Modelling the interplay between inadvertent social information use and a pesticide-induced foraging bias in bumblebees' crop visitation

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Abstract

Understanding the drivers of patch visitation in agricultural landscapes is essential for maintaining pollinator-mediated agroecosystem services, yet the extent to which pollinators' social and stress-altered foraging behaviour influences visitation rates has received little attention. Using an individual-based modelling framework (BEE-STEWARD), I examined how inadvertent social information (ISI) use and a recently documented pesticide-induced foraging bias influence nectar and pollen visitation rates of buff-tailed bumblebees (*Bombus terrestris*) to crop patches located near or far from the colony. Simulations showed that ISI use primarily increased nectar visitation to all OSR patches and pollen visitation to nearby crop patches, whereas foraging bias increased pollen visitation to all OSR patches and nectar visitation to nearby OSR patches. However, the two behavioural parameters had no interactive effects on crop visitation rates during nectar or pollen foraging. Despite the pronounced shifts in visitation patterns, total resource collection and colony demographics, including colony growth and reproductive output, remained largely unaffected. Overall, results indicate that ISI use and foraging biases represent behavioural pathways that are likely to alter exposure risk to pesticides by altering spatial foraging patterns, and thus may facilitate the emergence of pesticide-induced sublethal effects in bumblebees. Incorporating such pathways into pesticide risk assessment frameworks may improve current predictions of exposure routes in these pollinators and support more effective mitigation strategies in pesticide-treated agricultural landscapes.

Keywords: inadvertent social information, foraging behaviour, pesticide exposure, sublethal effects, individual-based modelling, *Bombus terrestris*

Introduction

Social information provided by conspecifics or heterospecifics (Danchin et al. 2004; Seppänen et al. 2007) has a crucial role in adaptation to predation risk (Coolen et al. 2005; Martín et al. 2006; Pays et al. 2013; Cruz et al. 2020), human disturbance (Banks et al. 2013; Schmidt et al. 2015; Gruber et al. 2019), or resource exploitation (Laidre 2010; Webster and Laland 2017; Pouca et al. 2020; Curk et al. 2025) in a wide variety of animals. Social information can originate either from evolved, actively produced signals, or from inadvertently produced cues. Inadvertently produced cues (visual, acoustic, chemical, etc.) include the presence, routine behaviour, or the product of the behaviour of others that convey information to the observers about the environment or the physiological state of the cue producing individual (Tóth et al. 2020; Gil et al. 2018). Unlike social learning, inadvertent social information (ISI) use is not characterised by the presence of a conditioning phase (Leadbeater 2015), and not necessarily results in the acquisition of new knowledge or skill. ISI use, nevertheless, often generates strong correlations in the behaviours and space use of multiple individuals (Laundré et al. 2010; Gil et al. 2017), making the social environment capable of shaping population dynamics and even ecosystem structure and function (Parejo and Avilés 2016; Gil and Hein 2017; Gil et al. 2018, 2019). ISI use can also contribute to the spread of maladaptive behaviours, potentially accelerating population declines and increasing extinction risk through information-mediated Allee effects (Kokko and Sutherland 2001; Donaldson et al. 2012; Sigaud et al. 2017; Wilson et al. 2020).

ISI use is a well-known phenomenon in pollinating insects (Grüter and Leadbeater 2014). Bumblebees, like many bee species (Yokoi et al. 2007; Dawson et al. 2013), often rely on social cues when they do not have personal information about surrounding food sources or finding them would be more costly (Kawaguchi et al. 2006; Jones et al. 2015). These pollinators have remarkable cognitive abilities and social learning has been demonstrated in various contexts in these model animals (e.g., Loukola et al. 2017). However, the possibility of social learning does not preclude ISI use and several studies have proved its fundamental role in bumblebees' foraging (e.g., Baude et al. 2008; Leadbeater and Florent 2014; Smolla et al. 2016). Under natural conditions, ISI use is often advantageous as it facilitates the localisation and exploitation of new food sources (Worden and Papaj 2005; Avarguès-Weber and Chittka 2014) and helps to avoid predators (Dawson and Chittka 2014). The presence of conspecifics can also provide a cue about the quantity of resources in the visited flowers (Leadbeater and Chittka 2009). However, relying on social information may also limit the time spent searching for alternative food sources and making independent foraging decisions,

resulting in suboptimal foraging (Avarguès-Weber et al. 2018). In agricultural habitats, the

attraction of conspecifics to pesticide-treated food patches may lead to increased pesticide

exposure. Frequent foraging at such patches can, in turn, lead to pesticide-induced sublethal

effects, impairing foragers' behaviour or physiology and, ultimately, reducing colony growth

or fitness (e.g., Baron et al. 2017; Crall et al. 2019). Interestingly, the effects of ISI use on

crop visitation patterns and pesticide exposure risk are largely understudied in most pollinator

species and remained unaddressed in recent theoretical models (e.g., Blasi et al. 2022;

Carturan et al. 2023; Lonsdorf et al. 2024).

