other studies have considered the overall effect of SNL on ecosystem service provision, but 344 we have highlighted the different spatial and temporal aspects of these services. We have shown how a relatively straightforward statistical technique can be used to consider multiple spatial scales of 346 landscape composition, providing richer inference about the processes acting on beneficial arthropods. Our results show empirical evidence of arthropod migration between cover types at different times of 348 year, which is rarely considered beyond single-field studies. Since we did not directly track individual movements, future studies should directly examine arthropod movement and life-history within matrices 350 of cover types, with the goal of integrating landscape ecology and behavioural processes into a single 351 model (Lima & Zollner, 1996). This would also allow direct inference about landscape categories that were combined in our dataset, allowing us to consider different landscape categories independently. 353 Finally, future work should explicitly link landscape structure, arthropod abundance, and ecosystem 354 services (Gagic et al., 2017). 355

Our work has three main applications. First, it adds to the basis of scientific evidence showing
that SNL can act as reservoir habitats in intensely-managed agroecosystems. Based on our findings,
farmers and land managers should consider preserving existing grassland and wetland habitat to
act as habitat for pest-suppressing arthropods in crops. While local habitat appears valuable, even
landscape-level habitat can be valuable, as it can increase the number of beneficial arthropods at larger
spatial scales (e.g. Figures 3, 5). Secondly, our work suggests that *P. melanarius* and *P. moesta* should
be investigated for their role as predators in canola crops (Table 2). They appear to aggregate in canola
crops early in the season, meaning that they have the potential to provide valuable pest-control services
to growers. Finally, we demonstrate that timing and spatial scale of spillover should be considered
in future landscape studies of ecosystem services. We show important differences in the timing of

- landscape effects on activity-density (early-season vs. late-season spillover) that are typically ignored,
- ₃₆₇ but should be considered when assessing how SNL can deliver ecosystem services.

5. Authors' contributions

- SVJR and PG conceived of the project. JV, DE, and LB conducted field work and taxonomic work.
- 370 SVJR designed the statistical analysis and wrote the manuscript. JV, DE, LB, and PG all gave critical
- feedback on the manuscript, and give their approval for the final version.

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614 Supplementary Material

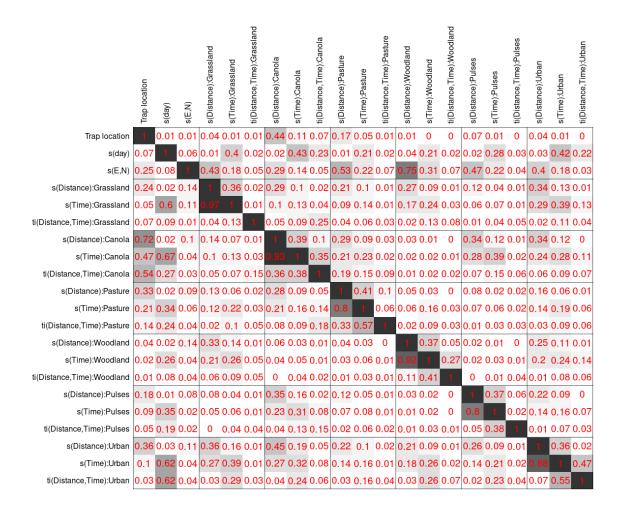


Figure S1: Concurvity estimates for reduced cover classes used in models.

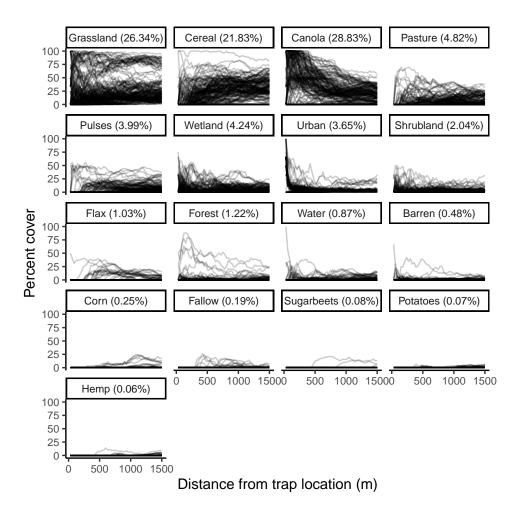


Figure S2: Percent cover of landscape cover classes in annuli surrounding each trap location. Sites are represented by individual lines. Mean cover for each class is listed in each sub-heading.

