

1 **INSECT HERBIVORY ON URBAN TREES:** 2 **COMPLEMENTARY EFFECTS OF TREE** 3 **NEIGHBOURS AND PREDATION**

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19

20 **Abstract**

- 21 1. Insect herbivory is an important component of forest ecosystems functioning and can
22 affect tree growth and survival. Tree diversity is known to influence insect herbivory
23 in natural forest, with most studies reporting a decrease in herbivory with increasing
24 tree diversity. Urban ecosystems, on the other hand, differ in many ways from the
25 forest ecosystem and the drivers of insect herbivory in cities are still debated.
- 26 2. We monitored 48 urban trees from five species – three native and two exotic – in three
27 parks of Montreal (Canada) for leaf insect herbivory and predator activity on artificial
28 larvae, evaluated the relationships between herbivory, predation, and tree diversity in
29 the vicinity of focal trees.
- 30 3. Insect herbivory on leaves decreased with increasing tree diversity and with increasing
31 predator attack rate.
- 32 4. Our findings indicate that tree diversity is a key determinant of multitrophic interactions
33 between trees, herbivores and predators in urban environments and that managing tree
34 diversity could contribute to pest control in cities.

35

36 **Keywords** : Artificial prey, Insect herbivory, Tree diversity, Top-down control, Urban
37 biodiversity

38 Introduction

39 Insect herbivores have a major impact on tree growth and survival, hence on the functioning
40 of forest ecosystems (Metcalf et al., 2014; Visakorpi et al., 2018; Zvereva, Zverev, & Kozlov,
41 2012). Tree diversity significantly influences insect herbivory in forest ecosystems
42 (Castagneyrol, Jactel, Vacher, Brockerhoff, & Koricheva, 2014; Jactel et al., 2017). Most
43 studies report that herbivory declines as tree diversity increases (*i.e.*, associational resistance,
44 Barbosa et al., 2009), although the opposite pattern has also been found (Haase et al., 2015;
45 Schuldt et al., 2011). Recently, the interest in how tree diversity affects insect herbivory has
46 expanded to include urban forests (Clem & Held, 2018; Dale & Frank, 2018; Frank, 2014),
47 where pest damage can compromise the ecological and aesthetic values of urban trees (Nuckols
48 & Connor, 1995; Tooker & Hanks, 2000; Tubby & Webber, 2010). Urban forests differ from
49 natural forests in many ways. For example, most of the trees in cities are planted, found in
50 lower density and/or mixed with native and exotic ornamental species that are rarely
51 encountered in natural forests. Thus, given these specific characteristics of urban forests, it is
52 still unclear how and why tree diversity might influence insect herbivory on urban trees.

53 The density and diversity of trees determine the amount and the quality of food and habitat
54 resources available to herbivores and their enemies, and thus can have strong impact on the
55 bottom-up and top-down forces acting upon insect herbivores (Haase et al., 2015; Muiruri,
56 Rainio, & Koricheva, 2016; Setiawan, Vanhellefont, Baeten, Dillen, & Verheyen, 2014). For
57 example, some insect herbivores, in particular generalist species, could take advantage of tree
58 diversity to acquire more abundant, complementary food resources or benefit from a more
59 balanced food mix, thus causing more damage in mixed forests (Lefcheck, Whalen, Davenport,
60 Stone, & Duffy, 2013). In contrast, insect herbivores generally find it easier to identify and
61 orientate towards the signals emitted by their host trees when the latter are more concentrated
62 (*the resource concentration hypothesis*, Hambäck & Englund, 2005; Root, 1973) while non-
63 host trees can emit volatile compounds that interfere with the ability of herbivores to detect
64 their preferred host (Jactel, Birgersson, Andersson, & Schlyter, 2011). Finally, the abundance
65 and diversity of predatory birds and arthropods generally increases with plant density and
66 diversity, which would result in a better top-down regulation of insect herbivores (*the enemies*
67 *hypothesis*, Risch, Andow, & Altieri, 1983; Root, 1973). However, the evidence available to
68 support the enemies hypothesis in forest is controversial (Muiruri et al., 2016; Riihimäki,
69 Kaitaniemi, Koricheva, & Vehviläinen, 2005; Staab & Schuldt, 2020) and the contribution of
70 natural enemies to the control of herbivores in urban area remains poorly explored.