The quantity of pesticide residues found in agricultural landscapes is highly

inconsistent, and it is not well understood if pollinators can influence their exposure risk when

pesticide contamination varies spatially and temporally (Olaya⎕Arenas et al. 2020). Evidence

suggests that some pollinating insects, such as honeybees and bumblebees, cannot detect even

lethal concentrations of pesticides in nectar (Kessler et al. 2015; Parkinson et al. 2023), so

foraging individuals cannot behaviourally mitigate their pesticide exposure level by avoiding

contaminated food sources (but see Olaya⎕Arenas et al. (2020) for a lepidopteran

counterexample). Evidence suggests that pesticide exposure can alter individual foraging

behaviour. Such pesticide-induced behavioural biases may manifest in preference to pesticide-

treated food sources (Kessler et al. 2015; Arce et al. 2018), impaired cognitive abilities and

feeding motivation (Siviter and Muth 2022; Gray et al. 2024), decreased resource collection

(e.g.; Lämsä et al. 2018; Muth and Leonard 2019), or reduced ability to orient toward a

reward-associated floral scent source (Kárpáti et al. 2024). Such behavioural changes may, in

turn, result in an elevated exposure risk by contributing to higher visitation rates of those food

patches that are more likely to receive – directly or indirectly (Ward et al. 2022) – regular

pesticide treatments (Pohorecka et al. 2012; Byrne et al. 2014). However, the potential

contribution of such behavioural pathways to pesticide exposure risk in pollinators remains

largely unexplored.

In this study, I extended the existing individual-based model of BEE-STEWARD

(Twiston⎕Davies et al. 2021) to explore how ISI use and a recently documented pesticide-

induced foraging bias influence crop patch visitation rates in buff-tailed bumblebees (*Bombus*

terrestris) in a simulated agricultural landscape. This species is an important pollinator of

both crops and wildflowers in temperate regions (Hutchinson et al. 2021), and a common

visitor of the mass-flowering oilseed rape (*Brassica napus*; 'OSR' henceforward), a dominant

oil crop in Europe (Perrot et al. 2024). By including the possibility of ISI use among foragers

and reduced ability to locate rewarding food sources as a pesticide-induced foraging bias

(following Kárpáti et al. 2024) in the model, I predicted that bumblebees will exploit the

available food sources more efficiently in the presence of ISI use, leading to a shift in patch

visitations and more resources collected compared to the original model setting. In the

presence of foraging bias, a shift in patch visitations accompanied by reduced foraging

efficiency is expected, the extent of which may be amplified by the presence of ISI use by

attracting other foragers to suboptimal resource patches. In this study, I simulated the

attraction of conspecifics to the same resource patches, which does not involve the acquisition

of a new knowledge (i.e., it is not social learning) or the exploitation of evolved signals (i.e., it

is not communication), but falls within the definition of ISI use (i.e., the exploitation of social

cues originating from the routine behaviour of others).

Materials and methods

Description of the applied model

Simulations were run in BEE-STEWARD (www.beesteward.co.uk; Twiston☐Davies et al.

2021). This simulation tool incorporates landscape data and the full life cycle of bumblebees,

allowing to test the impact of land-use characteristics and environmental stressors on colony

and population dynamics in various bumblebee species. Individual behaviour is determined

by stimuli and thresholds that scale up to colony-level processes, enabling the simulation of

population survival and performance on the landscape level over multiple years. The

underlying parts of the software are parameterised with empirical data from the UK,

describing colony dynamics of six bumblebee species and plant species characteristics,

including details on flower shape, resource production, and flowering period. The model also

incorporates energy budgets and depletion of resources to facilitate the emergence of

competition both within and between pollinator species. With sensitivity analysis and

validation, the incorporated Bumble-BEEHAVE model has been demonstrated to make

realistic predictions for buff-tailed bumblebee colonies within the boundaries of independent

empirical observations (Becher et al. 2018). For a detailed description of the underlying

models and procedures, see Becher et al.'s (2016, 2018) supplementary materials (Appendices

S02 and S05, respectively).

In short, the landscape is 2-dimensional and comprises patches that provide foraging

and/or nesting opportunities. Patches are spatially explicit and implemented as one or more

food sources, each representing a single plant species. If a patch contains multiple food

sources, these can be layered, i.e., placed at the same location and linked with each other. The

simulations start at the beginning of the year with an initial number of hibernating queens

that, after emerging from hibernation, search for suitable nest sites in semi-natural habitats

(exact dates drawn from a characteristic distribution). Queens that survived the searching

process and successfully founded a colony must collect resources to lay the first batch of

eggs, then forage and care for the brood simultaneously until the first adult workers emerge.