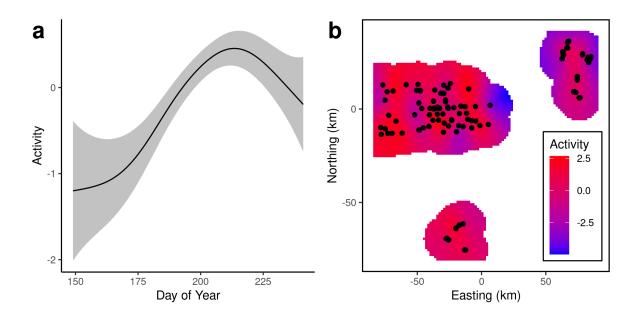


Figure S3: Temporal and spatial components of *Pterostichus melanarius* activity-density (after accounting for landscape composition and trap location). a) shows the effect of day of year, and b) shows the spatially smoothed effect of site location.

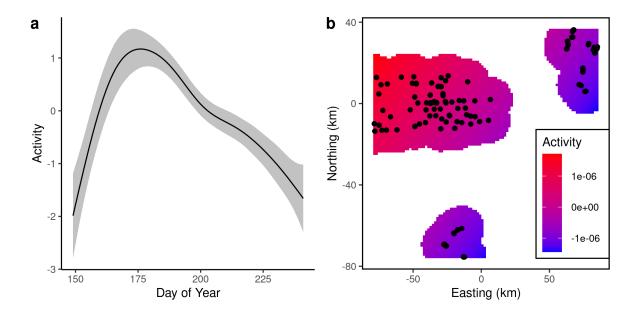


Figure S4: Temporal and spatial components of *Pardosa distincta* activity-density (after accounting for landscape composition and trap location). a) shows the effect of day of year, and b) shows the spatially smoothed effect of site location.

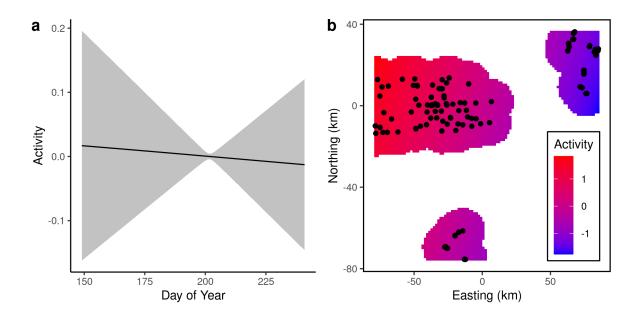


Figure S5: Temporal and spatial components of *Pardosa moesta* activity-density (after accounting for landscape composition and trap location). a) shows the effect of day of year, and b) shows the spatially smoothed effect of site location.

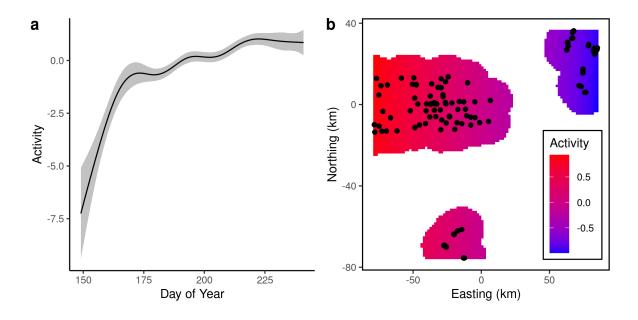


Figure S6: Temporal and spatial components of *Phalangium opilio* activity-density (after accounting for landscape composition and trap location). a) shows the effect of day of year, and b) shows the spatially smoothed effect of site location.

