1   The interplay of landscape composition and configuration: new pathways to manage

2   functional biodiversity and agro-ecosystem services across Europe

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124   Abstract

125   Managing agricultural landscapes to support biodiversity and ecosystem services are key aims

126   of a sustainable agriculture. However, how the spatial arrangement of crop fields and other

127   habitats in landscapes impacts arthropods and their functions is poorly known. Synthesizing

128   data from 49 studies (1,515 landscapes) across Europe, we examined effects of landscape

129   composition (% habitats) and configuration (edge density) on arthropods in fields and their

130   margins, pest control, pollination and yields. Configuration effects interacted with proportions

131   of crop and non-crop habitats, and species' dietary, dispersal and overwintering traits led to

132   contrasting responses to landscape variables. Overall, however, in landscapes with high edge

133   density, 70% of pollinator and 44% of natural enemy species reached highest abundances and

134   pollination and pest control improved 1.7 and 1.4-fold, respectively. Arable-dominated

135   landscapes with high edge densities achieved high yields. This suggests that enhancing edge

136   density in European agroecosystems can promote functional biodiversity and yield-enhancing

137   ecosystem services.

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146   INTRODUCTION

147   Worldwide, intensive agriculture threatens biodiversity and biodiversity-related ecosystem

148   services (Foley *et al.* 2005). At a local field scale, monocultures and pesticides restrict many

149   arthropods and plants to non-cropped areas (Geiger *et al.* 2010). Thus, the majority of

150   organisms that provide key regulating services to agriculture, such as pollination and natural

151   pest control, must colonize fields from non-cropped, semi-natural areas (e.g. road verges,

152   grass margins, hedgerows, fallows), neighboring fields or in the wider landscape (Blitzer *et*

153   *al.* 2012). Semi-natural habitats, however, are often removed to facilitate the use of modern

154   machinery or converted to crops to increase production (Naylor & Ehrlich 1997), resulting in

155   reduced populations of service providing organisms (Holland *et al.* 2016). Consequently, the

156   sustainability of modern food production is increasingly questioned (Garnett *et al.* 2013).

157   'Ecological intensification' has the potential to enhance the sustainability of agricultural

158   production by increasing the benefits agriculture derives from ecosystem services (Bommarco

159   *et al.* 2013). Supporting populations of ecosystem service providers is a key component of

160   ecological intensification (Bommarco *et al.* 2013). However, we currently lack detailed

161   knowledge on the landscape-scale management choices needed to achieve ecological

162   intensification with a high degree of certainty (Kleijn et al. 2019). For example, semi-natural

163   habitats are prerequisite for many organisms, but effects are often taxon-specific. In addition,

164   the presence or abundance of functional groups of organisms in a landscape does not always

165   correlate with the services they provide to crops (Tscharntke *et al.* 2016; Karp *et al.* 2018).

166   The configuration of landscapes (size, shape and spatial arrangement of land-use patches), in

167   addition to their composition (proportion of land-use types), is increasingly suggested as a key

168   factor in determining biodiversity and associated ecosystem services in agricultural

169   landscapes (Fahrig 2013). However, studies have only begun to disentangle the relative roles

170   of the composition *vs*. the configuration of habitats and fields within landscapes (Fig. 1;

171   Fahrig 2013; Haddad *et al.* 2017). Landscape configuration can be measured as the density of

172   edges between crop fields and their surroundings, including neighboring crops and non-crop

173   areas. Complex landscapes where small and/or irregularly shaped fields and habitat patches

174   prevail have a high density of edges. Due to increased opportunities for exchange, these

175   landscapes are likely to support spillover of dispersal-limited populations between patches

176   (Smith *et al.* 2014; Fahrig 2017). This may enhance populations' survival in the face of

177   disturbance and their potential to provide services in crops (Boetzl *et al.* 2019). Further, if

178   landscapes with high edge density are also spatially and temporally diverse in their

179   composition, organisms in these landscapes may benefit from landscape-scale resource

180   complementation and supplementation (Dunning *et al.* 1992). In this context, areas offering

181   refuges or complementary food resources may encompass uncropped (semi-natural) areas, but

182   also neighboring crops with asynchronous phenology, different host species and/or variable

183   timing and intensity of management interventions (Vasseur *et al.* 2013; Schellhorn *et al.*

184   2015). However, previous studies have found contrasting effects of increasing configurational

185   complexity for different taxa (Concepción *et al.* 2012; Plećaš *et al.* 2014; Duflot *et al.* 2015;

186   Fahrig *et al.* 2015; Gámez-Virués *et al.* 2015; Perović *et al.* 2015; Martin *et al.* 2016; Bosem

187   Baillod *et al.* 2017; Hass *et al.* 2018). Thus, there is currently no consensus on the importance

188   of landscape configuration for arthropods and the services they provide in crops (Seppelt *et*

189   *al.* 2016; Perović *et al.* 2018). Further, interactions between landscape composition and

190   configuration might explain seemingly contradictory results, but have rarely been tested in

191   part due to a lack of independent landscape gradients (but see Coudrain *et al.* 2014; Bosem

192   Baillod *et al.* 2017).

193   Species' responses to environmental filters depend on sets of biological traits ('response

194   traits'), such as diet breadth and dispersal ability, that constrain species' reactions to

195   environmental predictors (Lavorel & Garnier 2002). The resulting filtering of ecological

196   communities determines the presence or abundance of arthropods able to provide ecosystem

197   services (Gámez-Virués *et al.* 2015). Organisms with similar responses to environmental

198   filters may share specific combinations of response traits, known as trait syndromes.

199   Characterizing these syndromes and their responses to landscape gradients is critical to

200   predict the consequences of land-use change for biological communities (Mouillot *et al.*

201   2013) and the services they provide. However, trait-based responses of arthropods in cropland

202   to landscape gradients have only recently been investigated (Bartomeus *et al.* 2018; Perović *et*

203   *al.* 2018) and cross-taxonomic approaches in agroecosystems are lacking (but see Gámez-

204   Virués *et al.* 2015). For pollinators, natural enemies and pests in agricultural landscapes, a

205   high diversity of responses due to trait variation within and between groups ('response

206   diversity') is likely to underlie observed abundance patterns. In turn, this may affect our

207   ability to manage landscapes for maximum abundance and/or effectiveness of crop ecosystem

208   service-providers, and for minimum impacts of pests.