After that, the queen's activity is limited to egg laying, while the workers are responsible for

nursing and foraging. Foragers' choices on which patch to exploit are based on maximising

the foraging rate for pollen and the energetic efficiency for nectar, primarily determined by

the distance from the colony, handling time, and the amount of the given resource available in

the patch (Knapp et al. 2019). Colony growth depends on resource collection, whereas

population dynamics result from the number of mated queens that leave the colony and enter

into hibernation until the following spring (Becher et al. 2018).

Simulated landscape with specific foraging patches

The simulated landscape, consisting of 300×210 grid cells, contains one grassland and eight

oilseed rape (*Brassica napus*) patches (Fig. 1). The central grassland patch provide nectar and

pollen for the bumblebees' year-around and also serves as a nesting habitat where queens can

establish new colonies (randomly distributed within the grassland patch) after emerging from

winter hibernation (Figure S1). This patch is a good-quality habitat for buff-tailed bumblebees

and contains 22 different flowering plant species (Becher et al. 2018). The OSR patches

represent habitats that are unsuitable for nesting and solely consist of oilseed rape crop that

flowers from day 114 until day 136 in each simulation year. All patches have a 200 m radius,

resulting in a 125 663.7 m^2 patch area. Due to the spatial distribution of OSR patches around

the grassland patch, four OSR patches are closer to the central patch (distance between patch

edges: 15 m) than the diagonally located OSR patches (186.9 m). The former patches are

termed 'close OSR patches', whereas the latter 'far OSR patches'; the exploitation of close

and far OSR patches was examined separately.

Model extensions

I modified the original code by incorporating the parameters 'ISI use' and 'foraging bias' into

the model (Table 1). The model does not explicitly simulate pesticide exposure level as it

would require detailed information on pesticide concentrations in nectar and pollen and their

temporal dynamics in treated crops, which remained poorly characterized for many pesticides.

Instead, it simulates two behavioural pathways that may influence patch visitations, which, in

turn, has been documented to be positively related to pesticide exposure risk in several

empirical studies (e.g., Pohorecka et al. 2012; Byrne et al. 2014; Arce et al. 2018). I also

included additional functions in the extended version to obtain patch-specific nectar and

pollen visitation data. To reduce the computational time of the simulations, the code of the

original model was converted from NetLogo 5.3.1 to 6.4.0 (Wilensky 1999).

In the original foraging module, the model sorts the list of known foraging patches by

their distances to the bee's colony for each forager. In the presence of 'ISI use', foragers'

choice of where to collect nectar or pollen is influenced by the presence of conspecifics.

Individuals' known patches are first divided into two types: one type is where others are

currently foraging, and the other type is the unoccupied patches. Then, both lists are sorted by

the distances of patches and the bee's colony, and the lists are compiled so the occupied

patches appear first in the compiled list (see the 'Foraging_SortKnownPatchesListREP'

function in the extended code). The modified function is later called during both nectar and

pollen foraging. Because of the way ISI use is coded, freshly emerged foragers are not

affected by this parameter until they find at least two accessible patches (please note that

foragers become experienced by acquiring knowledge of more and more foraging patches

during the simulations). Thus, naïve individuals will deplete food sources within the grassland

patch first, because these are always the closest to the nest, which assumption corroborates

with empirical findings indicating that young workers fly much shorter distances than older

ones (Gilgenreiner & Kurze 2024). Foragers can be attracted to the OSR patches by others

only after they discovered those patches by themselves. Because of that, the contribution of

ISI use to resource exploitation can be regarded conservative in the extended model.

The foraging bias included in the extended model is based on a recent finding of

Kárpáti et al. (2024). This study examined the peripheral olfactory detection of a floral blend

and foraging behaviour of buff-tailed bumblebee workers exposed to a formulation containing

acetamiprid, a neonicotinoid insecticide. The key result was that treated bumblebees were less

likely to locate the floral blend source that was previously associated with syrup during

chronic pesticide exposure, which led to a random choice between the control and reward-

associated scent sources in the behavioural trials. Acetamiprid is frequently applied in OSR

cultivations, particularly during the flowering period (Kathage et al. 2018; Dworzańska et al.

2020). As floral scent is generally associated with the presence of resources in flowers

(Wright and Schiestl 2009; Knauer and Schiestl 2015), this finding suggests that exposed

individuals could not correctly assess which presented scent source was rewarding. In BEE-