71 Tree diversity and density vary widely between and within cities (Ortega-Álvarez, Rodríguez-
72 Correa, & MacGregor-Fors, 2011; Sjöman, Östberg, & Bühler, 2012). A consequence of this
73 variability is that even within a common urban environment, herbivory may be reduced in some
74 tree species and increased in others (Clem & Held, 2018; Frank, 2014), and the relative
75 importance of bottom-up and top-down forces responsible for these effects may also differ. In
76 addition, non-native trees have been widely planted in urban habitats (Cowett & Bassuk, 2014;
77 Moro, Westerkamp, & de Araújo, 2014). While they often escape from herbivory by native

87 insects (*'the enemy escape hypothesis'*, Adams et al., 2009; Keane & Crawley, 2002), cases of
88 native herbivores spilling-over onto exotic trees have been recorded (e.g. Branco, Brouckhoff,
89 Castagnyrol, Orazio, & Jactel, 2015). Non-native tree species can also provide habitats to
90 insectivorous birds or predatory arthropods (Gray & van Heezik, 2016). It is thus difficult to
91 predict the effect of mixing native and exotic trees on insect herbivory in urban habitats (Clem
92 & Held, 2018; Frank, 2014).

84 In this study, we investigated the effect of tree density, tree diversity, presence of conspecific
85 trees, tree origin and predator activity on insect herbivory in urban trees of the city of Montreal
86 (Quebec, Canada). We measured leaf area removed or otherwise damaged by insect herbivores
87 on 48 trees of five species – three native and two exotic – in three urban parks. We
88 concomitantly assessed predator activity by using artificial caterpillars exposed on tree
89 branches. We tested the following hypotheses: (1) insect herbivory decreases with tree density,
90 number of non-conspecific trees (host dilution) and diversity (associational resistance) around
91 focal trees, (2) predator activity increases with increasing tree density and diversity and (3)
92 predation and herbivory have different responses to tree diversity on native and exotic trees.
93 By doing so, our study builds toward a better understanding of the drivers of pest insect damage
94 on urban trees.

95 **Materials and methods**

96 **Study site**

97 The study was conducted in the city of Montreal (Canada, 45°50'N, -73°55'W), where the
98 climate is temperate cold, with 6.8°C average temperature and 1000.3 mm annual rainfall
99 during the 1981-2010 period (Pierre Elliott Trudeau airport weather station, www.canada.ca).
100 The study took place in three parks of the southwest part of the city: Angrignon, Marguerite
101 Bourgeoys and Ignace-Bourget (Table 1).

102 **Tree selection**

103 Every tree in Angrignon, Ignace-Bourget and Marguerite-Bourgeoys parks had been
104 previously geolocalized and identified to the species level. This information was accessible
105 through the city database for urban trees (<http://donnees.ville.montreal.qc.ca/dataset/arbres>).
106 We selected a total of 48 trees of five deciduous species (Table 1). Three species are native to
107 the study area (*Acer saccharinum* L., *Tilia americana* L., *Quercus rubra* L.) while two are
108 exotics, from Europe (*Acer platanoides* L., *Tilia cordata* Mill.). These species are amongst the
109 most abundant tree species in the city of Montreal where together they represent 37% of all the
110 tree species of the public domain. In agreement with the city of Montreal administration, we
111 only selected trees with a diameter at breast height (DBH) greater than 8 cm (mean \pm SD: 18.38
112 \pm 9.36) (to withstand the sampling of leaves required for the experiment) and with low
113 branches that could be easily accessed using a stepladder (for safety).

Table 1. Mean (\pm SD) diameter at breast height (in cm) and number of trees initially selected for each park and species.

Species	Angrignon (AN) (45°26'N, -73°36'W)	Marguerite-Bourgeoys (MB) (45°47'N, -73°36'W)	Ignace-Bourguet (IB) (45°45'N, -73°60'W)
<i>Acer saccharinum</i>	37.55 (n = 5)	37.55 (n = 2)	15.1 (n = 2)
<i>Acer platanoides</i>	21.60 (n = 1)	23.68 \pm 1.97 (n = 6)	26.25 (n = 2)
<i>Tilia cordata</i>	22.40 (n = 1)	30.60 \pm 3.37 (n = 5)	9.67 \pm 0.51 (n = 4)
<i>Tilia americana</i>	10.52 \pm 0.55 (n = 4)	22.06 \pm 1.87 (n = 3)	27.60 \pm 1.20 (n = 3)
<i>Quercus rubra</i>	8.96 \pm 0.37 (n = 5)	NA	12.30 \pm 1.45 (n = 5)