209   Here, using data from 49 studies covering 1,515 European agricultural landscapes and more

210   than 15 crops, we aim to disentangle arthropod responses to landscape gradients and their

211   consequences for agricultural production by performing the first empirical quantitative

212   synthesis of the effects of landscape configuration (edge density) and composition (amount of

213   crop and semi-natural habitats) on arthropods and their services in cropland. We include

214   observations of the abundance of pollinators, pests and pests' natural enemies (predators and

215   parasitoids) sampled in fields and their margins, and measures of natural pest control,

216   pollination, and crop yields. We use landscape predictors calculated similarly for all studies

217   from high resolution maps with standard land use-land cover classification. We test the

218   following *predictions*:

219   1. *Within functional groups of pollinators, pests and natural enemies, responses to landscape*

220   *predictors differ among trait syndromes.* Thus, considering key trait syndromes of arthropods

221   should increase our ability to predict the effects of landscape variables on functional groups.

222   On one hand, species that use specific crop or non-crop resources should benefit from

223   increased proportions of these resources (habitats) in the landscape (Tscharntke *et al.* 2012).

224   On the other hand, species with medium to low dispersal ability and diet or habitat needs

225   outside crops should be most abundant in fields and margins of landscapes with high edge

226   density, due to shorter travel distances and/or greater resource complementation between

227   habitats and crops (Smith *et al.* 2014).

228   2. *Effects of landscape composition and configuration interact.* Increasing resources in

229   surrounding arable and semi-natural areas should support arthropods and arthropod-driven

230   services in crops most effectively when travel distances are short (edge density high),

231   promoting spillover between surrounding areas and crops. Further, short travel distances

232   promoting spillover may compensate for scarce arable or semi-natural resources.

233   Consequently, positive effects of edge density on abundance and services in crops may be

234   strongest at low amounts of non-crop habitat (Fig. 1; Holland *et al.* 2016).

235   3. *Effects of landscape variables on arthropods and services are hump-shaped across Europe*

236   (Fig. 1d; Concepción *et al.* 2012). Indeed, resource complementation may be optimal at

237   intermediate habitat amount, but insufficient at high amounts of crop or non-crop habitat

238   (Tscharntke *et al.* 2012). Similarly, edges may facilitate spillover at low to medium density,

239   but hinder dispersal at high edge density due to barrier effects (e.g. in the presence of hedges;

240   Wratten *et al.* 2003) or high spatiotemporal heterogeneity of the agricultural mosaic (Díaz &

241   Concepción 2016). Due to interactions (prediction 2), decreases in abundance or services at

242   extreme values of habitat amount may be lifted under conditions of high edge density, and

243   vice versa (shaded grey areas in Fig. 1d).

244 To date, interactive and non-linear effects of landscape variables on arthropods have rarely

245 been explored, and to our knowledge never in the context of trait-based responses to

246 landscape gradients. We test these predictions for a broad range of taxa and three production-

247 related ecosystem services. We show that the diversity of responses to landscape variables is

248 high among pollinators, enemies and pests, and effects of landscape composition and

249 configuration depend on each other. But overall, high landscape edge density benefitted a

250 large proportion of service-providing arthropods. It was also positive for service provision

251 and harmful for pests, indicating a landscape-scale solution for ecological intensification that

252 does not require setting-aside large amounts of arable land and comes with strong benefits for

253 arthropod functional diversity.

254

255 MATERIAL AND METHODS

256 *Data collection and collation*

257 Data holders were approached through networks of researchers with the aim of collecting raw

258 data from a representative sample of studies performed in European crops. After initial

259 collection, data were screened for missing countries or crops systems, and requests were

260 targeted at researchers having published in these areas. Of 77 proposed studies, 58 provided

261 data with sufficient site replication and high resolution land-use maps (Table S1, Appendices

262 S1, S2 in Supporting Information). Requested data were arthropod abundance per unit area

263 and time (species richness when available) and measures of pollination, pest control and

264 yields, sampled along gradients of landscape composition and configuration in ≥ 8 sites. Sites

265 included annual and perennial crop fields, managed grasslands, field margins and orchards.

266 Farms were conventional, low-input conventional or organic. Data were collated and

267 standardized as described in Appendix S1. After preliminary analyses, we excluded organic

268   sites because few studies compared conventional and organic farms in similar landscapes.

269   This led to a total of 49 studies and 1,637 site replicates from 1,515 distinct landscapes

270   (circular map sectors; Appendix S1, Fig. S1), some sites having been sampled in multiple

271   studies.

272   *Landscape variables*

273   We used land-use maps provided by data holders to calculate landscape variables for all

274   studies. First, we standardized map classification to five land-use classes (arable, forest, semi-

275   natural habitat, urban and water). Semi-natural habitat included hedges, grassy margins,

276   unmanaged grasslands, shrubs, fallows (Appendix S1). We then calculated variables in six

277   circular sectors of 0.1 to 3 km radius around sites (Appendix S1, Fig. S1). Several indices can

278   be used to describe landscape composition, including % arable land and % semi-natural

279   habitat (SNH) (e.g. Chaplin-Kramer *et al.* 2011). To test the importance of these land-use

280   classes, we selected % SNH and % arable land as measures of landscape composition and

281   used them in parallel sets of models to avoid collinearity (see Statistical analyses).

282   Similarly, several measures of landscape configuration exist. Among them, the density of

283   edges available for exchange between landscape patches theoretically underpins mechanisms

284   of spillover and resource complementarity for biodiversity and services (see Introduction),

285   and has been frequently used in other studies (e.g. Holzschuh et al. 2010; Concepción et al.

286   2012). We thus measured landscape configuration as the total length of edges per area of each

287   landscape sector (edge density ED, in km/ha) between crop fields and their surroundings.