STEWARD, floral scent cues are not represented explicitly; instead, the profitability of a food

source is indicated by its resource filling level. To represent the above experiment finding, I

implemented a pesticide-induced 'foraging bias', in which affected bees choose among patches independently of filling level. This mechanism can be interpreted as an abstraction of reduced ability to use reward-associated cues during patch choice, rather than as a direct preference for any particular crop types. I modelled this effect by modifying the 'Foraging_bestLayerREP' function that reports the most profitable layer within the currently used flower patch based on minimal handling time. In the original model, handling time ('handlingTimePerFlower_s' function) for nectar foraging is determined by the time needed to travel to the next flower plus the time to test whether it contains nectar, divided by the filling level of the flower, plus the time needed to collect the nectar once a filled flower is found. During pollen foraging, handling time is calculated from the time needed to travel to the next flower multiplied by the number of flowers needed to be visited, divided by the filling level of the flower, plus the time needed to collect the pollen once a flower with pollen is found. In the presence of 'foraging bias', filling levels are set to 1 within foragers' decision-making process, so handling times are not dependent on actual filling levels when bumblebees are exposed to a pesticide on the crop patches. Still, only those patches are considered where the filling level > 0, i.e., flowering plants are present. I assumed that all bees are affected by the pesticide application on OSR patches irrespective of the patch that individuals are exploiting (as resources stored within the colonies can also be contaminated), but only during the OSR flowering period (i.e., from day 114 to day 136).

Characteristics of the simulation runs

Simulations were run over 50 years in 50 iterations to determine if the default number of total hibernating queens (set to 500) is appropriate for the model and from which year the population can be assumed to be in equilibrium (Hui 2006; Fig. S1A). Based on the results, I took year 10 as the year when the bumblebee population was likely to reach equilibrium in the simulated landscape and thus subsequently analysed the data obtained in the 10th year. The initial number of hibernating queens was also 500 in all iterations. I ran the simulations 50 times in all combinations of 'ISI use' (absent or present) and 'foraging bias' (absent or present), totalling 200 simulations. Models were run individual-based under the 'intermediate foraging mortality' scenario. I also used the setting where males were not unlimited.

Data preparation

From the simulation outputs, I obtained the following measures from the 10th simulation year for subsequent analysis: the number of visits associated with nectar foraging on each patch

between day 114 and day 136 (i.e., during the OSR flowering period), the number of visits

related to pollen foraging on each patch between day 114 and day 136, the quantities of nectar

and pollen collected and stored in all colonies during this period, the total number of adult

workers and number of colonies on day 149, and the total number of hibernating queens on

day 365. The total number of adult workers and the number of colonies variables were used as

proxies for the pollination capacity of the simulated bumblebee population, as this time point

falls within the period of peak annual population size. The total number of hibernating queens

was considered as a proxy for fitness. These approximations are consistent with those applied

in a similar simulation in BEE-STEWARD by Knapp et al. (2019). From the patch visits data,

I calculated the proportions of nectar and pollen visits for the close and far OSR patches,

respectively. For the distribution of total nectar and pollen visits between day 114 and day

136, see Fig. S2.

Statistical analysis

All calculations were performed using R 4.5.0 (R Core Team 2025) and RStudio 2025.05.0

(Posit Team 2025). Instead of applying frequentist hypothesis testing during the analysis, I

evaluated the magnitude of differences between simulation runs with different parameter

settings and reported parameter estimates with corresponding uncertainty following the

recommendations of White et al. (2014). I fitted generalised linear models (GLMs) with a

beta binomial error distribution using the Template Model Builder ('glmmTMB' R Package;

Brooks et al. 2017) with ISI use (as a factor), foraging bias (as a factor), and their two-way

interaction as explanatory variables on the proportion of nectar and pollen visits, separately on

the close and far OSR patches. With this design, I could explore whether the effects of these

two processes are independent, additive, multiplicative, or whether they mitigate each other.

In the model fitted on the proportion of pollen visits on the close OSR patches, ISI use was

also included in the dispersion formula to account for variance differences between its levels.

As a decision framework for evaluating differences in these models, I calculated odds ratios

(OR) with corresponding 95% CIs as measures of effect size. To assess the impact of ISI use

and foraging bias on the amount of collected resources during the OSR flowering period, I

fitted linear models (with Gaussian error distribution; LMs) on the amount of stored nectar

and pollen. To investigate the potential consequences of these behavioural parameters on

pollination capacity and fitness, I also fitted LMs on the total number of adult workers,

number of colonies, and total number of hibernating queens. Model assumptions were

checked by plot diagnosis using the 'DHARMa' R package (Hartig 2022). I calculated

estimated marginal means (EMMs) and contrasts on the response scale with 95% CIs using

the 'emmeans' R package (Lenth 2023). For LMs, effects were considered negligible when

the 95% confidence interval of these contrasts included zero.

Results

Nectar visits on close and far OSR patches

In the baseline setting, in which neither ISI use nor foraging bias was included, bumblebees

visited the close OSR patches in 26.8% [25.6, 28.1] of the nectar foraging bouts (Fig. 2a).

