Title - Negative effects of agricultural intensification on energetic gain rates of a declining aerial insectivore

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1 Abstract

2 The historical rise of intensive agricultural practices is hypothesized to be related to declines of 3 grassland and aerial insectivorous birds. Drivers of declines may also influence the overall 4 abundance and spatial distribution of insects within agricultural landscapes. Subsequently, 5 average energetic gain rates of birds breeding within more agro-intensive landscapes may be 6 impacted. Lower energetic gain rates in agro-intensive landscapes may lead to reduced growth 7 rate, body condition or fledging success of nestlings but also to diminished body condition of 8 food provisioning adults. In this study, we assessed if energetic gain of nestlings and food 9 provisioning behavior of adults varied across a gradient of agricultural intensification in a 10 declining aerial insectivore, the Tree swallow (Tachycineta bicolor). We found that hourly gain 11 rate was lower in agro-intensive landscapes, and yet travel distances were longest within less 12 agro-intensive landscapes. Our results highlight that, in order to maximize long term average 13 gain rates, Tree swallows breeding within agro-intensive landscapes must forage with greater 14 intensity, perhaps at a cost to themselves, or else costs will transfer to growing broods. Our work 15 provides further evidence that agricultural intensification on the breeding grounds can contribute 16 to the declines of aerial insectivores in part through a trophic pathway.

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18 Key words: aerial insectivores; agricultural intensification; breeding costs; central-place
19 foraging; Diptera; marginal value theorem; PinPoint GPS; resource availability; *Tachycineta*20 *bicolor*; Tree swallows.

21 Introduction

22 Populations of both grassland and farmland birds have severely declined since the 1970s 23 (Stanton et al. 2018, Rosenberg et al. 2019, Bowler et al. 2019). These groups include many 24 aerial insectivores, a foraging guild presenting steeper population declines than most other avian 25 guilds (Nebel et al. 2010, Michel et al. 2016). Concurrent changes in landscape composition and 26 avian population indices have led to the hypothesis that these declines are driven, at least partly, 27 by the increasing use of intensive agricultural practices (Benton et al. 2003, Hallmann et al. 28 2014, Stanton et al. 2018, Spiller and Dettmers 2019). Indeed, intensive agricultural practices 29 increasing agricultural yield per unit area of cultivated land, such as row cropping and the use of 30 both synthetic fertilizers and pesticides, have risen dramatically in the past decades (Tscharntke 31 et al. 2005, DiBartolomeis et al. 2019, Malaj et al. 2020).

32 Although many aerial insectivores share a common foraging strategy and diet, the ecological 33 diversity of this guild invalidates many of the hypothesized drivers of farmland bird declines 34 (Stanton et al. 2018, Spiller and Dettmers 2019). For instance, increased mechanization, 35 including the frequency of forage harvesting, results in the significant nest destruction of ground 36 nesting farmland birds (Tews et al. 2013, Stanton et al. 2018). Yet, most aerial insectivores do 37 not nest on the ground and thus many hypothesized mechanisms underlying their declines 38 involve a trophic pathway. One of those mechanisms postulates that intensive agricultural 39 practices have reduced the overall abundance and changed the species composition or nutritive 40 value of their prey (Stanton et al. 2018, Twining et al. 2018, Spiller and Dettmers 2019). Many 41 factors of agricultural intensification can indeed affect the abundance, species assemblages, 42 distributions, and trophic interactions of insects (Raven and Wagner 2021, Wagner et al. 2021).

43 Therefore, intensive agricultural practices have the potential to affect the fitness of insectivorous44 birds through food limitation.

45 Previous studies have indeed observed that agricultural intensification, including habitats 46 impactful to insect communities and abundances, correlate with changes in dietary composition 47 (Nocera et al. 2012, English et al. 2017, Bellavance et al. 2018), reductions in annual fitness 48 proxies (Benton et al. 2002, Ghilain and Bélisle 2008, Garrett et al. 2021), and altered population 49 growth (Stanton et al. 2018) of several aerial insectivore species. Studies linking prey availability 50 to aerial insectivore breeding success have, however, often failed to observe such relationships 51 (Dawson and Bortolotti 2000, Dunn et al. 2011, Imlay et al. 2017, McClenaghan et al. 2019). A 52 lack of association between food availability and fitness proxies is potentially explained by 53 breeding individuals compensating for reduced food availability or quality by altering foraging 54 effort (Stephens et al. 2008). For altricial birds, food-provisioning is considered one of the most 55 costly behaviors as individuals must meet the increasing metabolic demands of their brood while 56 faring for themselves (Drent and Daan 1980, Wiersma 2005, Schifferli et al. 2014). Moreover, 57 aerial insectivores are mostly short-lived species and thus life history theory suggests 58 impediments to breeding will have significant impacts on lifetime reproductive success (Stearns 59 1992, Sæther and Bakke 2000, Berzins et al. 2020). Therefore, if compensation of reduced prey 60 availability is through increased foraging effort, then it potentially comes at a cost to an individual's own survival and future reproduction (Stearns 1992, Reznick et al. 2000, Harrison et 61 62 al. 2011).

63 The large-scale homogenization of landscapes resulting from agricultural intensification (Benton
64 et al. 2003) may thus alter foraging behavior and exacerbate the stress experienced by food-

65 provisioning adults (Hinsley 2000). Assuming food-provisioning birds are maximizing their 66 average long-term rate of energy gain (Charnov 1976, Stephens et al. 2008), and if food patches 67 are indeed of lower quality and/or more sparsely distributed within intensively cultivated 68 landscapes, then central-place foraging theory predicts lower gain and nest visitation rates within 69 these landscapes (Orians and Pearson 1979, Olsson et al. 2008). Therefore, parents may optimize 70 foraging strategies, potentially by increasing foraging efforts, to compensate for lowered gain 71 rates and to overcome the consequences of nesting within such agro-intensive landscapes 72 (Olsson et al. 2008). Empirical evidence for such elevated foraging workloads includes foraging 73 at greater distances, lower nest visitation rates, and reduced nest attendance in landscapes 74 dominated by agriculture (Bruun and Smith 2003, Poulin et al. 2010, Catry et al. 2014, Stanton et 75 al. 2016, Staggenborg et al. 2017, Evens et al. 2018). Furthermore, birds whose central places are 76 located within more intensely cultivated landscapes present several conditions potentially related 77 to increased foraging costs (Evens et al. 2018), including greater oxidative stress (Stanton et al. 78 2017, Evens et al. 2018) and lower immunocompetency (Belant et al. 2000, Pigeon et al. 2013). 79 It is therefore imperative to assess the influence of landscape context on the capacity of breeding 80 individuals to compensate for poor foraging conditions and the resultant workload.

A recent 11-year study observed that, upon controlling for local prey availability, several indices of breeding success in Tree swallow (*Tachycineta bicolor*) remained associated with an agricultural gradient in southern Québec, Canada (Garrett et al. 2021). Breeding attempts within landscapes dominated by intensively managed row crops resulted in reduced fledging success, longer nestling periods and shorter fledging wing lengths (Garrett et al. 2021). If agro-intensive habitats do indeed reduce the functional connectivity of landscapes by reducing the amount and distribution of suitable foraging patches, then central-place foragers in these landscapes potentially expend more time or energy traveling between food patches and exploiting them, with the additional cost of settling for patches of lower quality (Hinsley 2000, Bélisle 2005, Evens et al. 2018). Therefore, one hypothesized mechanism explaining the results of Garrett et al. (2021) is that swallows breeding in agro-intensive landscapes provide food at lower rates and thus nestlings experience lower energetic gain rates.

93 Presented here are the results of a study evaluating the strength of evidence for this hypothesis. 94 Our principal goals were to evaluate how a gradient of agricultural intensification influenced the 95 provisioning rate by adults, the energetic gain rates of nestlings, and travel distances of food 96 provisioning parents. Specifically, we evaluated a series of hypotheses grounded in central-place 97 foraging theory (Stephens et al. 2008, Olsson et al. 2008). We firstly assumed the duration of 98 each foraging bout is composed of the travel to, between, and within foraging patches, and that 99 the biomass of food collected while exploiting a patch increases, yet at a decreasing rate with 100 patch residency time (Olsson et al. 2008). We further assumed food bolus biomass is positively 101 related to its energy content, that traveling and foraging costs increase linearly with time, and 102 that nestling rearing females are attempting to maximize their average long-term net rate of 103 energy gains per unit time (Charnov 1976, Stephens et al. 2008). Therefore, females whose nests 104 are within landscapes composed of more dispersed or poorer foraging patches, as expected in 105 landscapes dominated by agro-intensive cultures, will experience longer travel times (Figure 1A 106 and C) or lower instantaneous gain rates while exploiting patches (Figure 1B and C) and thereby 107 show lower food delivery rates. Finally, shorter travel times can counterbalance slower gain rates 108 within food patches, resulting in different food provisioning rates and yet similar energetic gain 109 rates between contrasting landscapes (Figure 1D). In addition, we tested the prediction that 110 foraging distances would be furthest within more agro-intensive landscapes, indicating greater 6

travel time and costs in such landscapes. We also tested the expectation that these effects would be at least partly alleviated with increasing local prey availability (i.e. prey availability close to the nest), but exacerbated by increasing food-demand from the brood (e.g. larger and older broods; Geary et al. 2020).

115 Methods

116 Study sites and nest box visits

117 Data were collected between 2011 and 2016 within a nest box system placed across 40 farms 118 throughout a gradient of agricultural intensification in southern Québec, Canada (Figure 2, but 119 see Ghilain and Bélisle 2008 for details). The western portion of the study area is composed 120 primarily of large-scale row cropping, resulting in expanses denuded of forest cover and 121 dominated by corn and soybean monocultures, and to a lesser extent, other cereals (principally 122 wheat) (Bélanger and Grenier 2002, Ruiz and Domon 2009). This focus on monocultures has led 123 to the near ubiquitous use of fertilizers, herbicides, insecticides and fungicides, as well as 124 increased mechanization and removal of both surface waters and wetlands in this region (Jobin et 125 al. 1996). These agro-intensive landscapes transition to ones composed primarily of pastures and 126 forage crops (e.g. hay, alfalfa and clover) in the eastern parts of the study system and are 127 interspersed within large areas of forest cover and a systematic drop in pesticide detections 128 (Sigouin 2020).