Table S1: Trap location (cover type that trap was located in) intercept estimates for Pterostichus melanarius

Trap location	β	S.E.	Z	p
Canola	0.96	0.52	1.86	0.063
Road margin	-0.52	0.45	-1.16	0.247
Grassland	-0.43	0.78	-0.56	0.578
Field edge	-0.02	0.59	-0.04	0.972
Wetland	0.09	0.56	0.16	0.874

Table S2: Smooth terms for Pterostichus melanarius. s indicates a thin-plate spline, ti indicates a tensor-product interaction. All terms except for day and (E,N) are functional regression fits.

Smoothing term	E.d.f.	χ^2	р
s(Day)	3.48	31.11	< 0.001
s(E,N)	39.47	309.8	< 0.001
s(Distance):Grassland	1.51	11.24	< 0.001
s(Day):Grassland	< 0.01	< 0.01	0.769
ti(Distance,Day):Grassland	0.93	3.87	0.026
s(Distance):Canola	1.71	7.36	0.004
s(Day):Canola	< 0.01	< 0.01	0.824
ti(Distance,Day):Canola	1.64	8.86	0.002
s(Distance):Pasture	< 0.01	< 0.01	0.979
s(Day):Pasture	0.67	0.92	0.23
ti(Distance,Day):Pasture	< 0.01	< 0.01	0.907
s(Distance):Woodland	< 0.01	< 0.01	0.439
s(Day):Woodland	< 0.01	< 0.01	0.939
ti(Distance,Day):Woodland	< 0.01	< 0.01	0.983
s(Distance):Pulses	< 0.01	< 0.01	0.498
s(Day):Pulses	1.27	3.96	0.033
ti(Distance,Day):Pulses	1.4	2.24	0.16
s(Distance):Urban	< 0.01	< 0.01	0.277
s(Day):Urban	< 0.01	< 0.01	0.582
ti(Distance,Day):Urban	< 0.01	< 0.01	0.591

Table S3: Trap location (cover type that trap was located in) intercept estimates for Pardosa distincta

Trap location	β	S.E.	Z	p
Canola	-2.91	0.16	-17.77	< 0.001
Road margin	-0.42	0.11	-3.88	< 0.001
Grassland	-0.94	0.23	-4.02	< 0.001
Field edge	-0.24	0.33	-0.71	0.477
Wetland	-0.49	0.23	-2.12	0.034

Table S4: Smooth terms for $Pardosa\ distincta.\ s$ indicates a thin-plate spline, ti indicates a tensor-product interaction. All terms except for day and (E,N) are functional regression fits.

Smoothing term	E.d.f.	χ^2	р
s(Day)	5.59	120.42	< 0.001
s(E,N)	< 0.01	< 0.01	0.958
s(Distance):Grassland	< 0.01	< 0.01	0.966
s(Day):Grassland	< 0.01	< 0.01	0.94
ti(Distance,Day):Grassland	< 0.01	< 0.01	0.616
s(Distance):Canola	< 0.01	< 0.01	0.927
s(Day):Canola	< 0.01	< 0.01	0.645
ti(Distance,Day):Canola	0.75	2.18	0.077
s(Distance):Pasture	1.76	14.9	< 0.001
s(Day):Pasture	0.46	0.66	0.202
ti(Distance,Day):Pasture	< 0.01	< 0.01	0.799
s(Distance):Woodland	0.01	0.01	0.098
s(Day):Woodland	1.61	7.74	0.005
ti(Distance,Day):Woodland	0.56	0.98	0.153
s(Distance):Pulses	< 0.01	< 0.01	0.693
s(Day):Pulses	< 0.01	< 0.01	0.921
ti(Distance,Day):Pulses	< 0.01	< 0.01	0.43
s(Distance):Urban	< 0.01	< 0.01	0.725
s(Day):Urban	0.7	1.03	0.224
ti(Distance,Day):Urban	< 0.01	< 0.01	0.919