Incorporating ISI use was associated with a higher estimated proportion of visits to these

patches (37.6% [36.2, 39.0]; contrast – presence vs. absence: 10.8 [8.9, 12.7]), and the

estimated odds of visiting close OSR patches for nectar were higher by 65% (OR = 1.65

[1.51, 1.80]). Including foraging bias alone also increased the estimated proportion, although

the magnitude of the contrast was smaller (31.1% [29.7, 32.4]; contrast: 4.2 [2.4, 6.1]; OR =

1.23 [1.12, 1.35]). When both behavioural parameters were included, the estimated proportion

was similar to the 'only ISI use' setting (37.6% [36.1, 39.1]; contrast to the 'only ISI use'

setting: <-0.01 [-2.1, 1.9]; OR = 1 [0.92, 1.08]).

On the far OSR patches, the estimated effects of ISI use and foraging bias followed a

similar directional pattern but with smaller contrasts. In the absence of both behavioural

parameters, the proportion of nectar visits was 7.1% [6.6, 7.7] (Fig. 2b). When ISI use was

included, the estimated proportion increased to 8.5% [7.9, 9.1] (contrast – presence vs.

absence: 1.3 [0.5, 2.2]; OR = 1.21 [1.08, 1.35]). When only foraging bias was present, the

estimate was 7.4% [6.9, 8.0] (contrast: 0.3 [-0.5, 1.1]; OR = 1.05 [0.93, 1.17]). When both

behavioural parameters were incorporated, the estimated proportion was similar to that

observed with ISI use alone (8.1% [7.5, 8.7]; contrast to the 'only ISI use' setting: -0.3 [-1.2,

0.5]; OR = 0.96 [0.86, 1.07]).

Pollen visits on close and far OSR patches

In the absence of ISI use and foraging bias, the proportion of pollen visits to the close OSR

patches was 10.9% [9.3, 12.8] (Fig. 3a). When ISI use was included, the estimated proportion

increased to 14.9% [13.3, 16.6] (contrast – presence vs. absence: 4.0 [1.5, 6.4]; OR = 1.43

[1.14, 1.78]). When only foraging bias was present, the proportion of pollen visits to the close

OSR patches was almost three times higher than the baseline value (31.0% [28.4, 33.7];

contrast: 20.1 [16.9, 23.2]), corresponding to 3.66-fold higher odds of visitation (OR = 3.66

[2.95, 4.54]). When both behavioural parameters were included, the estimated proportion

further increased to 34.3% [32.0, 36.6], corresponding to a contrast of 3.3 [<-0.1, 6.8] relative

to the 'only foraging bias' setting (OR = 1.16 [0.99, 1.36]).

On the far OSR patches, the estimated proportion of pollen visits was 2.9% [2.4, 3.6]

in the absence of both behavioural parameters. When only ISI use was present, the estimated

proportion was 3.8% [3.1, 4.6] (contrast: 0.8 [-0.1, 1.7]; OR = 1.3 [0.98, 1.71]). When

foraging bias was included, the proportion increased to 10.6 % [9.4, 11.9] (contrast – presence

vs. absence: 7.6 [6.3, 9.0]; Fig. 3b), and the odds of visiting far OSR patches for pollen were

3.9 times higher (OR = 3.9 [3.05, 5.0]). When both behavioural parameters were included, the

estimate was 9.1% [8.0, 10.3], corresponding to a contrast of -1.4 [-3.1, <0.01] relative to the

'only foraging bias' setting (OR = 0.85 [0.7, 1.03]).

Resources collected during the OSR flowering period

The amount of nectar stored in the baseline setting was 146 mL [127, 165]. When ISI use was

included, the estimated amount was 170 mL [152, 189] (contrast – presence-absence: 24.8 [-

1.8, 51.3]), and when only foraging bias was present, the estimate was 169 mL [150, 188]

(contrast: 23.6 [-3, 50.1]). When both behavioural parameters were included, the estimated

amount was 167 mL [148, 186] (contrast relative to the baseline setting: 21.5 [-5.1, 48.1]).

The amount of pollen stored in the baseline setting was 60.8 g [54.6, 67.1]. When ISI

use was present, the estimated amount was 70.9 g [64.6, 77.1] (contrast – presence vs.

absence: 10 [1.2, 18.9]). When only foraging bias was included, the estimate was 66.5 g [60.2,

72.8] (contrast: 5.7 [-3.2, 14.6]). When both parameters were included, the estimated amount

of pollen stored was 63.7 g [57.5, 70.0] (contrast relative to the baseline setting: 2.9 [-5.97,

11.8]).

Consequences on pollination capacity and fitness

In the baseline setting, the simulated landscape contained a total of 373 adult workers [336,

409], 37.3 colonies [33.7, 40.9], and 452 hibernating queens [420, 485]. The estimated

numbers of adult workers and hibernating queens were similar across scenarios that included

ISI use, foraging bias, or both behavioural parameters. When both ISI use and foraging bias

were present, the estimated number of colonies was slightly higher than in the baseline setting

(contrast relative to the baseline setting: 6.1 [1, 11.2]; Table 2).