114 Predation rate assessment

115 We used artificial caterpillars made with modelling clay to estimate predation rate on sampled
116 trees (Ferrante, Lo Cacciato, & Lovei, 2014; Howe, Lövei, & Nachman, 2009). We installed
117 15 artificial caterpillars per tree. We haphazardly selected three low (2.5-3.5 m from ground)
118 branches facing different directions and installed five artificial caterpillars per branch (total:
119 720 caterpillars). Caterpillars were 3 cm long, and modelled to match the approximate form
120 and size of real caterpillars. They were modelled using a 1-cm ball of non-toxic and odourless
121 green modelling clay (*Sculpey III String Bean colour*) and secured on thin branches using a 12-
122 cm long, 0.5 mm diameter, non-shiny metallic wire.

123 We exposed artificial caterpillars for 11 days in late spring (from May 29th to June 9th, 2018)
124 and for 6 days in early summer (from July 18th to July 24th, 2018). These seasons were chosen
125 to cover the main activity period of both predators and herbivores. Artificial caterpillars were
126 left untouched for the full duration of each survey. We estimated total predator attack rate as
127 the number of artificial larvae with any predation mark, divided by the total length of the
128 observation period in days. There were uncertainties regarding predator identity responsible
129 for predation marks. Most of the marks were attributable to birds or arthropods, while very few
130 were attributable to small mammals, therefore, we chose to combine predation marks primarily
131 attributed to birds or arthropods into a single category, which we refer to as total predation.

132 Branches of three trees were accidentally pruned by city workers in late spring so that the
133 predation rate could not be estimated on these trees for the first survey. Three new trees of the
134 same species were selected for the second survey, in early summer.

135 Leaf insect herbivory

136 We estimated insect herbivory on leaves (Kozlov et al., 2017) as the percentage of leaf area
137 removed or impacted by insect herbivores (including chewing, skeletonizing and mining
138 damage, collectively referred to as ‘herbivory’). At the end of the second predation survey, we
139 collected 10 leaves per branch on the same branches on which we had exposed artificial
140 caterpillars, starting with the most apical, fully-developed, leaf to the 10th leaf down to branch
141 basis (Total: 30 leaves per tree). We estimated total herbivory (i.e., total leaf area consumed or
142 impacted by herbivores, regardless of their identity) as well as damage made by chewing,
143 mining and sap-feeding herbivores at the level of individual leaves by using an ordinal scale of
144 eight percentage classes of defoliation: 0%; 0-1%, 1-5%; 6-10%; 11-25%; 26-50%; 51-75%
145 and 76-100%. We counted the number of galls per leaf. Most damage was made by leaf
146 chewers, while other damage had a skewed distribution, preventing detailed analyses for each
147 type of damage separately. We therefore analysed total herbivory by averaging herbivory at
148 the level of individual trees and using the median of each class of defoliation. Herbivory was
149 scored by a single observer (BC), who was blind to tree identity.

150 Tree neighbourhood

151 We used three variables to describe tree neighbourhood in a 20-m radius around each focal
152 tree: tree density (defined as the number of neighbouring trees in that radius), tree species
153 diversity (Shannon diversity index) and the number of conspecific trees around each focal tree.
154 Those variables were obtained using QGIS Geographic Information System software (QGIS
155 Development Team, 2018). Excluding focal tree species, the most common tree species in the
156 vicinity of focal trees were the smooth serviceberry (*Amelanchier leavis* Wiegand), the white
157 spruce (*Picea glauca* Voss), the green ash (*Fraxinus pennsylvanica* Marshall) and the eastern
158 cottonwood (*Populus deltoides* Marshall), all of them native to the region. We should note that,
159 as focal trees were not necessarily 20m or more apart, we could not avoid that some
160 “neighbour” trees were used in more than one neighbourhood, and some focal trees were also
161 within the neighbourhood of another focal tree.