288   Hereby, we consider the combined effects of crop / crop (between fields) and crop / non-crop

289   edges (Fig. 1). Both interfaces may enhance arthropod movements in and out of fields

290   (Schellhorn *et al.* 2015). At radii up to 0.5 km, ED is negatively related to mean field size and

291   positively to the density of edges per area of arable land (Fig. S2). Importantly, ED reflects

292   the grain of whole landscapes including non-crop elements and crops. Thus landscapes with

293   high ED have comparatively small fields and non-crop patches. A decrease in ED is related to

294   an increase in size of both field and/or non-crop patches, and reflects a lower total density of

295   edges available for exchange in the whole landscape.

296   *Functional groups and arthropod traits*

297   We classified above-ground arthropods into functional groups of pollinators, pests and natural

298   enemies of pests (Appendix S1, Table S2). Organisms that are predators or herbivores as

299   larvae, but pollinators as adults were classified according to the life stage sampled.

300   Arthropods that could not be classified into these groups (Appendix S1) were included in

301   analyses of total arthropod abundance, as they contribute to overall farmland biodiversity, but

302   not in separate analyses of pollinators, pests and natural enemies (see Statistical analyses).

303   Six categorical traits associated with dispersal mode, overwintering behavior and diet were

304   hypothesized to influence the response of arthropods to landscape variables, as they relate to

305   the need and/or ability to move or disperse between habitat types to access food, hosts,

306   nesting or overwintering resources (Table 1). We defined traits for all arthropod species or

307   families according to the availability of information on separate taxa and to dataset resolution

308   (Appendix S1, Table S2; 36 out of 58 datasets provided species-level identification). We used

309   hierarchical cluster regression to identify parsimonious combinations of shared traits for

310   organisms with shared responses to landscape filters (Appendix S1; Kleyer *et al.* 2012). These

311   combinations are defined as trait syndromes characterizing different responses of species

312   groups to the environment (see Introduction). As trait syndromes may vary according to the

313   functional group (Lavorel & Garnier 2002), we identified them separately for pollinators,

314   natural enemies and pests (Figs. S3, S4). Trait syndromes are defined parsimoniously based

315   on one or a few trait combinations. However, all traits contribute to whole syndrome

316   definition and are described in Figs. S3, S4.

317   *Statistical analyses*

318   We calculated arthropod abundance in each site at three nested levels of community structure

319   (all arthropods; pollinators, enemies and pests; trait syndromes within functional groups;

320   Appendix S1). Pest control, pollination and yields were available from a subset of studies

321   (Table S3). For this subset, we calculated an ecosystem service index representing the amount

322   of service provided (Appendix S1). We analyzed effects of landscape predictors on arthropod

323   abundance and services using linear mixed effects models in R package lme4 v.1.1-15 (Bates

324   et al. 2015). We focused on abundance because it has been found to drive ecosystem service

325   provision (Winfree *et al.* 2015). However, abundance and species richness were positively

326   related across groups (estimates of linear mixed models relating richness to abundance using

327   ln(x+1)-transformed data, with random intercept for study and year: 0.4±0.01, p<0.001 for all

328   arthropods, pollinators and enemies). We ln(x+1)-transformed abundance and services to

329   meet assumptions of normality and homoscedasticity. Predictors were % SNH and % arable

330   land as measures of landscape composition, and edge density as measure of configuration. We

331   expected changes at low values of predictors to have more impact than at high values, thus we

332   ln(x+1)-transformed the predictors. This transformation improved model fits (R^2, see below)

333   and was maintained for all analyses.

334   To account for collinearity of composition variables (Fig. S2), we performed two sets of

335   models including either % SNH or % arable. Correlations between edge density and

336   composition variables were low within and across studies (Fig. S2; mean within-study

337   Spearman rho 0.05, SD 0.2, mean variance inflation factor of models with all arthropods 2.7,

338   SD 1.8), but some studies showed high correlation in specific years and scales (Table S4). We

339   thus ran analyses including and excluding these studies. As no differences were found in

340   overall results, we present analyses including all studies (Appendix S1).

341   Full models took into account hypotheses of a) interactions between landscape variables, and

342   b) non-linearity by including quadratic model terms (Appendix S1). To reflect the ranges

343   covered by European landscape gradients, we did not standardize landscape predictors within

344   studies. In this way we were able to capture non-linear effects across full gradients, i.e. that

345   responses to landscape change within studies may differ across full European gradients in

346   landscape composition and configuration (Van de Pol & Wright 2009). For comparison, we

347   evaluate effects using i) landscape variables mean-centered within studies and ii) standardized

348   response variables in Appendix S3.

349   We accounted for the data's hierarchical structure by including random effects for study and

350   year, sampling method and block (Appendix S1), and scaled predictors across studies by

351   mean-centering and dividing them by two standard deviations (R package arm v.1.9-3,

352   Gelman & Su 2016). We ran separate models at successive scales of 0.1, 0.25, 0.5, 1, 2 and 3

353   km radius around fields. Results at all scales (estimates and boot-strapped 95% confidence

354   intervals [CI] of full models) are presented Figs. S5-7. Figs. 2-4 illustrate results at 1 km

355   radius. We calculated R^2 of the models as the variance explained by fixed (marginal R^2, R^2m),

356   and by fixed and random terms (conditional R^2, R^2c), respectively (Nakagawa & Schielzeth

357   2013). Successive spatial scales are inherently correlated, and results at one scale are likely to

358   be reflected at other scales (Martin *et al.* 2016). In results, we focus interpretation on effects

359   that were significant (CI do not overlap zero) at more than one scale, as these indicate

360   robustness across scales and have the broadest implications for landscape management

361   (Pascual-Hortal & Saura 2007).

362   Few studies sampled all taxa and services in the same sites. To avoid lack of common support

363   for contrasts (e.g. a functional group sampled only in a portion of the overall gradient;

364   Hainmueller *et al.* 2018), we performed separate models for each functional group and

365   service. Replicate numbers for all responses and sites are provided in Tables S5, S6. Residual

366   normality and homoscedasticity were validated graphically. We verified the absence of

367   residual spatial autocorrelation using spline correlograms across studies (Zuur *et al.* 2009).