 ${\it Table S5: Trap\ location\ (cover\ type\ that\ trap\ was\ located\ in)\ intercept\ estimates\ for\ {\it Pardosa\ moesta.}}$

Trap location	β	S.E.	Z	p
Canola	-6.24	0.40	-15.76	< 0.001
Road margin	-2.45	0.24	-10.22	< 0.001
Grassland	-1.83	0.58	-3.18	0.001
Field edge	-3.30	0.58	-5.67	< 0.001
Wetland	-2.57	0.40	-6.44	< 0.001

Table S6: Smooth terms for $Pardosa\ moesta.\ s$ indicates a thin-plate spline, ti indicates a tensor-product interaction. All terms except for day and (E,N) are functional regression fits.

Smoothing term	E.d.f.	χ^2	p
s(Day)	0.05	0.04	0.349
s(E,N)	1.96	50.7	< 0.001
s(Distance):Grassland	1.87	14.39	< 0.001
s(Day):Grassland	< 0.01	< 0.01	0.944
ti(Distance,Day):Grassland	< 0.01	< 0.01	0.976
s(Distance):Canola	< 0.01	< 0.01	0.341
s(Day):Canola	< 0.01	< 0.01	0.627
ti(Distance,Day):Canola	2.22	21.63	< 0.001
s(Distance):Pasture	< 0.01	< 0.01	0.936
s(Day):Pasture	< 0.01	< 0.01	0.835
ti(Distance,Day):Pasture	1.6	2.96	0.109
s(Distance):Woodland	< 0.01	< 0.01	0.646
s(Day):Woodland	0.95	2.23	0.085
ti(Distance,Day):Woodland	< 0.01	< 0.01	0.414
s(Distance):Pulses	1.48	3.94	0.039
s(Day):Pulses	0.4	0.5	0.221
ti(Distance,Day):Pulses	< 0.01	< 0.01	0.984
s(Distance):Urban	< 0.01	< 0.01	0.846
s(Day):Urban	0.63	0.8	0.252
ti(Distance,Day):Urban	3.15	20.23	< 0.001

Table S7: Trap location (cover type that trap was located in) intercept estimates for Phalangium opilio

Trap location	β	S.E.	Z	p
Canola	0.64	0.14	4.49	< 0.001
Road margin	1.20	0.14	8.64	< 0.001
Grassland	0.69	0.35	1.95	0.051
Field edge	1.68	0.30	5.52	< 0.001
Wetland	1.73	0.25	6.92	< 0.001

Table S8: Smooth terms for $Phalangium\ opilio.\ s$ indicates a thin-plate spline, ti indicates a tensor-product interaction. All terms except for day and (E,N) are functional regression fits.

Smoothing term	E.d.f.	χ^2	р
s(Day)	7.2	141.89	< 0.001
s(E,N)	1.95	59.8	< 0.001
s(Distance):Grassland	1.69	50.17	< 0.001
s(Day):Grassland	< 0.01	< 0.01	0.685
ti(Distance,Day):Grassland	1.39	5.79	0.013
s(Distance):Canola	< 0.01	< 0.01	0.466
s(Day):Canola	< 0.01	< 0.01	0.882
ti(Distance,Day):Canola	< 0.01	< 0.01	0.675
s(Distance):Pasture	0.36	0.42	0.291
s(Day):Pasture	< 0.01	< 0.01	0.436
ti(Distance,Day):Pasture	< 0.01	< 0.01	0.585
s(Distance):Woodland	1.44	14.8	< 0.001
s(Day):Woodland	< 0.01	< 0.01	0.601
ti(Distance,Day):Woodland	1.81	4.06	0.073
s(Distance):Pulses	< 0.01	< 0.01	0.716
s(Day):Pulses	< 0.01	< 0.01	0.713
ti(Distance,Day):Pulses	1.23	1.88	0.187
s(Distance):Urban	0.79	1.94	0.088
s(Day):Urban	< 0.01	< 0.01	0.419
ti(Distance,Day):Urban	0.36	0.53	0.224