Ten nest boxes were placed on the field margins of each farm (total N=400) and have been monitored every two days throughout each Tree swallow breeding season since 2006 (Porlier et al. 2009). The number of eggs and nestlings were counted during each nest box visit, and thus the age and phenology of each brood is known within two days of accuracy. We captured > 99%

of adult females (during incubation) and 80% of males (during the brooding period) of each
breeding attempt. Adults were banded with an aluminum US Geological Survey band containing
a unique identification code and a series of morpho-metrics and samples were collected (Ghilain
and Bélisle 2008, Lessard et al. 2014).

137 Food provisioning

138 Food provisioning was characterized via two rates defining hourly energetic gains of growing 139 broods: the total biomass of delivered food boluses per hour (bolus biomass; mg/hr) and the 140 number of food boluses delivered per hour (bolus delivery). These rates were estimated during 141 food-provisioning sessions conducted between 2011 and 2016. Sessions consisted of two 142 consecutive 30-min blocks in which collars were applied to each nestling within broods aged 6, 143 8, 10, and 12 days old (Bellavance et al. 2018). Collars were composed of a rubber band slid 144 through a 5-mm long feeding tube and adjusted like a bolo-tie to capture food boluses in the crop 145 yet loose enough to not restrict respiration. This method has often been used to document the 146 diet of Tree swallows, and has not been found to impact nestling survival (McCarty and Winkler 147 1999, Johnson and Lombardo 2000, Smits et al. 2005, Bellavance et al. 2018). At the end of each 148 block, boluses were carefully removed via forceps and the nest was checked for discarded 149 boluses. Boluses were then placed in sterile 50-ml conical tubes and immediately placed in a 150 portable freezer (< 12 hours) until being stored at -80°C. At the end of each session, collars were 151 removed and the starting time, the Julian date and the presence of precipitation was recorded. 152 The maximum temperature during sessions was also noted at the end of each breeding season via 153 iButtons (model DS1922L; Embedded Data Systems, Lawrenceburg, Kentucky, USA) attached 154 to the underside of a single nest box on each farm that recorded hourly temperatures.

In a tandem study occurring between 2013 and 2016, the food boluses collected during food provisioning sessions were analyzed for the presence of the principal active agents of many pesticides (Haroune et al. 2015). The methods of this study required recording the wet mass of food boluses (± 1 mg). We subsequently used this measure to estimate the total biomass of food boluses delivered during the one-hour food provisioning sessions.

160 Foraging distance

161 We used GPS locations of nestling rearing females (nestling age 4-12 days) to estimate their total 162 hourly distance traveled (DT) and mean hourly distance from their nest box (DNB). These 163 metrics were assumed to be indicative of the amount of travel undergone by each individual and 164 therefore to reflect the amount of energy expended while food provisioning. We used hourly 165 travel distances to have a comparable time scale to the hourly rates estimated by the food-166 provisioning sessions. Specifically, we fitted PinPoint-10 GPS tags (Lotek Wireless Inc, 167 Newmarket, Ontario, Canada) to 43 nestling rearing females across 14 farms in 2019 (Figure 2, 168 white points). Chosen farms represented landscapes dominated by either forage and pasture 169 fields or the principal row-crop monocultures of our study system (i.e. corn, soybean and other 170 cereals), with minimal tree cover. Tags weighed between 0.9 g and 1.2 g, including the harness 171 (Figure S1), and were fitted over the synsacrum using leg loops (Rappole and Tipton 1991). We 172 preprogrammed tags so GPS fixes ("SWIFT" fixes, nominal horizontal accuracy \pm 10 m) were 173 taken every 15 min between 5:00 and 20:00 local time (N = 60 locations) the day after 174 deployment, allowing for a habituation period. This schedule was the most intensive that we 175 could use to cover the majority of the daily food provisioning period given the battery-life 176 (McCarty 2002, Rose 2009). The day after data collection, individuals were caught, and their tag 177 removed. Once data were extracted, movements of each individual were derived by connecting GPS locations, in chronological order, and then calculating the distance along each segment of the resulting path using both the sp (Bivand et al. 2013) and rgeos (Bivand and Rundel 2019) packages in the R interface (R Core Team 2019). We estimated DT by summating the length of each segment within each one-hour sampling period. We then estimated DNB by grouping points within each one-hour block and then calculating the mean distance between each point and the female's nest box.

184 Landscape characterization

185 Landscape characterization started with delineating habitats, including agricultural parcels, 186 within 500 m of each nest box using orthophotos (scale 1:40 000) in QGIS (QGIS 2020) at the 187 end of each breeding season between 2011 to 2019. This spatial scale hosts close to 80% of 188 nestling rearing females' daily time budget. We then characterized each habitat parcel in situ, 189 defining agricultural habitats by the cultures grown. We then reclassified habitats into one of 190 four higher order categories represented by forest, corn and soybean, forage fields (including 191 pastures as well as hay, alfalfa, and clover fields), and cereals (e.g. wheat, oat, barley). Higher 192 order categories represent the hypothesized habitats Tree swallows use as indicators during nest-193 site selection, and the principal habitats composing landscapes throughout this system (Rendell 194 and Robertson 1990, Courtois 2020, Winkler et al. 2020).

We calculated the mean percent cover of each higher order category across all nest boxes on each farm for each breeding season, resulting in 240 separate landscape contexts (i.e., 40 farms x 6 years). To obtain an integrative measure of the percent cover of all higher order habitats, defined as the landscape context, we used a robust principal components analysis (PCA) for compositional data (Filzmoser et al. 2009) to assign "site scores" to each of the farms during each year. Site scores were defined by the values along the first two components of the resulting 10

201 compositional PCA as fitted using the robCompositions package in R (Templ et al. 2011). The 202 first component (Comp.1) explained 79.8% of the variance in landscape context and was 203 correlated positively with corn and soybean cultures and negatively with both forest cover and 204 forage fields (Figure 3A). The second component (Comp.2) explained 16.3% of the variance and 205 was positively correlated with forest cover, negatively with both forage and cereal covers. Site 206 scores were assigned to each food-provisioning session throughout the study.

207 The study of foraging distance occurred separately and on a subset of farms than those used to 208 estimate food provisioning (Figure 2). We thus repeated the landscape characterization used in 209 the food-provisioning study, yet for the subset of farms hosting the foraging distance study. The 210 resulting PCA was similar to that of the food provisioning sessions. The first component 211 explained 87.4% of the variance in landscape context and correlated positively with both corn 212 and soybean cultures, and negatively with both forest and forage cover (Figure 3B). The second 213 component explained 12.1% of the variance and correlated positively with forest cover as well as 214 with both corn and soybean cultures, and negatively with forage and cereal cultures.

215 For ease of interpretation, all statistical predictions and interpretations were made in reference to 216 five landscape context scenarios. These scenarios represent the predominant types of landscape 217 contexts experienced across the agricultural gradient and are defined by the site scores of five 218 farms (1-5) identified in Figure 3A and B. The scenarios for the food-provisioning sessions are: 219 (1) ones dominated by corn or soybean with 8.8% forest cover (corn/soybean with forest); (2) 220 ones denuded of forest cover and dominated by only corn or soybean (corn/soybean); (3) ones 221 denuded of forest cover and dominated by a mixture of corn, soybean and cereals (mixed row 222 crops); (4) ones with 9.0% forest cover and dominated by forage crops (forage); and (5) 223 landscapes with 69.0% forest cover and few cultivated parcels typically devoted to forage crops

(agro-forested). Although site score values differ between the PCA derived for the food provisioning and foraging distance studies, they are qualitatively similar and the above landscape
 context scenarios apply to both.

227 The distance of GPS locations from their respective nest boxes, greater than 0.5 km from nest 228 boxes (19.3% of locations), was positively related to the percent cover of water bodies (e.g. 229 rivers, wetlands and ponds) around a point (Figure S2), supporting the contention that these 230 features are important (Elgin et al. 2020, Berzins et al. 2020). These habitats are however rare 231 within our system and were considered as potential confounding factors of the observed foraging 232 responses. Moreover, we hypothesized that the influence of water features located far from a nest 233 may depend on their local availability (i.e. close to the nest). We therefore undertook a modeling 234 endeavor aimed at identifying the spatial extent and the type of water features most influential to 235 the rate of food provisioning (Appendix S1, Figure S3).

236 Due to several model convergence problems, we found numerical stability was greatest when 237 treating water features as the non-overlapping cover of wetlands and open bodies of water 238 (hitherto referred to as surface water). A local minimum of delta AICc was observed when 239 including the percent cover of surface water calculated within 1 km (local percent cover) and 240 10 km (regional percent cover) from the nest. These model terms were included in all models for 241 all foraging proxy. Surface water data were derived from vector layers acquired from the 242 Canadian National Hydro Network data sets (NHN 2020) and Ducks unlimited (Ducks 243 Unlimited, 2020). All calculations of percent cover were made using the sf (Pebesma 2018) R 244 package.

245 Local prey availability

246 Local prey availability was represented by the dry biomass of Diptera found within insect 247 samples from two separate insect traps placed on each of the 40 farms between 2006 and 2016. 248 We focused on dry biomass of Diptera (using the number of Diptera provides qualitatively 249 similar results, analyses not shown, see also Rioux Paquette et al. 2013, Garrett et al 2021) 250 because they represent a majority of the insects found in the diets of nestling Tree swallows 251 (McCarty and Winkler 1999, Johnson and Lombardo 2000, Mengelkoch et al. 2004, Twining et 252 al. 2018). Within our study system, 67% of the insects found in food boluses delivered to 253 nestlings are Diptera (Rioux Paquette et al. 2013, Bellavance et al. 2018). The focus of this work 254 being on the availability of food during the nestling period, we processed insect samples 255 collected between 1 June and 15 July, representing the nestling period of over 96% of first 256 breeding attempts. Diptera within each insect sample were removed, counted, and then placed in 257 an oven at 50°C for over 24 hours to ensure no further changes in mass. Sample dry mass was 258 then recorded (± 0.0001 g) without delay.