Discussion

In non-*Apis* bees, the extent to which ecological traits, including foraging habits, influence

pesticide exposure and shape its subsequent effects on foragers is still poorly understood

(Raine and Rundlöf 2024). By incorporating inadvertent social information (ISI) use and a

recently documented pesticide-induced foraging bias into the BEE-STEWARD model, I

found that patch visitation patterns of buff-tailed bumblebees shifted relative to the original

model. Specifically, ISI use enhanced nectar visitation to both close and far OSR patches and

also increased pollen visitation to the close crop patches. Foraging bias primarily increased

pollen visitation to all OSR patches and also increased nectar visitation to close OSR patches.

I have found no evidence of an interactive effect between the two parameters on crop patch

visitations. The observed shifts in patch visitation did not alter the amount of resource

collected or induced comprehensive changes in colony growth parameters and reproductive

performance. Overall, these simulations show that in a crop patch-dominated landscape where

the bumblebee population is at equilibrium, both behavioural parameters can increase

foragers' resource collection rate from crop patches, despite the availability of alternative

food sources during crop bloom (also in Carturan et al. 2023). The presented model does not

explicitly simulate pesticide residues, intake, or dose accumulation, and therefore does not

provide quantitative estimates of pesticide exposure levels. Nevertheless, it quantifies

visitation patterns to different patch types, which may influence the probability of exposure

when floral resources are treated with pesticides. Consequently, the model captures two,

previously overlooked behavioural pathways that are likely to alter exposure risk in these

pollinators.

360   Bumblebees are central place foragers (Heinrich 1979), and most recent models have

simulated their foraging behaviour as primarily depending on the distance to nest and floral

resource quality (Olsson et al. 2015; Häussler et al. 2017; Blasi et al. 2022; Carturan et al.

2023; Lonsdorf et al. 2024). These variables are regarded as key drivers of bumblebees'

foraging decision because they can accurately predict resource partitioning especially at larger

spatial scale and are well-supported by empirical data (Lihoreau et al. 2011; Pope and Jha

2018; Fragoso et al. 2021). Nevertheless, findings of this study indicate that considering

foraging decisions at a finer resolution and taking additional behavioural pathways into

account can be crucial for accurately estimating patch visitation patterns in bumblebees.

Conspecific attraction among foragers has been proposed to play a key role in the exploitation

of patchily distributed resources (Dubois et al. 2021), whereas pesticide-induced foraging

biases have not yet been integrated into similar modelling frameworks. Although these

behavioural parameters had limited effects on total resource intake and colony demographics,

both altered the origin of the collected resources, indicating a subtle modification of the

realised foraging domain. In agricultural landscapes, such shifts may result in a higher

exploitation of pesticide-treated crop patches. Empirical work has shown that such scenarios

can indeed elevate pollen-based pesticide risk in *B. terrestris* (Knapp et al. 2023).

Pesticides are known to disrupt within-nest social networks and social interactions in

bumblebees (Crall et al. 2018, 2019; Straub et al. 2023), but the extent to which ISI use

influences the visitation rates of different food patches and potential exposure to

agrochemicals is poorly known (unlike in honeybees; Hasenjager et al. 2024). In the context

of nectar foraging, ISI use emerged as an influential factor shaping spatial foraging patterns

by attracting conspecifics to crop patches. However, its effect was spatially limited but

additive with foraging bias on pollen visitation rate. I modelled ISI use as a density-dependent

response (effectively a 'copy-the-majority' rule) rather than a state-dependent 'copy-when-

uncertain' strategy (as suggested by Smolla et al. 2016), because formalising 'uncertainty' in

a spatially explicit context remains methodologically challenging. Furthermore, simple

attraction to conspecific density is often a proximate mechanism driving social aggregation,

even if the ultimate function aligns with reducing uncertainty (Leadbeater and Chittka 2007).

Nevertheless, this implementation corresponds to empirical findings indicating that social and

personal information interact synergistically, wherein individuals do not prioritize social cues

above direct experience (Leadbeater and Florent 2014; Smolla et al. 2016). I predicted that

bumblebees will alter their patch visitation pattern and exploit the available food sources less

efficiently in the presence of foraging bias, and this effect may be further amplified by the

presence of ISI use via attracting conspecifics to suboptimal resource patches. However, the

interplay between these two behavioural parameters were negligible, likely because the

simulated landscape was coarse-grained and included a proximal grassland patch that could

fully support all colonies throughout the year. Due to this landscape structure, foragers may

not need to copy others to locate additional resource patches, regardless of whether a foraging

bias is present. Consequently, the influence of ISI use could be substantially larger in more

heterogeneous, fine-grained environments, and may interact more strongly with foraging

biases, particularly if resources in the nesting habitat become insufficient during crop bloom.