368   Statistical analyses were performed in R Statistical Software v. 3.4.1 (R Core Team 2017).

369

370   RESULTS

371   *Abundance of arthropods and functional groups*

372   We synthesized effects of landscape predictors on the abundance of 132 arthropod families,

373   encompassing over 494,120 individuals and 1,711 identified species or morphospecies. Of

374   these individuals, 50%, 10% and 37% were classified as natural enemies, pollinators and

375   pests, respectively (44%, 33% and 1% of species; Table S2). Effects of % SNH on arthropod

376   abundance were convex at high edge density (Figs. 2, S5). Effects of edge density depended

377   on % SNH, and led to a 2-fold increase at high (>20%) and 1.6-fold increase at low (<2%)

378   SNH. However, in landscapes with low edge density, increasing % SNH had no effect on

379   arthropod abundance.

380   Pollinators, natural enemies and pests showed distinct patterns when considered separately

381   (Fig. 2). Pollinators showed a similar convex effect of % SNH and a negative effect of %

382   arable land (Fig. S5), but effects were scarce on all natural enemies or all pests. The

383   conditional R^2 of these models was high (mean maximal R^2c across scales 0.80, SD 0.06), but

384   the variance explained by landscape predictors was low (mean maximal R^2m across scales

385   0.04, SD 0.03). However, breaking up these groups into trait syndromes led to further

386   differentiation and a clearer picture.

387   *Trait syndromes of enemies, pollinators and pests*

388   Trait syndromes obtained by cluster regression varied between enemies, pollinators and pests,

389   with the most clusters identified among natural enemies (Figs. S3-4). Though scarce overall,

390   effects of landscape predictors on enemies were significant across scales and highly

391   contrasted between trait syndromes (Fig. 3a, S6). Three main patterns emerged: 1) Enemies

392   overwintering outside crops, including flight and ground-dispersers (327 species, 44% of

393   enemies), benefited from high edge density. This was especially true in landscapes with <10%

394   SNH for flyers, and <60% arable land for ground-dispersers (Fig. 3a, S6). These groups

395   increased with increasing % SNH and decreasing % arable land, but effects depended on edge

396   density: they occurred at low (flight) or high edge density (ground-dispersers). 2) In contrast,

397   enemies able to overwinter in crops were most abundant in landscapes with few edges (Fig.

398   3a, S6). Among these, ground-dispersers benefited from high % arable land, but flyers

399   benefited from high % SNH. 3) Effects of landscape predictors on wind-dispersers, mainly

400   ballooning spiders and parasitoid wasps (flight/wind), were scarce.

401   Different responses also emerged among pollinators. Similarly to all arthropods, non-

402   agricultural specialist pollinators increased with high edge density at high or low % SNH

403   (Fig. 3b, S6; 393 species, 70% of pollinators). In contrast, agricultural specialists (e.g.

404   aphidophagous syrphids) were most abundant in landscapes with few edges and high % arable

405   land.

406   Pests able to overwinter in crops showed few effects of landscape variables across scales. But

407   pests considered to leave crops over winter were six times less abundant in landscapes with

408   high edge density (0.2-0.4 km/ha), regardless of their composition (Fig 3c, S6). Due to an

409   increase beyond this range at intermediate % SNH, 0.2-0.4 km/ha of edges represented an

410   area of minimum pest density along the observed gradients.

411   Marginal R^2 of models including trait syndromes averaged 0.11, SD 0.07 (mean maximal R^2m

412   across scales). Thereby, landscape predictors had significantly higher explanatory power

413   when applied to trait syndromes within functional groups, than to whole groups of natural

414   enemies, pollinators and pests (Wilcoxon rank sum test, W=1289, p<0.001).

415   *Pest control, pollination and yields*

416   Pest control, pollination and yields are given for a subset of studies (Tables S3, S6; Figs. 4,

417   S7). Pest control by natural enemies was highest in landscapes with low % arable land

418   (<40%) and high edge density, where it increased 1.4-fold compared to landscapes with low

419   edge density. It was lowest in coarse-grained landscapes (low edge density) with either low or

420   high % arable land (Fig. 4a). Pollination increased with edge density: it was 1.7 times higher

421   in fine-grained compared to coarse-grained landscapes regardless of % SNH or % arable land.

422   Low pollination was observed in landscapes with >70% arable land and at edge densities <0.1

423   km/ha (Fig. 4b right panel). Yields showed a variable pattern (Fig. 4c, S7). They were highest

424   in landscapes with 10-20% SNH at high edge density (Fig. 4c left panel). Lowest yields were

425   achieved in landscapes with <40% arable land and high edge density (Fig. 4c right panel). In a

426   range of landscapes including a large range of edge density and % arable land, intermediate to

427   high yields were maintained. The variance explained by landscape predictors in models of

428   pest control, pollination and yields averaged 0.14, SD 0.08 (mean maximal R^2m across scales;

429   mean maximal R^2c 0.60, SD 0. 09).

430   Additional analyses show that effects occurred mainly across full gradients instead of within

431   standardized landscape ranges and were robust to standardization of response variables

432   (Appendix S3), as well as to the analytical method chosen (Appendix S4).

433

434   DISCUSSION

435   This synthesis shows that the response of arthropod abundance and services to landscape

436   predictors is non-linear across Europe and depends on interactions between landscape

437   composition and configuration, and on the response traits of arthropods. Overall, arthropods

438   were most abundant in landscapes that combine high edge density with high proportions of

439   semi-natural habitat. Functional groups of pollinators, enemies and pests did not strongly

440   reflect this pattern. Rather, trait syndromes within groups showed contrasting trends.