259 To obtain prey estimates more reflective of the farm vs. local scale of each trap, and avoid 260 punctuated events as insect swarms, we chose to use daily model predictions of Diptera biomass 261 from generalized additive models (GAMs) in which raw values were regressed against the Julian 262 date of sample collection for each farm and year. GAMs were fitted as a tensor product smoother 263 using the mgcv package in R (Wood 2017) with an identical degree of smoothness (k=10). 264 Model predictions are the accumulated biomass of Diptera over a roughly 48-hour period 265 occurring over a 3-day window (i.e. from the time the trap was reset on day one, the entirety of 266 day 2, and until the content of the trap was collected on day 3). As we wanted estimates of

267 Diptera biomass on the day of the food provisioning session, prey availability is represented by268 the predicted Diptera biomass on the day following the food provisioning session.

269 Statistical methods

270

Food provisioning

271 Bolus biomass was a continuous right-skewed (conditional) response containing zeros and was 272 consequently modeled with GLMMs using a tweedie distribution and a log link function (Lo and 273 Andrews 2015). It was noted that the likelihood to encounter food provisioning sessions with 274 zero delivered boluses varied with year, thus bolus delivery followed a (conditional) zero-275 inflated Poisson distribution, including a year-varying zero-inflation factor, and was modeled 276 with generalized linear mixed effect models (GLMMs) with a log link (Blasco Moreno et al. 277 2019). In order to account for differences in sampling effort (both 30-min blocks were completed 278 in 86% of sessions), we included whether sessions lasted half or an entire hour as an offset (i.e. 279 $\log(0.5)$ or $\log(1)$, respectively) in both sets of analyses. We further included year, farm, and 280 brood IDs as random factors in order to account for the data's hierarchical structure (i.e. broods 281 nested within farms nested within years).

We took an information theoretic and multimodel approach to evaluate links between landscape context and rates of food provisioning by comparing a set of 12 models representing biologically relevant alternative hypotheses (Table 1 and Table S1). Except for a Null model containing only random effects, we started with a Base model including potential confounding variables. This model contained two weather variables. First, the mean maximum temperature, as insects require a threshold temperature before being active (Taylor 1963, Grüebler et al. 2008), and the upper limit of foraging swallow behavior is potentially thermally constrained (Tapper et al. 2020).

289 Second, the presence of precipitation, as we expected precipitation to act as a hindrance to 290 foraging (Cox et al. 2019), and because food provisioning sessions did not last long enough to 291 estimate amount of precipitation. The base model also included the age of the female (second 292 year (SY) or after second year (ASY); Hussell 1983), time at the start of the provisioning session 293 as a second-order polynomial, the percent cover of surface water within 1 km and 10 km and 294 their interaction (Elgin et al. 2020), and a proxy of the brood's food demand (Geary et al. 2020). 295 This brood demand proxy was represented by the multiplication of the brood size with the brood 296 age. We did not expect foraging rates to respond linearly with this proxy, as the functional 297 response likely reaches a plateau whereby food provisioning parents can no longer increase their 298 efforts. We therefore treated this variable as a third-order polynomial. All model terms within 299 Base were included in all subsequent models of the candidate set. Our principal model of interest 300 was one in which sites scores and their interaction were added to assess landscape context effects 301 (Base + Land). Because potential negative effects of foraging within agro-intensive landscapes 302 may be alleviated with increasing prey availability closer to nests, we also considered models 303 with only local prey availability (Base + Food), one with both prey availability and site scores 304 (Base + Food + Land), and one also including the interaction between prey availability and 305 Comp.1 (Base + Food*Land). We further hypothesized provisioning rates may be influenced by 306 brood demand and the availability of prime foraging habitat and included four models with such 307 relationships. These included one in which the effect of land cover and local prey availability 308 varied with the demand proxy (Base + Food + Land * Demand and Base + Land + Food * 309 Demand) or the local (1 km) percent cover of surface water (Base + Land + Food * LSW and 310 Base + Food + Land * LSW).

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Foraging distance

The two response variables characterizing the foraging distances of females were the total hourly distance traveled (DT) and the mean hourly distance an individual was from its nest box (DNB). These response variables were modeled using linear mixed effects models (LMM). Both response variables were log-transformed to ensure normality of residuals. We included the individual and farm IDs as random effects to account for the hierarchical structure of the dataset (i.e. locations nested within individuals nested within farms).

318 We applied an information theoretic and multimodel approach to a set of six models to evaluate 319 hypotheses about foraging distances (Table 1 and Table S1). We started with a Null model only 320 including random effects and a Base model including potential confounding variables influential 321 to foraging distances (McCarty 2002, Dreelin et al. 2018, Elgin et al. 2020). Confounding factors 322 included the brood demand proxy as a third-order polynomial, as well as the percent cover of 323 surface water within 1 km and 10 km and their interaction. We further included proxies of local 324 prey availability, including the Julian date and the maximum observed temperature during each 325 one-hour sampling period as a second-order polynomial (Rioux Paquette et al. 2013, Bellavance 326 et al. 2018, Garrett et al. 2021). Though the principal model of interests was one including site 327 scores and their interaction (Base + Land), we compared another three models hypothesizing that 328 the effects of foraging within agro-intensive landscapes may either be alleviated or exacerbated 329 by confounding variables. We thus considered models also including an interaction term between 330 Comp.1 and each potential confounder, namely brood demand (Base + Land*Demand), local 331 (1 km) percent cover of surface water (Base + Land*LSW), and local prey availability (Base + 332 Land*Food).

In all analyses, the effects of key individual variables, including interactions, were estimated via multi-model inference whereby predictions were calculated by model-averaging with shrinkage and shown with their 95% unconditional confidence intervals (Burnham and Anderson 2002). All quantitative covariates were z-transformed and all analyses were performed in R version 3.6.2 (R Core Team 2019) using the glmmTMB (Brooks et al. 2017) and AICcmodavg (Mazerolle 2020) packages. Model validation was performed following Zuur et al. (2009) using the DHARMa package (Hartig 2020)

340 **Results**

341 Food provisioning

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Bolus biomass

343 We measured the total bolus biomass delivered during 741 food provisioning sessions from 420 344 broods, 92% of which consisted of both 30-min blocks. Overall mean bolus biomass was 345 284.5 mg per hour (SD = 344.9), ranging from 208.2 mg to 319.2 mg across years (Figure S4). 346 The Null and Base models were not supported by the data, contrary to models including 347 information on landscape context (Table S1). Compared to forage landscapes (#4 in Figure 4B), 348 bolus biomass was 54.0% lower within agro-forested landscapes (#5), 42.8% lower in mixed 349 row-crop landscapes (#3), and 32.8% lower within corn/soybean landscapes (#2). Unlike for 350 bolus delivery rate (see below), we did not find that the effect of corn/soybean cover was 351 dependent upon forest cover (#1). We further found bolus biomass increased almost linearly with 352 brood demand (Table 2 and Figure S5), and negatively correlated with the local percent cover of 353 surface water and was greater for ASY females (Table 2, Figure S5 and Figure S6). Lastly, bolus

354	biomass	increased	with	both	the	start	time	and	the	maximum	temperature	of	the	food
355	provision	ing sessior	n, until	l plate	auing	g at 14	:12 ar	nd 25	°C, r	espectively	(Figure S5).			

356

Bolus delivery

357 We conducted 1,208 food provisioning sessions from 662 different broods. Overall mean bolus 358 delivery was 2.00 boluses per hour (SD = 2.05), ranging from 1.69 to 2.25 across years (Figure 359 S4). While the Null and Base models were not supported by the data, several alternative models 360 received support (Table S1). We observed substantial landscape effects (Table 2, Figure 4A). 361 Compared to forage landscapes (#4 in Figure 4A), bolus delivery rate was 39.6% lower within 362 agro-forested landscapes (#5), 26.9% lower in mixed row-crop landscapes (#3), and 13.8% lower 363 within corn/soybean landscapes (#2). The effect of corn/soybean cover was however highly 364 dependent upon forest cover as the presence of remnant woodlots in landscapes highly 365 dominated by these crops (#1) was associated with a bolus delivery rate 7.8% lower than within 366 forage landscapes (#4). Although an increase of 18.4% in the mean bolus delivery rate was 367 predicted across the full range of local prey availability, the effect size of this variable lacked 368 precision (Table 2). We further found that bolus delivery rate was positively associated with 369 brood demand (third-order) and both the start time and the maximum temperature of the food 370 provisioning session, until plateauing at 14:23 and 25°C, respectively (Figure S7). It also 371 negatively correlated with the local percent cover of surface water (Figure S8).

372 Foraging distance

We successfully deployed and recovered 43 GPS tags. Of these tags, 40 successfully recorded all 60 attempted locations, resulting in 2,363 swift-fixes and 592 hourly estimates of foraging distances (Figure S9). The overall mean distance of each location from an individual's nest box

376	was 362 m (SD = 781) and 440 m (SD = 768) between each consecutive fix. Six females
377	recorded locations over 5 km from their nest box with a maximum distance of 9.557 km.

378

Hourly distance traveled

379 Overall mean DT was 1,841 m (SD=2,301, Figure S4). A single most predictive and 380 parsimonious model occurred when considering landscape context along with an interaction 381 between Comp.1 and local percent cover of surface water (Table 2 and Table S1). DT was 382 longest within agro-forested landscapes (#5 in Figure 4C) and was 60.1% further when compared 383 to forage landscapes (#4). Compared to forage landscapes, distances were 13.6% shorter within 384 mixed row-crop landscapes (#3), and yet 49.4% shorter within corn/soybean landscapes (#2). 385 The effect of corn/soybean cover was however highly dependent upon forest cover, as the 386 presence of remnant woodlots in landscapes highly dominated by these crops (#1) was associated with a DT 27.0% shorter than within forage landscapes (#4). DT increased with brood demand 387 until plateauing at mean demand values. DT was further associated in a quadratic fashion with 388 389 both the Julian date and maximum temperature, peaking on Julian date 185 (4 July) and at 390 26.1°C (Figure S10 and Figure S11).