Simulations indicated that the incorporated pesticide-induced foraging bias was a

major driver of pollen visitation rates to crop patches, regardless of patch distance and

outweighing the influence of ISI use. I propose that this bias did not directly cause bees to

shift from the grassland patch to OSR patches. Instead, it reduced their preference for more

rewarding resources and thereby weakened discrimination among available food sources

within the grassland patch, where multiple floral resources coexist. At the landscape scale,

this also diverted some foragers from the grassland patch to OSR patches. As a result, the

distribution of exploited patches more closely reflected the distribution of available patches

within the colony's foraging range. This also suggests that in landscapes containing multiple

semi-natural habitat patches, the effect of such a foraging bias may be less pronounced

compared to that observed in these simulations. Field studies have proved that some pesticide-

induced foraging biases can reduce colony-level resource collection in bumblebees (e.g., Gill

and Raine 2014), but their direct effects on visitation rates and their potential influence on

exposure risk have rarely been examined. Pesticide-induced foraging biases in pollinators,

such as the one modelled in this study, may emerge through multiple mechanisms. First,

insecticides can directly affect chemosensory orientation (Kárpáti et al. 2024), resulting in

suboptimal choices between rewarding and unrewarding patches. Alternatively,

agrochemicals can impair spatial and chemical learning (Samuelson et al. 2016; Muth et al.

2019), cognitive abilities (Andrione et al. 2016; Gray et al. 2024), floral preference (Gill and

Raine 2014; Phelps et al. 2018), or reduce foraging motivation (Siviter et al. 2021; Ohlinger et

al. 2022; Siviter and Muth 2022), all of which may indirectly lead to foraging decisions that

are similar to the modelled foraging bias. Even if such biases do not necessarily translate into

adverse changes in fitness, the resulting alteration of exposure risk may lead to the emergence

of multiple sublethal effects (Desneux et al. 2007; Siviter et al. 2021; Tosi et al. 2022). For

instance, exposure to acetamiprid was found to induce various colony- and individual-level

symptoms in bumblebees including significant reductions in drone count (Camp et al. 2020a;

Weitekamp et al. 2022), drone weight (Camp et al. 2020b; Weitekamp et al. 2022), number of

pupae (Weitekamp et al. 2022), number of interactions in the colony, immune response, and

worker body size (dependent also on food quality; Straub et al. 2023). Such life history trait-

related sublethal effects, in turn, can impair colony growth and reproductive success,

potentially leading to population declines and lowered pollination services in these species

(Stanley et al. 2015; Baron et al. 2017).

Patch choice decisions related to nectar and pollen visits differ in the Bumble-

BEEHAVE model incorporated in BEE-STEWARD. The calculation of handling time for

nectar foragers follows the model of Harder (1983) and energetic efficiency of exploiting a

patch (i.e., energy gained compared to the energy needed to exploit given layer) plays a

crucial role in determining the most profitable food source. However, pollen foraging is

modelled to maximise foraging rate. In addition, the priority of nectar and pollen foraging is

also different in the model, and individuals always search for nectar if the stimulus for this

task in the bee's colony is higher than the bee's associated threshold, regardless of the

stimulus-threshold relationship of the other tasks (including pollen foraging). This difference

may, at least partially, explain why the examined behavioural parameters had distinct impacts

on the nectar and pollen visits in the simulations.

In conclusion, this study shows that previously overlooked aspects of foraging

behaviour can alter the visitation rates of available resource patches in buff-tailed

bumblebees, which result corroborates with field observations that mass-flowering crops can

competitively dilute pollinator densities in agricultural landscapes (Holzschuh et al. 2011;

Riggi et al. 2024). In particular, simulation results imply that both ISI use and a pesticide-

induced foraging bias may lead to more frequent visitation of crop patches, without changing

overall resource consumption. Through positive feedback, such shifts in patch exploitation

may facilitate the accumulation of sublethal effects of pesticides in real-world scenarios.

Future efforts should extend the BEE-STEWARD model to more behavioural aspects of

pollinator foraging that could potentially influence visitation patterns to differently treated

resource patches. Incorporating pesticide concentrations in nectar and pollen and their

temporal dynamics in treated crops into the model, together with additional sublethal effects,

would also enable us to assess the conditions under which these effects on visitation rates are

likely to scale up, as well as their overall impact on colonies' survival and reproductive

success. In that way, we could have an even more powerful tool for predicting how pesticide

exposure influence bumblebee population dynamics and exploring how different intervention

options (e.g., installing hedgerows, application of crop rotation) might mitigate pesticide-

associated risks for pollinators in different agricultural landscapes.

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Author contribution statement

The author confirms sole responsibility for the following: conceptualisation, methodology,

data curation, formal analysis, investigation, validation, visualisation, and writing.

Data availability statement

Input files for the extended BEE-STEWARD model, R script for the analysis of model

outputs, and data supporting the results are archived and available at Figshare

(https://figshare.com/s/acee9613adfc7757267b).