441   Pollinators that do not feed on pests or crops as larvae (non-pest butterflies, non-

442   aphidophagous syrphids, bees), and flying and ground-dwelling enemies considered to

443   overwinter mainly outside crops, benefited from high edge density at low or high habitat

444   amount and may require a high density of ecotones as exchange interfaces in order to

445   spillover between and into crops (Concepción *et al.* 2012; Tscharntke *et al.* 2012; Hass *et al.*

446   2018). For organisms with limited dispersal ability, this requirement is likely due to the need

447   to recolonize crops in spring. However, the same driver affected strong aerial dispersers such

448   as wasps and butterflies, for which it may be more related to a high sensitivity to disturbance

449   within fields, and/or to the need for resource complementation through a high diversity of

450   available plants and prey (Sutter *et al.* 2017) or nesting sites. Such diverse resources can be

451   found in neighboring semi-natural habitats (e.g. nest sites; Holland *et al.* 2016), but also in

452   adjoining crops (pollen and nectar from crops and weeds, host plants or prey for herbivores

453   and predators). Indeed, a high number of separate field units is the first requirement to support

454   a high diversity of arable crops at organism-relevant scales. Landscapes with high vs. low

455   edge density may also differ in their crop composition and/or diversity, with associated

456   impacts on the arthropod community.

457 In contrast, ground-dispersing enemies with generalist overwintering needs, and pollinators

458 whose larvae feed on crops or pests, were most abundant in landscapes with few edges and

459 high % arable land. These groups benefit from agricultural resources and were able to

460 maintain populations in coarse-grained landscapes with high % arable land that other

461 organisms avoided. They thus represent important insurance organisms contributing to

462 arthropod response diversity (Cariveau *et al.* 2013), and may continue to provide services in

463 coarse-grained landscapes with little non-crop habitat (Rader *et al.* 2016; but see Stavert *et al.*

464 2017). However, abundances were too low for these trends to be reflected in overall patterns.

465 In addition, pests also benefited from landscapes with low edge density. The services

466 provided by agriculture-resilient enemies and pollinators are thus likely insufficient to balance

467 the bottom-up effects of high crop resource availability on pests in such low complexity

468 landscapes (Walker & Jones 2003).

469 Pests overwintering outside crops were least abundant, and pollination and pest control were

470 highest, in landscapes with high edge density, particularly within the range of 0.2-0.4 km/ha.

471 In agreement with Rusch *et al.* (2016), pest control was also highest at low % arable land. But

472 for pests and pollination, edge density effects occurred largely independently of landscape

473 composition. Based on trait syndrome patterns, pest control and pollination appear to have

474 been largely driven by organisms without strong links to agricultural resources, which

475 benefitted from high edge density to spillover and provide services in crops (ground- and to a

476 lesser extent flight-dispersing enemies overwintering outside crops for pest control; non-

477 agricultural specialists for pollination). Due to positive impacts on services and many service

478 providers and negative impacts on pests, edge density thus appeared a more consistent driver

479 for functional biodiversity and service provision than the presence of semi-natural habitat

480 alone (Concepción *et al.* 2012). High diversity of arthropod service providers in such

481 landscapes, confirmed by a positive correlation between abundance and species richness, may

482   further imply functional redundancy. As a result, services supported by these landscapes may

483   be more resilient to environmental change (Oliver *et al.* 2015, Martin *et al.* in press).

484   Landscapes with high edge density did not have lower yields/area than coarse-grained

485   landscapes, in a large portion of composition gradients with varying % SNH and arable land.

486   Though only available from a subset of the data (Table S6), this result indicates that high edge

487   density and its benefits can be combined with maintaining crop yields, within the range of

488   edge density observed here. Accordingly, productive landscapes with edge density between

489   0.2 and 0.4 km/ha may be ideally suited to implement ecological intensification. Cascading

490   (positive) effects on yields of higher service provision and less pests in landscapes with high

491   edge density were not, however, apparent from the available data. Reduced pollination and

492   pest control at low edge density may have been compensated by external inputs in productive

493   landscapes. In addition, other factors combine to impact yields (Gagic *et al.* 2017) and may

494   mask the impact of biodiversity-driven services in the absence of careful standardization

495   (Pywell *et al.* 2015). Intermediate to low yields in landscapes with high % arable, low % SNH

496   and low edge density may underpin the risks of ongoing conventional intensification resulting

497   in yield stagnation or reduction despite high agricultural inputs (Ray *et al.* 2012).

498   Non-linear and interacting effects of landscape predictors denote the importance of variation

499   in the ranges occupied by European landscape gradients between studies. In combination with

500   trait-based response syndromes, these results explain several inconsistencies highlighted in

501   previous work (Kennedy *et al.* 2013; Veres *et al.* 2013; Díaz & Concepción 2016; Holzschuh

502   *et al.* 2016; Rader *et al.* 2016; Tscharntke *et al.* 2016; Karp *et al.* 2018). By covering a wide

503   range of landscapes and responses, this study helps resolve why responses to landscape

504   configuration and composition of arthropod functional groups differ along landscape

505   gradients. In particular, we show that landscape effects and the potential effectiveness of

506   landscape management measures vary according to the ranges of landscape variables captured

507   in each study region, in agreement with theory underlying non-linear responses of organisms

508   to landscape gradients (Concepción *et al.* 2012). Increasing edge density was most effective

509   for arthropods in landscapes with low (<5%) or high (>20%) % SNH. In landscapes with

510   intermediate % SNH, small increases in SNH may dilute populations, evening out the benefits

511   of many edges, before reaching sufficient levels to contribute positively to spillover into

512   fields. In these landscapes, extensive practices such as low-input farming may be the most

513   effective way to enhance arthropod diversity and services in crops (Jonsson *et al.* 2015).

514   Contrary to our hypotheses (Fig. 1), few effects were hump-shaped within the range of tested

515   gradients, thus maxima may not be reached within the measured European gradients.