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Hourly mean distance from nest box

Overall mean DNB was 368 m (SD=563, Figure S4). While no single model could be considered the most predictive or parsimonious, models taking landscape context into account were clearly better supported (Table 2 and Table S1). DNB was furthest within agro-forested landscapes (#5 in Figure 4D) and was 49.3% further when compared to forage landscapes (#4). Compared to forage landscapes, DNB was 3.3% further within mixed row-crop landscapes (#3), and yet 39.3% closer within corn/soybean landscapes (#2). The effect of corn/soybean cover was however highly dependent upon forest cover as the presence of remnant woodlots in landscapes

highly dominated by these crops (#1) was associated with DNB being 23.9% closer than within forage landscapes (#4). DNB increased significantly until plateauing at mean values of brood demand. Lastly, DNB was associated in a quadratic fashion with both the Julian date and maximum temperature, peaking on Julian date 184 (3 July) and at 26.7°C (Figure S12 and Figure S13).

404 **Discussion**

405 This study aimed to assess whether central-place foragers in agro-intensive landscapes expend 406 more time or energy reaching food patches and traveling between them with the additional cost 407 of settling for patches of lower quality, the combination of which resulting in lower food 408 provisioning rates (Figure 1). We found that female Tree swallows provided food to their 409 nestlings at lower rates within increasingly agro-intensive landscapes. The hourly rate of delivery 410 of both bolus biomass and bolus number were indeed lowest within more agro-intensive 411 landscapes and greatest within forage landscapes. Female Tree swallows traveled most and 412 furthest from their nest within forest-dominated landscapes, and unexpectedly less and closer to 413 their nest in agro-intensive compared to forage landscapes.

414 The relationship between the rate at which boluses were delivered and landscape context 415 resembled that at which prey biomass was provided, suggesting that the biomass of individual 416 boluses was similar throughout the agricultural intensification gradient studied (McCarty 2002). 417 This outcome was corroborated by a supplemental analysis (Appendix S2, Table S2 and Table 418 S3). Coupled with the result that females traveled furthest within predominantly forested 419 landscapes and to a lesser extent forage landscapes, the relatively low variance in the biomass of 420 individual boluses suggests that forage landscapes implied marginally greater travel times, yet 421 more rapid loading curves, which subsequently led to greater gain rates than agro-intensive

422 landscapes (Figure 5). Lower gain rates within more agro-intensive landscapes may need to be 423 compensated by parents elongating the entire foraging period of a given day, foraging more 424 intensely when conditions allow, forgoing recuperation and foraging for themselves (Olsson et 425 al. 2008, Schifferli et al. 2014). In all cases, increases in foraging effort may come at a cost to 426 themselves through reduced body condition and thus potentially impactful to their future 427 reproductive success (Stearns 1992, Reznick et al. 2000, Harrison et al. 2011, Lagrange 2015), as 428 increases in effort may result in foraging parents being in too poor of body condition for molting 429 prior to migration (Jenni-Eiermann and Jenni 1996, Rubolini et al. 2002). Furthermore, if parents 430 cannot compensate for lowered gain rates, then broods may incur the cost through reduced 431 growth rates or elevated mortality (Naef-Daenzer et al. 1999, Tremblay et al. 2004). In fact, 432 evidence suggest food provisioning Tree swallows as well as their broods already incur these 433 types of costs, as agro-intensive landscapes result in parents spending less time in the nest 434 (Lamoureux 2010, Stanton et al. 2016), expressing reduced body condition (Pigeon et al. 2013, 435 Stanton et al. 2017), and both reduced fledging success and nestling growth rates (Houle et al. 436 2020, Garrett et al. 2021). Some of these natal effects are of concern as they may be carried over 437 into later life stages (Harrison et al. 2011).

One fundamental characteristic of our gradient of agricultural intensification, and as an extension, most agricultural landscapes, is a loss of forest cover (Fischer and Lindenmayer 2007). Within our study area, landscapes dominated by a mixture of forest cover and agricultural fields resulted in the lowest gain rates and furthest foraging distances. The foraging habitat of Tree swallows, as well as of most swallow species, is frequently recognized as being open terrestrial habitats or over large water surfaces (Evans et al. 2007, Boynton et al. 2020, Elgin et al. 2020). Thus, forest cover may reduce landscape functional connectivity for these species, at

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445 least with respect to foraging (Bélisle 2005, Evens et al. 2018). In this way, as forest cover is 446 reduced, the cover of foraging habitats, close to central places and conducive to Tree swallows, 447 increases. In fact, second to forage landscapes, we found foraging rates were superior within 448 agro-intensive landscapes with remnant forest cover. Residual woodlots within these agro-449 intensive landscapes add to the tree lines separating fields to increase the length of "forest" 450 edges, where insect abundance tends to be greatest (Jokimäki et al. 1998, Evans et al. 2003, 451 Boetzl et al. 2020). The fledging success of Tree swallows is greater in agro-intensive landscapes 452 containing residual woodlots than ones denuded of forest cover (Garrett et al. 2021). It may 453 therefore be hypothesized that the elevated insect abundance associated with these residual 454 woodlots lead to greater gain rates and thus greater fledging success in this landscape context.

455 It should be noted that during the years covered by this study, areas intensively-managed for the 456 growing of corn and soybean harbored elevated levels of pesticides in their surface waters, 457 notably several herbicides (e.g. glyphosate, atrazine, S-metolachlor, imazethapyr) and 458 neonicotinoid insecticides [e.g. clothianidin, thiacloprid, thiametoxam (Giroux 2019, Montiel-459 León et al. 2019)]. For instance, a previous study found that, on average, exposure to active 460 pesticide agents within the food boluses was greatest within these landscapes (Sigouin 2020). 461 The consumption or prolonged exposure to such chemicals can influence the neurological, 462 endocrine and immunological systems of non-target animals, including vertebrates, via "skin" 463 contact, breathing or consumption of food or water (Mayne et al. 2005, Mineau et al. 2013, 464 Lopez-Antia et al. 2015, Pisa et al. 2015). Importantly, neurotoxic insecticides can induce 465 anorexia, alter orientation and influence motor functions required for flight in birds (Eng et al. 466 2017, 2019). Foraging within agro-intensive landscapes increases exposure of foragers to these

467 chemicals, likely reducing their motivation and ability to forage, and thus potentially resulting in468 decreased provisioning rates.

469 Finally, foraging is a metabolically expensive behavior, made worse when foraging bouts are 470 during the nestling periods (Bryant and Tatner 2008, Yap et al. 2017). Therefore, incoming food 471 need not only be frequent, but also energetically rich (Twining et al. 2018). The diversity and 472 abundance of insects found within food boluses suggests that the nutritive value of food varies 473 across the agricultural gradient of our study system (Bellavance et al. 2018). Most notably, 474 insects with aquatic life stages (principally Ephemeroptera) are more abundant within the food 475 boluses retrieved from forage landscapes than from those dominated by more agro-intensive 476 cultures. These insects are greater in highly unsaturated omega-3 fatty acids (HUFA; (Twining et 477 al. 2018). Nestlings with diets rich in HUFA can grow faster, gain a better body condition and 478 have greater immunocompetence, leading to elevated brood survival and fledging success 479 (Twining et al. 2018). Such effect can logically be extended to foraging adults. Subsequently, the 480 energetic losses of foraging may be more easily replenished within forage landscapes.

481 Global climate change results in increased global temperatures and predicts increases in both the 482 frequency and intensity of extreme weather events, such as prolonged periods of low or high 483 temperatures or precipitation (Rahmstorf and Coumou 2011, Wuebbles et al. 2014). Such 484 weather events may drastically alter the availability of insects as a food resource due to their thermally dependent nature (Grüebler et al. 2008, Shipley et al. 2020, Garrett et al. 2021). 485 486 Furthermore, recent evidence suggests Tree swallows may already be foraging at some upper 487 thermal threshold (Tapper et al. 2020). Our results, combined with other pieces of evidence, 488 suggests food provisioning adults may also be foraging at some upper energetic limit inasmuch, 489 as proxies of body condition suggest they are lower where gain rates are lower (Pigeon et al.

490 2013, Stanton et al. 2016, 2017). It is thus imperative that future work focuses not only on how 491 agricultural practices or global climate change independently impact declining avian species, but 492 also on how these factors may interact with one another. This importance is due to the ability of 493 growing broods or breeding parents to manage extreme weather is potentially lower within more 494 agro-intensive landscapes (Grue et al. 1997, Evans et al. 2003, Pérez et al. 2008, Eng et al. 2019, 495 Tapper et al. 2020). Finally, our results provide evidence that aerial insectivore declines may be 496 related to agricultural intensification through a trophic pathway.

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508 Tables and figures

509 Table 1: Mean ± SD of covariates used to model response variables. Covariates for the 510 hourly number of boluses and total bolus biomass delivered to nestlings were representative of 511 the hourly food provisioning sessions conducted between 2011 and 2016. Covariates of foraging 512 distances were calculated for the hourly interval's representative of the hourly GPS locations. 513 Dashes imply the covariate was not used in the modeling of the response variable. Covariate 514 groups identify which covariates were present within models found in Table 1.