Declaration of competing interests

The author has no conflict of interest to declare.

Ethics approval

Not applicable.

Declaration of generative AI and AI-assisted technologies in the writing process

During the preparation of this work, the author used the GPT-4o and GPT-5 models (OpenAI

Inc., USA) to assist writing and improve the readability and language of the manuscript. After

using this product, the author reviewed and edited the content as needed and take full

responsibility for the content of the published article.

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Table 1. Additional parameters in the extended BEE-STEWARD model.

Parameter	Description	Range (unit)
ISI use	When using ISI, foragers collect nectar or pollen from the closest patches where conspecifics are also present, provided such foraging sites are known.	absent or present
Foraging bias	With foraging bias, filling levels (if >0) are set to 1, making handling times independent of actual filling levels when bumblebees are exposed to the pesticide during OSR flowering period.	absent or present

738   Table 2. Estimated marginal means (EMMs) of the proxy measures for pollination capacity

739   and fitness in different combinations of the investigated behavioural parameters. EMMs and

740   contrasts are on the response scale and reported with their corresponding 95% confidence

741   intervals (in square brackets). All values were estimated from the fitted LMs.

Simulation measures	Baseline setting	With ISI use (presence vs. absence contrast)	With foraging bias (presence vs. absence contrast)	With both ISI use and foraging bias (contrast relative to the baseline setting)
Total number of adult workers	373 [336, 409]	420 [384, 457] (47.8 [-4, 99.5])	390 [353, 426] (17 [-34.7, 68.8])	384 [348, 421] (11.8 [-40, 63.5])
Number of colonies	37.3 [33.7, 40.9]	41.3 [37.7, 44.9] (4 [-1.1, 9.12])	40.5 [36.9, 44.1] (3.2 [-1.9, 8.3])	43.4 [39.8, 47.0] (6.1 [1, 11.2])
Total number of hibernating queens	452 [420, 485]	443 [410, 475] (-9.6 [-55.6, 36.4])	439 [407, 472] (-12.7 [-58.7, 33.3])	458 [426, 491] (6 [-40, 52])

 Figure legends

Figure 1. Schematic diagram of the simulated landscape. The simulated landscape consisted

of a grassland (green) and multiple oilseed rape (OSR) patches (yellow). All patches had a

200 m radius (r), resulting in a 1256.64 m^2 patch area. Due to the spatial distribution of OSR

patches around the grassland patch, four OSR patches were closer to the central patch (d_1: 15

m) than the diagonally located OSR patches (d_2: 186.9 m). The former patches were termed

'close OSR patches' (with dotted pattern), whereas the latter 'far OSR patches' (with diagonal

pattern). Distances on the diagram are not on scale.

Alt text for Figure 1. Schematic of a simulated landscape with a central grassland patch

surrounded by eight equally sized oilseed rape patches (all with 200 m radius). Four 'close'

patches are positioned 15 m from the grassland, while additional 'far' patches are located

diagonally at 186.9 m. Close and far patches are visually distinguished by different patterns.

Figure 2. The proportion of nectar visits on the close (panel A) and far (panel B) OSR

patches. Boxplots show the median and interquartile range; whiskers show values within 1.5-

fold of the interquartile range, dots indicate individual values. The x-axis denotes the absence

('No foraging bias') or presence ('With foraging bias') of a pesticide-induced foraging bias.

The filling colour of the boxplots indicate the absence ('No ISI use': teal) or presence of

inadvertent social information (ISI) use ('With ISI use': orange).

Alt text for Figure 2. Two-panel boxplot figure showing the proportion of nectar visits to

oilseed rape (OSR) patches. Panel A represents close OSR patches and Panel B far OSR

patches. The x-axis compares absence versus presence of pesticide-induced foraging bias, and

box colour indicates absence (teal) or presence (orange) of inadvertent social information

(ISI) use. Boxplots display medians, interquartile ranges, whiskers (1.5× IQR), and individual

data points.

Figure 3. The proportion of pollen visits on the close (panel A) and far (panel B) OSR

patches. Boxplots show the median and interquartile range (IQR). Whiskers denote the range

of 1.5 times the IQR, while dots indicate individual data points. The x-axis denotes the

absence ('No foraging bias') or presence ('With foraging bias') of a pesticide-induced

foraging bias. The filling colour of the boxplots indicate the absence ('No ISI use': purple) or

presence of ISI use ('With ISI use': magenta).

Alt text for Figure 3. Two-panel boxplot figure showing the proportion of pollen visits to

oilseed rape (OSR) patches. Panel A represents close OSR patches and Panel B far OSR

patches. The x-axis compares absence versus presence of pesticide-induced foraging bias, and

box colour indicates absence (purple) or presence (magenta) of ISI use. Boxplots display

medians, interquartile ranges, whiskers ($1.5\times$ IQR), and individual data points.