516   We applied a trait-based framework for agroecosystem communities using response traits that

517   have not been considered in previous work on pollinators (Williams *et al.* 2010; De Palma *et*

518   *al.* 2015; Carrié *et al.* 2017) or grassland arthropods (Gámez-Virués *et al.* 2015), but were

519   important determinants of species' responses to landscape structure. We found that syndromes

520   combining several response traits effectively disentangled pollinator, pest and enemy

521   responses compared to single-trait approaches. Considering such traits with strong

522   mechanistic underpinnings (Bartomeus *et al.* 2018) will increase our ability to derive

523   predictions of the effects of environmental change on communities. Clarification is needed,

524   however, on which trait syndromes correlate with strong impacts on service provision in

525   crops. For instance, non-bees may complement bees for provision of pollination services

526   (Rader *et al.* 2016), but the separate contribution of non-bee pollinators in intensive

527   landscapes is unknown, and according to our results, may be considerably lower. In addition,

528   relative contributions to pest control of natural enemies with different landscape responses,

529   and the importance of high enemy diversity for pest control in real-world landscapes, have yet

530   to be elucidated.

531   *Conclusion*

532   In this synthesis across Europe, we show that within European gradients, a high edge density

533   is beneficial for a wide range of arthropods and the services they provide, and can be

534   combined with high yields in productive landscapes with over 50% arable land. In addition to

535   managing semi-natural habitat amounts, increasing the edge density of these landscapes is a

536   promising pathway to combine the maintenance of arthropod biodiversity and services with

537   continued and sustainable agricultural production. While the strength of these effects for

538   arthropods depends on habitat amount, fine-grained landscapes provided benefits such as less

539   pests and more pollination, which were largely independent of their composition. We further

540   demonstrate a high response diversity of arthropod service providers leading to differing

541   impacts of landscape change within groups of natural enemies, pests and pollinators. We thus

542   call for consideration of mechanism-relevant response traits to catalyze modelling and

543   prediction of the consequences of land-use change on arthropods and ecosystem services in

544   crops.

545

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560

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772   Table 1. Functional response traits included in cluster analyses. Details on classification of

773   traits for all organisms are provided in Appendix S1, Table S2. The full database of traits for

774   all species is published at [doi://].

Trait name	Trait level	Abbreviation	Description
Diet breadth	specialist	(specialist)	Diet restricted to no more than two families of host plants or prey*
	generalist	(generalist)	Generalist diet including a broad range of families
Agricultural specialism	yes	(agsp)	Diet specialists for which hosts or prey are agricultural (crops or pests)
	no	(non agsp)	Diet generalists or diet specialists for which hosts or prey are not agricultural
Diet life history	same diet	(same diet)	Organisms have a similar diet across their life cycle
	different diet	(diff. diet)	Organisms switch diets between life stages (e.g. carnivore larva to nectarivore adult)
Overwintering habitat	crop	(crop)	Organisms may overwinter in or outside crops
	non crop	(non crop)	Organisms overwinter mainly outside crops
Dispersal	ground	(gd)	Dispersal by moving on the ground (wingless or undeveloped wings)
	flight	(flight)	Dispersal by active flight (organisms with developed wings)

	flight / wind	(fl/wind)	Active flyers known to disperse on wind currents
	wind	(wind)	Dispersal by wind or electrostatic currents (ballooning spiders)
Stratum	ground / vegetation	(ground/veg)	Forages by walking or web-building on the ground or in vegetation
	aerial	(aerial)	Forages by flying between target hosts

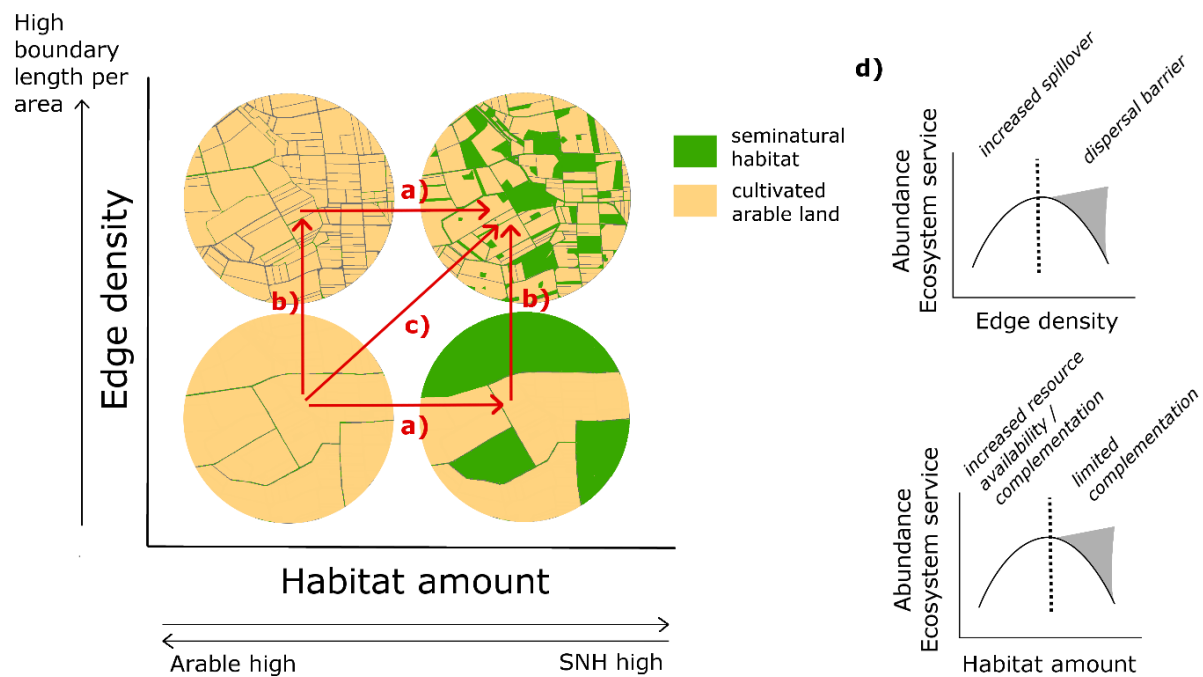775   * Diet restricted to one larval 'microhabitat' for hoverflies; see Appendix S1.