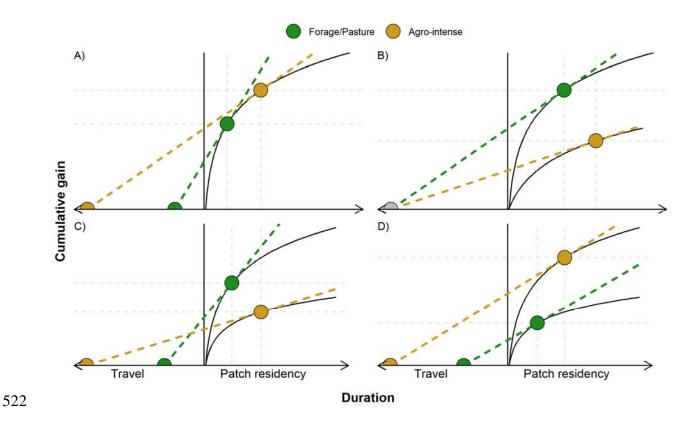
Covariate	Acronym	Unit	Covariate group	Biomass	Number	Foraging distance
Starting time	ST	Time	Base	262.31 ± 150.16	251.57 ± 153.55	-
Maximum temperature	TMAX	°C	Base/Food	$\begin{array}{c} 26.20 \pm \\ 4.08 \end{array}$	25.54 ± 3.94	$\begin{array}{c} 20.85 \pm \\ 5.08 \end{array}$
Brood demand proxy	DM	Brood size * Brood age	Base/Demand	$\begin{array}{c} 40.50 \pm \\ 13.89 \end{array}$	39.14 ± 12.8	$\begin{array}{c} 29.48 \pm \\ 10.87 \end{array}$
Local surface water (1 km)	LSW	%	Base/ LSW	$\begin{array}{c} 0.80 \pm \\ 1.43 \end{array}$	$\begin{array}{c} 0.87 \pm \\ 1.58 \end{array}$	1.38 ± 1.68
Regional surface water (10 km)	RSW	%	Base	1.58 ± 1.46	1.56 ± 1.42	2.62 ± 1.93
Prey availability (Diptera biomass)	PA	g	Food	$\begin{array}{c} 0.02 \pm \\ 0.01 \end{array}$	$\begin{array}{c} 0.02 \pm \\ 0.01 \end{array}$	-
Site score Comp.1	Comp.1	-	Land	0.34 ± 1.73	0.42 ± 1.72	0.07 ± 2.35
Site score Comp.2	Comp.2	-	Land	$\begin{array}{c} \textbf{-0.04} \pm \\ \textbf{0.88} \end{array}$	-0.01 ± 0.86	-0.09 ± 1.38
Julian day	JD	days	Base	-	-	170.83 ± 7.04

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516

517	Table 2:	Standardized coefficient estimates and their 95% confidence intervals included in
518	the top mode	of each response variable. See Table 1 for definitions of covariates and Table S1
519	for outcome	of model selection. Coefficients in bold indicate confidence intervals not
520	overlapping z	ero. Numbers next to acronym refer to the order of the polynomial term.

Covariate	Biomass	Delivery	Travel	Distance		
Comm 1	0.03	0.02	-0.11	-0.08		
Comp.1	(-0.01, 0.07)	(-0.04, 0.07)	(-0.16, -0.06)	(-0.15, -0.02)		
	0.07	0.09	-0.13	-0.10		
Comp.1:Comp.2	(0.02, 0.11)	(0.02, 0.16)	(-0.19, -0.06)	(-0.17, -0.03)		
Comm 2	-0.10	-0.15	0.07	0.03		
Comp.2	(-0.18, -0.01)	(-0.27, -0.02)	(-0.02, 0.16)	(-0.10, 0.15)		
Esmale and (ASV)	0.05	0.13				
Female age (ASY)	(-0.11, 0.21)	(-0.13, 0.40)	-	-		
$\mathbf{D}_{\mathbf{r}}$ ($\mathbf{V}_{\mathbf{r}}$)	-0.16	-0.18				
Rain (Yes)	(-0.40, 0.07)	(-0.52, 0.16)	-	-		
DM1	4.04	9.78	3.11	4.09		
DM1	(1.99, 6.10)	(7.09, 12.47)	(0.56, 5.66)	(1.17, 7.01)		
DM0	-5.21	-3.34	-1.86	-4.23		
DM2	(-7.29, -3.12)	(-6.11, -0.57)	(-4.35, 0.64)	(-6.99, -1.48)		
D1/2	2.81	1.07	-0.01	0.17		
DM3	(0.89, 4.72)	(-1.64, 3.78)	(-2.78, 2.76)	(-2.94, 3.28)		
0751	2.08	2.53		· · · /		
ST1	(-0.03, 4.18)	(-0.19, 5.26)	-	-		
0.070	-2.33	-3.02				
ST2	(-4.43, -0.22)	(-5.61, -0.43)	-	-		
	-0.23	0.13	8.84	8.28		
TMAX1	(-2.43, 1.97)	(-2.58, 2.83)	(6.43, 11.25)	(5.67, 10.90)		
	-3.87	-3.99	-5.14	-4.46		
TMAX2	(-6.04, -1.69)	(-6.75, -1.23)	(-7.48, -2.80)	(-7.02, -1.89)		
I GIV	-0.08	-0.12	0.16	0.05		
LSW	(-0.14, -0.01)	(-0.22, -0.02)	(-0.01, 0.33)	(-0.15, 0.25)		
	0.04	0.12	-0.22	-0.10		
LSW:RSW	(-0.05, 0.14)	(-0.03, 0.28)	(-0.41, -0.03)	(-0.35, 0.14)		
DOW	0.06	0.05	-0.27	-0.23		
RSW	(-0.01, 0.13)	(-0.06, 0.16)	(-0.41, -0.13)	(-0.43, -0.04)		
			11.17	11.85		
JD1	-	-	(8.04, 14.30)	(8.14, 15.56)		
IDA			-4.77	-5.20		
JD2	-	-	(-7.54, -1.99)	(-8.43, -1.97)		
			-0.07	(,)		
RSW:Comp.1	-	-	(-0.13, -0.01)	-		



523 Figure 1: Optimal energetic gain rates expected under different scenarios where food 524 provisioning individuals forage according to the marginal value theorem and experience different 525 travel times and gain curves that depend (or not) on landscape context. The x-axis is the total 526 duration of a foraging bout composed of both the travel time to foraging patches and the time 527 spent foraging within patches. Relationship between cumulative gain and patch residency time is 528 represented by a decelerating gain curve. The rate of energetic gains of a foraging bout is given 529 by the ratio between the cumulative gain and the bout's total duration. Optimal gain rate is thus 530 graphically represented by the slope of the dashed line connecting travel time to the tangent of 531 the gain curve. Assuming more agro-intensive landscapes result in more dispersed or poorer 532 foraging patches, landscape mediated differences in energetic gains potentially occur through 533 longer travel times (A and C) or lower instantaneous gain rates while exploiting patches (B and

534 C). Lastly, shorter travel times can counterbalance slower gain rates within food patches,

535 resulting in different food provisioning rates and yet similar energetic gain rates between

536 contrasting landscapes (D).

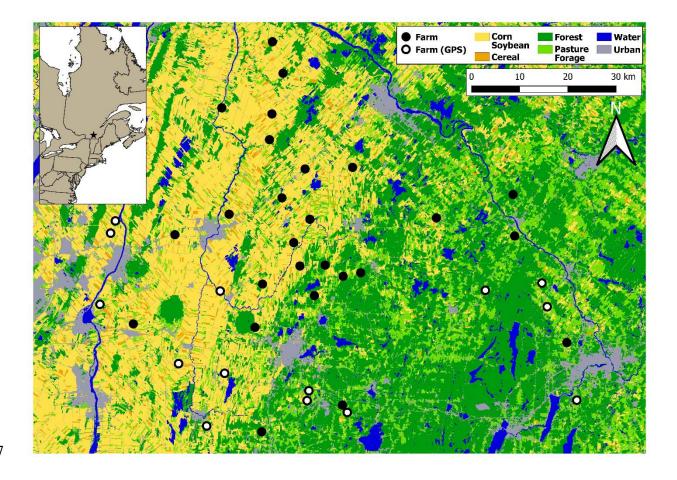




Figure 2: Map of study system. Each point represents the location of a farm, filled white points are farms where GPS were applied to 43 females. Underlying image represents the agricultural gradient derived from the Annual crop inventory (AAFC 2013). Light yellow represents the more agro-intensive areas while light green represents less agro-intensive areas.

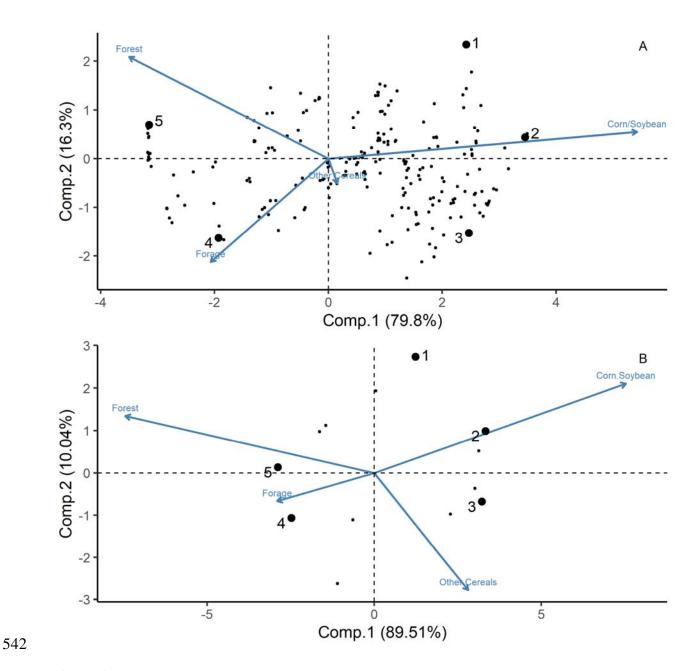
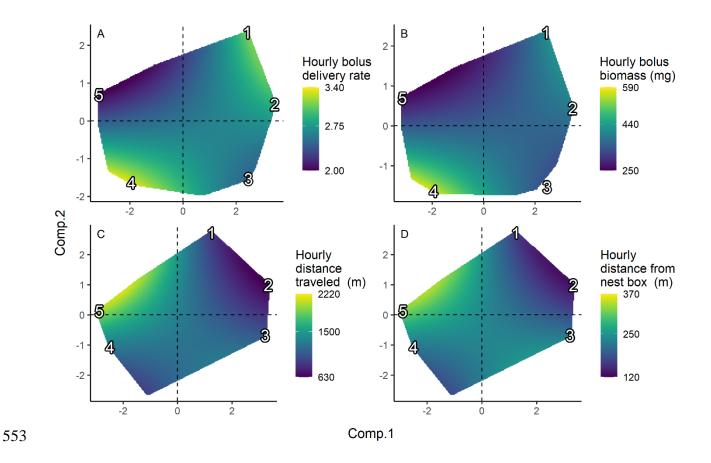