162 Statistical analyses

163 We used the information theory framework to identify the best model fitting our data and
164 applied model averaging whenever necessary to estimate model coefficient parameters
165 (Grueber, Nakagawa, Laws, & Jamieson, 2011). We first built a full model including tree
166 density (*Density*), tree diversity (*Diversity*), number of conspecifics (*Conspecific*), origin of the
167 focal tree (*Origin*, native of exotic), park (*Park*), and predation rate (*Predation*) as fixed factors
168 and tree species identity (*Species*) as a random factor:

$$\begin{aligned} 169 \quad Y_{ij} \sim & \beta_0 + \beta_1 \times \text{Density}_{ij} + \beta_2 \times \text{Diversity}_{ij} + \\ 170 & \beta_3 \times \text{Conspecific}_{ij} + \beta_4 \times \text{Origin}_{\text{exotic}, ij} + \beta_5 \times \text{Park}_{\text{TB}, ij} + \\ 171 & \beta_6 \times \text{Park}_{\text{MB}, ij} + \beta_6 \times \text{Predation}_{ij} + \\ 172 & \gamma_j + \varepsilon_{ij} \end{aligned} \quad (1)$$

173
$$\gamma \sim N(0; \sigma_{\text{species}}^2) \quad \varepsilon \sim N(0; \sigma_e^2)$$

174 Where Y_{ij} is the herbivory on tree individual i in tree species j , β are model coefficient
175 parameters for fixed effects, γ_j is the random effect of tree species identity and ε the residuals.

176 To ease the interpretation of parameter estimates after model averaging, we standardized the
177 input variables using Gelman's approach (Gelman, 2008). We then applied a procedure of
178 model selection based on the Akaike's criterion corrected for small sample size (AICc) by
179 running every model nested within the full model. As tree density and tree diversity were
180 correlated (Pearson's correlation: $r = 0.71$), we excluded all sub-models that included these
181 predictors together. We ranked all models based on difference in AICc between each model
182 and the top ranked model with the lowest AICc (ΔAICc). Models with a $\Delta\text{AICc} < 2$ are
183 generally considered equally supported by the data or not differentiable from the top ranked
184 model. Finally, we estimated model fit by calculating marginal (R^2_m) and conditional (R^2_c) R^2
185 values, corresponding to variance explained by fixed effects only (R^2_m) and by fixed and
186 random effects (R^2_c) (Nakagawa & Schielzeth, 2013). When multiple models had a $\Delta\text{AICc} <$
187 2 , we used a model averaging approach to build a consensus model including all variables
188 found in the set of best models. We considered that a given predictor had a significant effect if
189 its 95% confidence interval did not overlap zero. When only one model had a $\Delta\text{AICc} < 2$, we
190 used it as the best model. We used a square-root transformation of insect herbivory to satisfy
191 model assumptions of normality and homogeneity of residuals.

192 We used the same approach to test the effect of tree neighbourhood on predation rate, log-
193 transforming predation rate to satisfy model assumptions. Model equation (2) included the
194 fixed effect of sampling season (*Season*) and the random effect of tree identity (τ_k), nested
195 within tree species identity as an additional random factor accounting for repeated
196 measurements of the same individuals:

197
$$Y_{ijk} \sim \beta_0 + \beta_1 \times \text{Density}_{ijk} + \beta_2 \times \text{Diversity}_{ijk} +$$

198
$$\beta_3 \times \text{Conspecific}_{ijk} + \beta_4 \times \text{Origin}_{\text{exotic}, ijk} + \beta_5 \times \text{Park}_{\text{IB}, ijk} +$$

199
$$\beta_6 \times \text{Park}_{\text{MB}, ijk} + \beta_6 \times \text{Season}_{\text{summer}, ijk} +$$

200
$$\gamma_j + \tau_{k|j} + \varepsilon_{ijk} \quad (2)$$

201
$$\gamma \sim N(0; \sigma_{\text{species}}^2) \quad \tau \sim N(0; \sigma_{\text{individual}}^2) \quad \varepsilon \sim N(0; \sigma_e^2)$$

202
203 Statistical analyses were performed using the R software version 3.4.4 (R Core Team 2019)
204 with packages *lme4* (Bates, Mächler, Bolker, & Walker, 2015) and *MuMIn* (Barton 2019).