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Fig. 1. Conceptual representation of the distinction between landscape composition and

configuration and their possible effects. Four theoretical farming landscapes are viewed from

above (left panel). **a) Landscape composition (increasing habitat amount):** an increase in

the proportion of seminatural habitat (SNH) is reflected, in the absence of forest, by a

decrease in the proportion of cultivated area as arable land is taken out of production. **b)**

Landscape configuration (increasing edge density): for the same total amount of crop and

non-crop habitat, decreasing patch sizes and complex shapes lead to an increase in the length

per area of edges (ecotones) among crop fields and between crop and non-crop habitat. **c)**

Simultaneous increase vs. interactions: habitat amount and landscape edge density may

increase simultaneously, making it difficult to disentangle the contribution of each to

biodiversity and ecosystem services. However, these variables are not intrinsically correlated.

In addition, interactions may take place that lead to different effects of edge density according

to the relative proportion of crop and non-crop habitats in a landscape. **d) Non-linear effects:**

we hypothesize that effects of landscape composition and configuration on abundance and

services are unimodal, with different mechanisms operating at each end of the gradients. For

800   instance, low resource complementarity due to high amounts of semi-natural habitat, but little

801   cropland may decrease arthropod abundance and/or ecosystem services if organisms benefit

802   from both resource types (e.g. pollinators that benefit from flowering crops or enemies that

803   feed on pest prey, but require resources from semi-natural habitats for reproduction).

804   However, the shape of curve tail ends (grey area) should depend on the state of other

805   variables. For instance, constraints on resource complementation when habitat amount is high

806   should be lifted when edge density increases, facilitating spillover.
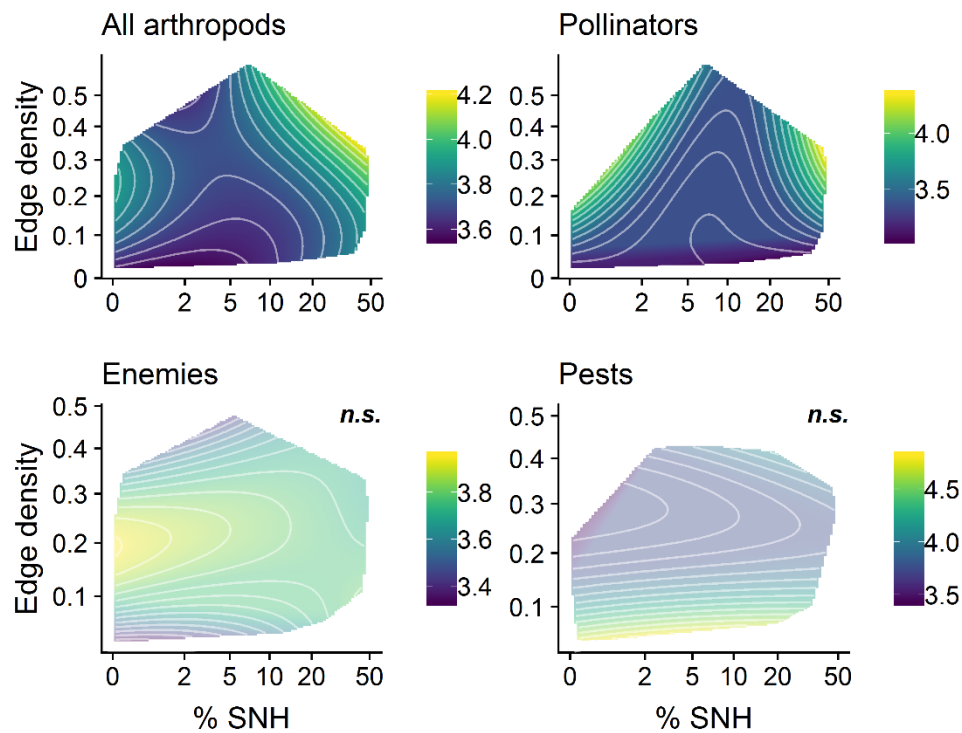
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Fig. 2. Heatmaps of the effects of seminatural habitat amount (SNH; composition variable)

and landscape edge density (ED in km/ha; configuration variable) on the abundance of

arthropods (top left) and on functional groups of pollinators, natural enemies, and pests. The

heatmaps can be read like a topographic map, with yellow peaks and blue valleys, and steeper

slopes where line density is high. Yellow indicates areas of highest abundance, blue areas of

lowest abundance (see ln(x+1)-transformed abundance scale at the right of each panel).

Estimates and 95% confidence intervals (CI) of effects are shown for all radii in Fig. S4.

Results at 1 km radius are shown here. Results are not interpreted (marked 'n.s.' and faded) if

significant effects were obtained at less than two out of six tested radii. Only the area