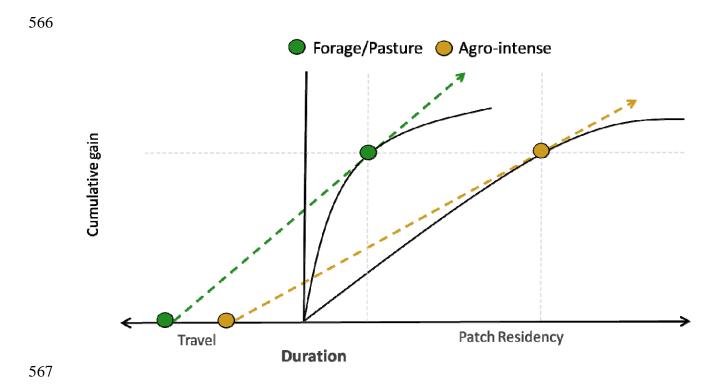
Figure 3: Results of the robust PCA for compositional data including the site scores of the five landscape context scenarios for the food provisioning sessions (A) and foraging distance studies (B). Points in the background represent the site scores assigned to each food provisioning session. The arrows lengths are eigenvectors representing the direction and magnitude of correlation between habitats. The five landscape scenarios are: ones dominated by corn or

548 soybean with 8.8% forest cover (Corn/soybean with forest, 1), ones denuded of forest cover and 549 dominated by only corn or soybean (Corn/Soybean,2), ones denuded of forest cover and 550 dominated by a mixture of corn, soybean and cereals (mixed row crop, 3), ones minimizing 551 forest cover (9.0%) and dominated by forage crops (Forage, 4), and agricultural landscapes with 552 69.0% forest cover (Agro-forested, 5)



554 Figure 4: Response surfaces of the principal response variables against the first two 555 components deriving the landscape context. Deeper blue represents lower predicted values while 556 vellow represents greater values. Predicted hourly bolus delivery rate (A), hourly bolus biomass 557 (B) delivered to nestlings. Predicted hourly distance traveled (C) and hourly distance from nest 558 box (D). Numbers in each panel are landscape context scenarios and represent the predominant 559 types of landscape contexts experienced across the agricultural gradient and correspond to the 560 points in numbers in Figure 3. Scenarios are: (1) ones dominated by corn or soybean with 8.8% 561 forest cover (corn/soybean with forest); (2) ones denuded of forest cover and dominated by only 562 corn or soybean (corn/soybean); (3) ones denuded of forest cover and dominated by a mixture of 563 corn, soybean and cereals (mixed row crops); (4) ones with 9.0% forest cover and dominated by

- 564 forage crops (forage); and (5) landscapes with 69.0% forest cover and few cultivated parcels
- 565 typically devoted to forage crops (agro-forested).



568 Figure 5: Reasoning underlying our hypothesis that food patches within forage landscapes 569 allow for a more rapid accumulation of food resources than those found in more agro-intensive 570 landscapes. Assuming that food provisioning Tree swallow females foraged according to the 571 marginal value theorem, and given we found (1) that the individual bolus biomass was similar 572 across the agricultural intensification gradient, (2) that travel distances were greater in forage 573 than in agro-intense landscapes, then only a lower gain curve in agro-intense landscapes is 574 compatible with our findings that hourly bolus biomass and hourly number of boluses delivered 575 were higher in forage than in agro-intense landscapes. Gain rate is graphically represented by the 576 slope of the dashed line connecting mean travel time to the tangent of the gain curve.

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578 Bibliography

- 579 Agriculture and Agri-Food Canada (AAFC). 2018. Annual Crop Inventory. 580 https://open.canada.ca/data/en/dataset/.
- 581 Bélanger, L., and M. Grenier. 2002. Agriculture intensification and forest fragmentation in the
 582 St. Lawrence valley, Québec, Canada. Landscape Ecology 17:495–507.
- Belant, J. L., L. A. Tyson, and T. W. Seamans. 2000. Wildlife Management: Cropping to
 Manage or Managing to Crop? Wildlife Society Bulletin 28:774–779.
- 585 Bélisle, M. 2005. Measuring landscape connectivity: the challenge of behavioral landscape
 586 ecology. Ecology 86:1988–1995.
- Bellavance, V., M. Bélisle, J. Savage, F. Pelletier, and D. Garant. 2018. Influence of agricultural
 intensification on prey availability and nestling diet in Tree Swallows (*Tachycineta bicolor*). Canadian Journal of Zoology 96:1053–1065.
- Benton, T. G., D. M. Bryant, L. Cole, and H. Q. P. Crick. 2002. Linking agricultural practice to
 insect and bird populations: a historical study over three decades. Journal of Applied
 Ecology 39:673–687.
- Benton, T. G., J. A. Vickery, and J. D. Wilson. 2003. Farmland biodiversity: is habitat
 heterogeneity the key? Trends in Ecology & Evolution 18:182–188.
- Berzins, L. L., R. D. Dawson, C. A. Morrissey, and R. G. Clark. 2020. The relative contribution
 of individual quality and changing climate as drivers of lifetime reproductive success in a
 short-lived avian species. Scientific Reports 10:19766.
- Bivand, R. S., E. J. Pebesma, and V. Gomez-Rubio. 2013. Applied spatial data analysis with R,
 Second edition. Springer.
- Bivand, R. S., and C. Rundel. 2019. Package 'rgeos' Interface to Geometry Engine Open
 Source ('GEOS').
- Blasco Moreno, A., M. Pérez Casany, P. Puig, M. Morante, and E. Castells. 2019. What does
 a zero mean? Understanding false, random and structural zeros in ecology. Methods in
 Ecology and Evolution 10:949–959.
- Boetzl, F. A., M. Schuele, J. Krauss, and I. Steffan Dewenter. 2020. Pest control potential of
 adjacent agri □environment schemes varies with crop type and is shaped by landscape
 context and within field position. Journal of Applied Ecology 57:1482–1493.
- Bowler, D. E., H. Heldbjerg, A. D. Fox, M. Jong, and K. Böhning□Gaese. 2019. Long□term
 declines of European insectivorous bird populations and potential causes. Conservation

- 610 Biology 33:1120–1130.
- Boynton, C. K., N. A. Mahony, and T. D. Williams. 2020. Barn Swallow (*Hirundo rustica*)
 fledglings use crop habitat more frequently in relation to its availability than pasture and
 other habitat types. The Condor 122.
- Brooks, M. E., K. Kristensen, K. J. van Benthem, A. Magnusson, C. W. Berg, A. Nielsen, H. J.
 Skaug, M. Mächler, and B. M. Bolker. 2017. glmmTMB balances speed and flexibility
 among packages for zero-inflated generalized linear mixed modeling. R Journal.
- Bruun, M., and H. G. Smith. 2003. Landscape composition affects habitat use and foraging flight
 distances in breeding European starlings. Biological Conservation 114:179–187.
- Bryant, D. M., and P. Tatner. 2008. Intraspecies variation in avian energy expenditure: correlates
 and constraints. Ibis 133:236–245.
- Burnham, K. P., and D. R. Anderson. 2002. Model Selection and Multimodel Inference. second.
 Springer.
- 623 2020. National Network NHN Canada, N. R. Hydro _ -GeoBase Series. 624 https://www.nrcan.gc.ca/science-and-data/science-and-research/earth-625 sciences/geography/topographic-information/geobase-surface-water-program-626 geeau/national-hydrographic-network/21361.
- 627 Catry, I., A. M. A. Franco, and F. Moreira. 2014. Easy but ephemeral food: exploring the trade628 offs of agricultural practices in the foraging decisions of Lesser Kestrels on farmland. Bird
 629 Study 61:447–456.
- 630 Charnov, E. L. 1976. Optimal foraging, the marginal value theorem. Theoretical Population
 631 Biology 9:129–136.
- 632 Courtois, È. 2020. Sélection De Nichoir Chez L'hirondelle Bicolore (*Tachycineta Bicolor*) En
 633 Milieux Agricoles : Une Trappe Écologique?
- Cox, A. R., R. J. Robertson, Á. Z. Lendvai, K. Everitt, and F. Bonier. 2019. Rainy springs linked
 to poor nestling growth in a declining avian aerial insectivore (*Tachycineta bicolor*).
 Proceedings of the Royal Society B: Biological Sciences 286:20190018.
- Dawson, R. D., and G. R. Bortolotti. 2000. Reproductive Success of American Kestrels: The
 Role of Prey Abundance and Weather. The Condor 102:814–822.
- DiBartolomeis, M., S. Kegley, P. Mineau, R. Radford, and K. Klein. 2019. An assessment of
 acute insecticide toxicity loading (AITL) of chemical pesticides used on agricultural land in
 the United States. PLOS ONE 14:e0220029.

- Dreelin, R. A., J. R. Shipley, and D. W. Winkler. 2018. Flight Behavior of Individual Aerial
 Insectivores Revealed by Novel Altitudinal Dataloggers. Frontiers in Ecology and
 Evolution 6:1–7.
- Drent, R. H., and S. Daan. 1980. The Prudent Parent: Energetic Adjustments in Avian Breeding.
 Ardea 68:225–252.
- 647 Ducks Unlimited. 2020. Ducks unlimited GIS mapping applications.
 648 https://www.ducks.ca/initiatives/gis-mapping-applications/.
- Dunn, P. O., D. W. Winkler, L. a. Whittingham, S. J. Hannon, and R. J. Robertson. 2011. A test
 of the mismatch hypothesis: How is timing of reproduction related to food abundance in an
 aerial insectivore? Ecology 92:450–461.
- Elgin, A. S., R. G. Clark, and C. A. Morrissey. 2020. Tree Swallow selection for wetlands in
 agricultural landscapes predicted by central-place foraging theory. The Condor 122:1–12.
- Eng, M. L., B. J. M. Stutchbury, and C. A. Morrissey. 2017. Imidacloprid and chlorpyrifos
 insecticides impair migratory ability in a seed-eating songbird. Scientific Reports 7:15176.
- Eng, M. L., B. J. M. Stutchbury, and C. A. Morrissey. 2019. A neonicotinoid insecticide reduces
 fueling and delays migration in songbirds. Science 365:1177–1180.
- English, P. A., J. J. Nocera, B. A. Pond, and D. J. Green. 2017. Habitat and food supply across
 multiple spatial scales influence the distribution and abundance of a nocturnal aerial
 insectivore. Landscape Ecology 32:343–359.
- Evans, K. L., R. B. Bradbury, and J. D. Wilson. 2003. Selection of hedgerows by Swallows
 Hirundo rustica foraging on farmland: the influence of local habitat and weather. Bird
 Study 50:8–14.
- Evans, K. L., J. D. Wilson, and R. B. Bradbury. 2007. Effects of crop type and aerial invertebrate
 abundance on foraging barn swallows *Hirundo rustica*. Agriculture, Ecosystems &
 Environment 122:267–273.
- Evens, R., N. Beenaerts, T. Neyens, N. Witters, K. Smeets, and T. Artois. 2018. Proximity of
 breeding and foraging areas affects foraging effort of a crepuscular, insectivorous bird.
 Scientific Reports 8:3008.
- Filzmoser, P., K. Hron, and C. Reimann. 2009. Principal component analysis for compositional
 data with outliers. Environmetrics 20:621–632.
- Fischer, J., and D. B. Lindenmayer. 2007. Landscape modification and habitat fragmentation: a
 synthesis. Global Ecology and Biogeography 16:265–280.

- 674 Garrett, D. R., F. Pelletier, D. Garant, and M. Bélisle. 2021. Combined influence of food 675 availability and agricultural intensification on a declining aerial insectivore. bioRxiv.
- Geary, B., P. L. Leberg, K. M. Purcell, S. T. Walter, and J. Karubian. 2020. Breeding Brown
 Pelicans Improve Foraging Performance as Energetic Needs Rise. Scientific Reports
 10:1686.
- Ghilain, A., and M. Bélisle. 2008. Breeding success of tree swallows along a gradient of
 agricultural intensification. Ecological Applications 18:1140–1154.
- 681 Giroux, I. 2019. Présence de pesticides dans l'eau au Québec□: Portrait et tendances dans les
 682 zones de maïs et de soya 2015 à 2017, Québec, ministère de l'Environnement et de la
 683 Lutte contre les changements climatiques, Direction générale du suivi de l'état de l'environ.
- 684 Grue, C. E., P. L. Gibert, and M. E. Seeley. 1997. Neurophysiological and Behavioral Changes
 685 in Non-Target Wildlife Exposed to Organophosphate and Carbamate Pesticides:
 686 Thermoregulation, Food Consumption, and Reproduction. American Zoologist 37:369–388.
- 687 Grüebler, M. U., M. Morand, and B. Naef-Daenzer. 2008. A predictive model of the density of
 688 airborne insects in agricultural environments. Agriculture, Ecosystems & Environment
 689 123:75–80.
- Hallmann, C. A., R. P. B. Foppen, C. A. M. van Turnhout, H. de Kroon, and E. Jongejans. 2014.
 Declines in insectivorous birds are associated with high neonicotinoid concentrations.
 Nature 511:341–343.
- Haroune, L., R. Cassoulet, M.-P. Lafontaine, M. Bélisle, D. Garant, F. Pelletier, H. Cabana, and
 J.-P. Bellenger. 2015. Liquid chromatography-tandem mass spectrometry determination for
 multiclass pesticides from insect samples by microwave-assisted solvent extraction
 followed by a salt-out effect and micro-dispersion purification. Analytica Chimica Acta
 891:160–170.
- Harrison, X. A., J. D. Blount, R. Inger, D. R. Norris, and S. Bearhop. 2011. Carry-over effects as
 drivers of fitness differences in animals. Journal of Animal Ecology 80:4–18.
- Hartig, F. 2020. DHARMa: Residual Diagnostics for Hierarchical (Multi-Level / Mixed)
 Regression Models. R package version 0.2.0.
- Hinsley, S. A. 2000. The costs of multiple patch use by birds. Landscape Ecology 15:765–775.
- Houle, C., F. Pelletier, M. Bélisle, and D. Garant. 2020. Impacts of environmental heterogeneity
 on natural selection in a wild bird population. Evolution 74:1142–1154.
- Hussell, D. J. T. 1983. Age and Plumage Color in Female Tree Swallows. Journal of Field
 Ornithology 54:312–318.

- Imlay, T. L., H. A. R. Mann, and M. L. Leonard. 2017. No effect of insect abundance on nestling
 survival or mass for three aerial insectivores. Avian Conservation and Ecology 12:art19.
- Jenni-Eiermann, S., and L. Jenni. 1996. Metabolic Differences Between the Postbreeding,
 Moulting and Migratory Periods in Feeding and Fasting Passerine Birds. Functional
 Ecology 10:62.
- Jobin, B., J.-L. DesGranges, and C. Boutin. 1996. Population trends in selected species of
 farmland birds in relation to recent developments in agriculture in the St. Lawrence Valley.
 Agriculture, Ecosystems & Environment 57:103–116.
- Johnson, M. E., and M. P. Lombardo. 2000. Nestling Tree Swallow (*Tachycineta bicolor*) Diets
 in an Upland Old Field in Western Michigan. The American Midland Naturalist 144:216–
 219.
- Jokimäki, J., E. Huhta, J. Itämies, and P. Rahko. 1998. Distribution of arthropods in relation to
 forest patch size, edge, and stand characteristics. Canadian Journal of Forest Research
 28:1068–1072.
- Jones, G. 1987. Parent-offspring resource allocation in swallows during nestling rearing: an
 experimental study. Ardea 75:145–168.
- Jones, G. 1988. Concurrent Demands of Parent and Offspring Swallows *Hirundo rustica* in a
 Variable Feeding Environment. Ornis Scandinavica 19:145.
- Lagrange, P. 2015. Déterminants De La Survie Et De La Dispersion De Reproduction Par Une
 Approche Capture-Marquage-Recapture Chez L'Hirondelle Bicolore Au Québec.
- Lamoureux, S. 2010. Impact de l'intensification agricole sur l'effort parental, la croissance et la
 survie des oisillons chez l'hirondelle bicolore (*Tachycineta bicolor*). Université de
 Sherbrooke.
- Lessard, A., A. Bourret, M. Bélisle, F. Pelletier, and D. Garant. 2014. Individual and
 environmental determinants of reproductive success in male tree swallow (*Tachycineta bicolor*). Behavioral Ecology and Sociobiology 68:733–742.
- Lo, S., and S. Andrews. 2015. To transform or not to transform: using generalized linear mixed
 models to analyse reaction time data. Frontiers in Psychology 6:1–16.
- Lopez-Antia, A., M. E. Ortiz-Santaliestra, F. Mougeot, and R. Mateo. 2015. Imidacloprid-treated
 seed ingestion has lethal effect on adult partridges and reduces both breeding investment
 and offspring immunity. Environmental Research 136:97–107.
- Malaj, E., L. Freistadt, and C. A. Morrissey. 2020. Spatio-Temporal Patterns of Crops and
 Agrochemicals in Canada Over 35 Years. Frontiers in Environmental Science 8:1–12.

- Mayne, G. J., C. A. Bishop, P. A. Martin, H. J. Boermans, and B. Hunter. 2005. Thyroid
 Function in Nestling Tree Swallows and Eastern Bluebirds Exposed to Non-Persistent
 Pesticides and p, p-DDE in Apple Orchards of Southern Ontario, Canada. Ecotoxicology
 14:381–396.
- Mazerolle, M. J. 2020. AICcmodavg: model selection and multimodel inference based on(Q)AIC(c).
- McCarty, J. P. 2002. The number of visits to the nest by parents is an accurate measure of food
 delivered to nestlings in Tree Swallows. Journal of Field Ornithology 73:9–14.
- McCarty, J. P., and D. W. Winkler. 1999. Foraging Ecology and Diet Selectivity of Tree
 Swallows Feeding Nestlings. The Condor 101:246–254.
- McClenaghan, B., K. C. R. Kerr, and E. Nol. 2019. Does prey availability affect the reproductive
 performance of Barn Swallows (*Hirundo rustica*) breeding in Ontario, Canada? Canadian
 Journal of Zoology 97:979–987.
- Mengelkoch, J. M., G. J. Niemi, and R. R. Regal. 2004. Diet of the nestling tree swallow. The
 Condor 106:423–429.
- Michel, N. L., A. C. Smith, R. G. Clark, C. A. Morrissey, and K. A. Hobson. 2016. Differences
 in spatial synchrony and interspecific concordance inform guild-level population trends for
 aerial insectivorous birds. Ecography 39:774–786.
- Mineau, P., C. Palmer, J. Melorose, R. Perroy, S. Careas, D. R. Tobergte, and S. Curtis. 2013.
 Summary for Policymakers. Pages 1–30 *in* Intergovernmental Panel on Climate Change,
 editor. Climate Change 2013 The Physical Science Basis. Cambridge University Press.
- Montiel-León, J. M., G. Munoz, S. Vo Duy, D. T. Do, M. Vaudreuil, K. Goeury, F. Guillemette,
 M. Amyot, and S. Sauvé. 2019. Widespread occurrence and spatial distribution of
 glyphosate, atrazine, and neonicotinoids pesticides in the St. Lawrence and tributary rivers.
 Environmental Pollution 250:29–39.
- Naef-Daenzer, B., L. F. Keller, B. N. Daenzer, and L. F. Keller. 1999. The foraging performance
 of great and blue tits (*Parus major* and *P. caeruleus*) in relation to caterpillar development,
 and its consequences for nestling growth and fledging weight. Journal of Animal Ecology
 68:708–718.
- Nebel, S., A. Mills, J. D. McCracken, and P. D. Taylor. 2010. Declines of Aerial Insectivores in
 North America Follow a Geographic Gradient. Avian Conservation and Ecology 5:art1.
- Nocera, J. J., J. M. Blais, D. V. Beresford, L. K. Finity, C. Grooms, L. E. Kimpe, K. Kyser, N.
 Michelutti, M. W. Reudink, and J. P. Smol. 2012. Historical pesticide applications