covering the range of both variables for each response is plotted. Note a log-scale of predictor
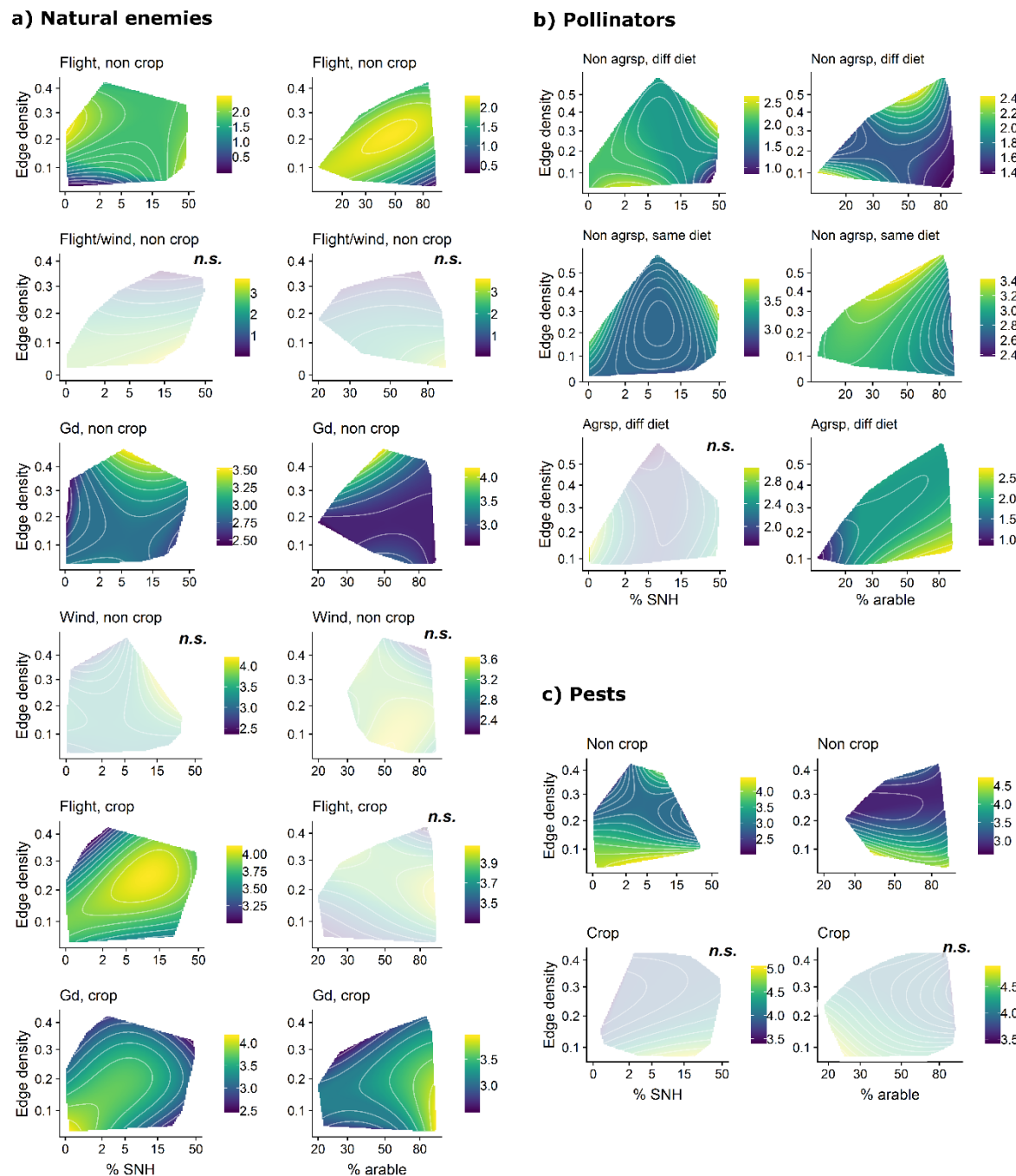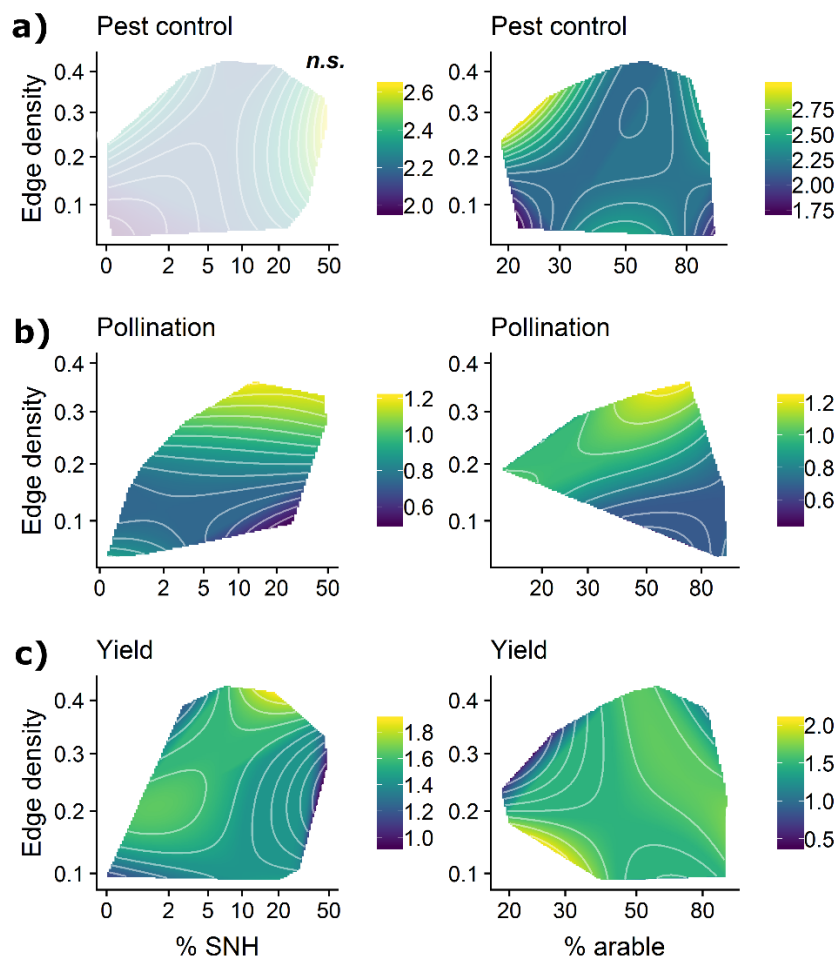
variables.

Fig. 3. Heatmaps of the effects of landscape composition (% SNH, left columns; % arable land, right columns) and landscape configuration (edge density in km/ha) on the abundance of functional response groups of a) natural enemies, b) pollinators, and c) pests. Functional groups were separated into trait syndromes based on cluster regression of six categorical traits (see abbreviations in Table 1; Figs. S2-3). Estimates and 95% CI are shown at all radii in Fig. S5; results are shown here at the 1 km radius. See further graph details in the legend of Fig. 2.

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Fig. 4. Heatmaps of the effects of landscape composition (% SNH, left columns; % arable land, right columns) and landscape configuration (edge density in km/ha) on a) pest control, b) pollination and c) crop yield in weight per unit area. Response variables represent an ecosystem service index accounting for differences in methods within and between studies (see Appendix S1). See Table S3 for detailed units and measurements per study. Blue: lowest service provision; yellow: highest service provision. Estimates and 95% CI are shown at all radii in Fig. S6; results are shown here at the 1 km radius. See further graph details in the legend of Fig. 2.