- coincided with an altered diet of aerially foraging insectivorous chimney swifts.
 Proceedings of the Royal Society B: Biological Sciences 279:3114–3120.
- Olsson, O., J. S. Brown, and K. L. Helf. 2008. A guide to central place effects in foraging.
 Theoretical Population Biology 74:22–33.
- Orians, G. H., and E. Pearson. 1979. On the theory of central place foraging. Analysis of
 Ecological Systems:154–1177.
- Pebesma, E. J. 2018. Simple Features for R: Standardized Support for Spatial Vector Data. TheR Journal 10:439.
- Pérez, J. H., D. R. Ardia, E. K. Chad, and E. D. Clotfelter. 2008. Experimental heating reveals
 nest temperature affects nestling condition in tree swallows (Tachycineta bicolor). Biology
 Letters 4:468–471.
- Pigeon, G., R. Baeta, M. Bélisle, D. Garant, and F. Pelletier. 2013. Effects of agricultural
 intensification and temperature on immune response to phytohemagglutinin in Tree
 Swallows (*Tachycineta bicolor*). Canadian Journal of Zoology 91:56–63.
- Pisa, L. W., V. Amaral-Rogers, L. P. Belzunces, J. M. Bonmatin, C. A. Downs, D. Goulson, D.
 P. Kreutzweiser, C. Krupke, M. Liess, M. McField, C. A. Morrissey, D. A. Noome, J.
 Settele, N. Simon-Delso, J. D. Stark, J. P. Van der Sluijs, H. Van Dyck, and M. Wiemers.
 2015. Effects of neonicotinoids and fipronil on non-target invertebrates. Environmental
 Science and Pollution Research 22:68–102.
- Porlier, M., M. Bélisle, and D. Garant. 2009. Non-random distribution of individual genetic
 diversity along an environmental gradient. Philosophical Transactions of the Royal Society
 B: Biological Sciences 364:1543–1554.
- Poulin, B., G. Lefebvre, and L. Paz. 2010. Red flag for green spray: adverse trophic effects of Bti
 on breeding birds. Journal of Applied Ecology 47:884–889.
- 797 QGIS. 2020. QGIS Geographic Information System. QGIS Association. http://www.qgis.org.
- 798 R Core Team. 2019. R: A language and environment for statistical computing.
- Rahmstorf, S., and D. Coumou. 2011. Increase of extreme events in a warming world.
 Proceedings of the National Academy of Sciences 108:17905–17909.
- Rappole, J. H., and A. R. Tipton. 1991. New harness design for attachment of radio transmitters
 to small passerines. Journal of field ornithology 62:335–337.
- Raven, P. H., and D. L. Wagner. 2021. Agricultural intensification and climate change are
 rapidly decreasing insect biodiversity. Proceedings of the National Academy of Sciences

- 805 118:e2002548117.
- Rendell, W. B., and R. J. Robertson. 1990. Influence of Forest Edge on Nest-Site Selection by
 Tree Swallows. Wilson Journal of Ornithology 102:634–644.
- Reznick, D., L. Nunney, and A. Tessier. 2000. Big houses, big cars, superfleas and the costs of
 reproduction. Trends in Ecology & Evolution 15:421–425.
- Rioux Paquette, S., D. Garant, F. Pelletier, and M. Bélisle. 2013. Seasonal patterns in Tree
 Swallow prey (Diptera) abundance are affected by agricultural intensification. Ecological
 Applications 23:122–133.
- Rose, A. P. 2009. Temporal and Individual Variation in Offspring Provisioning by Tree
 Swallows: A New Method of Automated Nest Attendance Monitoring. PLoS ONE 4:e4111.
- Rosenberg, K. V., A. M. Dokter, P. J. Blancher, J. R. Sauer, A. C. Smith, P. A. Smith, J. C.
 Stanton, A. Panjabi, L. Helft, M. Parr, and P. P. Marra. 2019. Decline of the North
 American avifauna. Science 366:120–124.
- Rubolini, D., A. Massi, and F. Spina. 2002. Replacement of body feathers is associated with low
 pre-migratory energy stores in a long-distance migratory bird, the barn swallow (*Hirundo rustica*). Journal of Zoology 258:S0952836902001590.
- Ruiz, J., and G. Domon. 2009. Analysis of landscape pattern change trajectories within areas of
 intensive agricultural use: case study in a watershed of southern Québec, Canada.
 Landscape Ecology 24:419–432.
- 824 Sæther, B.-E., and O. Bakke. 2000. Avian Life History Variation and Contribution of
 825 Demographic Traits to the Population Growth Rate. Ecology 81:642.
- Schifferli, L., M. U. Grüebler, H. A. J. Meijer, G. H. Visser, and B. Naef-Daenzer. 2014. Barn
 Swallow *Hirundo rustica* parents work harder when foraging conditions are good. Ibis
 156:777–787.
- Shipley, J. R., C. W. Twining, C. C. Taff, M. N. Vitousek, A. Flack, and D. W. Winkler. 2020.
 Birds advancing lay dates with warming springs face greater risk of chick mortality.
 Proceedings of the National Academy of Sciences 117:25590–25594.
- 832 Sigouin, A. 2020. Effets Combinés De L'exposition Aux Pesticides Et À Des Ectoparasites
 833 Hématophages Sur La Physiologie Des Oisillons De L'hirondelle Bicolore. Université de
 834 Sherbrooke.
- Smits, J. E. G., G. R. Bortolotti, M. Sebastian, and J. J. H. Ciborowski. 2005. Spatial, temporal,
 and dietary determinants of organic contaminants in nestling tree swallows in point Pelee
 national park, Ontario, Canada. Environmental Toxicology and Chemistry 24:3159.

- Spiller, K. J., and R. Dettmers. 2019. Evidence for multiple drivers of aerial insectivore declines
 in North America. The Condor 121:1–13.
- Staggenborg, J., H. M. Schaefer, C. Stange, B. Naef-Daenzer, and M. U. Grüebler. 2017. Time
 and travelling costs during chick-rearing in relation to habitat quality in Little Owls Athene
 noctua. Ibis 159:519–531.
- Stanton, R., R. G. Clark, and C. A. Morrissey. 2017. Intensive agriculture and insect prey
 availability influence oxidative status and return rates of an aerial insectivore. Ecosphere
 845 8:e01746.
- Stanton, R., C. A. Morrissey, and R. G. Clark. 2016. Tree Swallow (Tachycineta bicolor)
 foraging responses to agricultural land use and abundance of insect prey. Canadian Journal
 of Zoology 94:637–642.
- Stanton, R., C. A. Morrissey, and R. G. Clark. 2018. Analysis of trends and agricultural drivers
 of farmland bird declines in North America: A review. Agriculture, Ecosystems &
 Environment 254:244–254.
- 852 Stearns, S. C. 1992. The evolution of life histories. Oxford University Press.
- Stephens, D. W., J. S. Brown, and R. C. Ydenberg. 2008. Foraging: behavior and ecology.
 University of Chicago Press.
- Tapper, S., J. J. Nocera, and G. Burness. 2020. Experimental evidence that hyperthermia limits
 offspring provisioning in a temperate- breeding bird. Royal Societty Open Science 7.
- Taylor, L. R. 1963. Analysis of the Effect of Temperature on Insects in Flight. The Journal ofAnimal Ecology 32:99.
- Templ, M., K. Hron, and P. Filzmoser. 2011. robCompositions: An R-package for Robust
 Statistical Analysis of Compositional Data. Pages 341–355 Compositional Data Analysis.
 John Wiley & Sons, Ltd.
- Tews, J., D. G. Bert, and P. Mineau. 2013. Estimated Mortality of Selected Migratory Bird
 Species from Mowing and Other Mechanical Operations in Canadian Agriculture. Avian
 Conservation and Ecology 8:art8.
- Tremblay, I., D. W. Thomas, J. Blondel, P. Perret, and M. M. Lambrechts. 2004. The effect of
 habitat quality on foraging patterns, provisioning rate and nestling growth in Corsican Blue
 Tits Parus caeruleus. Ibis 147:17–24.
- Tscharntke, T., A. M. Klein, A. Kruess, I. Steffan Dewenter, and C. Thies. 2005. Landscape
 perspectives on agricultural intensification and biodiversity ecosystem service
 management. Ecology Letters 8:857–874.

- Twining, C. W., J. R. Shipley, and D. W. Winkler. 2018. Aquatic insects rich in omega-3 fatty
 acids drive breeding success in a widespread bird. Ecology Letters 21:1812–1820.
- Wagner, D. L., E. M. Grames, M. L. Forister, M. R. Berenbaum, and D. Stopak. 2021. Insect
 decline in the Anthropocene: Death by a thousand cuts. Proceedings of the National
 Academy of Sciences 118:e2023989118.
- Wiersma, P. 2005. Metabolic adjustments to increasing foraging costs of starlings in a closed
 economy. Journal of Experimental Biology 208:4099–4108.
- Winkler, D. W., K. K. Hallinger, D. R. Ardia, R. J. Robertson, B. J. M. Stutchbury, and R. R.
 Cohen. 2020. Tree Swallow (*Tachycineta bicolor*). Page *in* A. F. Poole, editor. Birds of the
 World. Cornell Lab of Ornithology.
- 881 Wood, S. N. 2017. Generalized additive models: an introduction with R. CRC press.
- Wuebbles, D. J., K. Kunkel, M. Wehner, and Z. Zobel. 2014. Severe Weather in United States
 Under a Changing Climate. Eos, Transactions American Geophysical Union 95:149–150.
- Yap, K. N., O. R. Kim, K. C. Harris, and T. D. Williams. 2017. Physiological effects of
 increased foraging effort in a small passerine. The Journal of Experimental Biology
 220:4282–4291.
- Zuur, A. F., E. N. Ieno, N. Walker, A. A. Saveliev, and G. M. Smith. 2009. Mixed effects models
 and extensions in ecology with R. Page Smart Society. Springer New York.

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