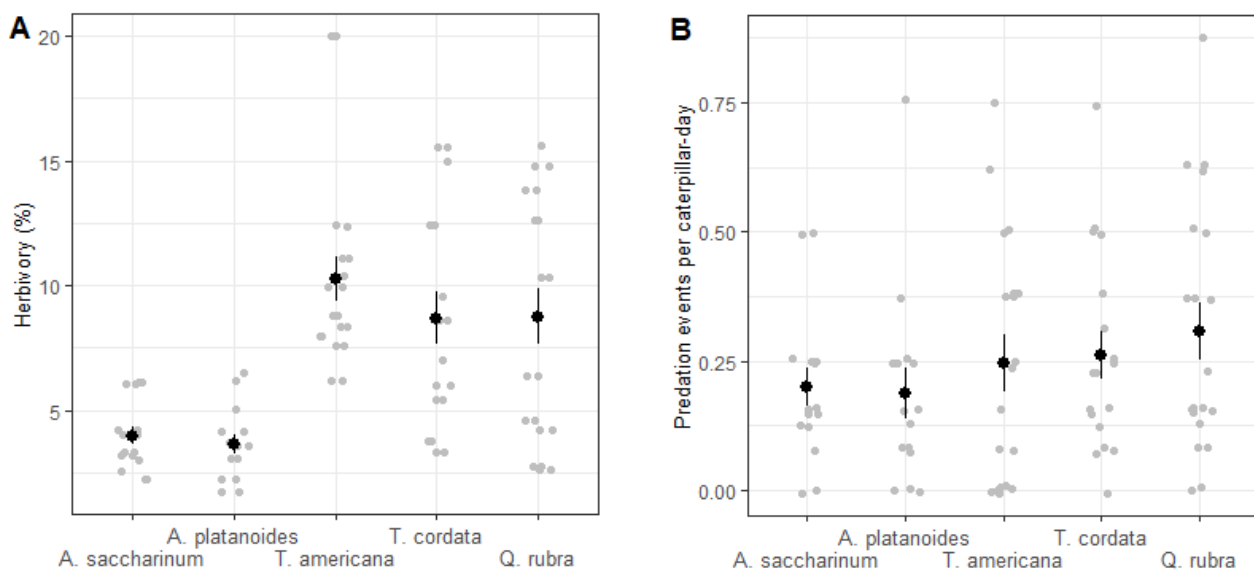
205 Results

206 **Insect herbivory** – Herbivory was on average (\pm SE) 7.19 ± 0.70 % ($n = 48$). Leaf damage was
207 lower in *Acer platanoides* (3.53 ± 0.54) and *A. saccharinum* (3.86 ± 0.47) than in *Quercus*
208 *rubra* (8.77 ± 1.65), *Tilia americana* (10.3 ± 1.37) and *T. cordata* (8.75 ± 1.75) (Fig. 1A).

209

210 There were six models competing with the top ranked model in a range of 2 units of $\Delta AICc$
211 (Table 2). These models included tree Shannon diversity, predation rate and tree origin as
212 predictors. Insect herbivory decreased significantly with increasing tree diversity (average
213 model coefficient parameter estimate \pm CI: $-0.482 \pm [-0.91; -0.05]$, Fig. 2A, Table 3) and with
214 increasing predation rate ($-0.473 \pm [-0.91; -0.003]$) (Fig. 2B, Table 3). Among the set of best
215 models, fixed effects explained between 7 and 12% of variability in insect herbivory. Fixed
216 and random effects together explained between 47 and 65% of variability in insect herbivory.

Figure 1. Effect of tree species identity on insect herbivory (A) and predation rate (B). Black dots and solid lines represents mean \pm SE calculated on raw data. Herbivory is the percentage of leaf area removed or impacted by herbivores in early summer. Predation events per caterpillar-day is the number of caterpillars attacked per day in late spring



217 **Predation** – Of the 1,315 artificial caterpillars that we installed, 198 displayed marks
218 unambiguously attributable to predators (*i.e.*, 15%). Predation rate varied between 0 and 0.87
219 per caterpillar-day (Fig. 1B).

220 Only one model had a $\Delta AICc < 2$ and was thus selected as the best model. This best model
221 included only Season, with predation rate two times higher in late spring (mean \pm CI: $0.44 \pm$
222 $[0.31, 0.58]$ caterpillars \cdot day $^{-1}$) than in early summer ($0.20 \pm [0.16, 0.24]$ caterpillars \cdot day $^{-1}$).
223 Season explained 56 % of variability in predation rate and, collectively, fixed and random
224 effects explained 59 % of variability in predation rate.

Table 2. Summary results of model selection of tree neighbourhood effect on herbivory rate: set of models with $\Delta AICc < 2$. Only predictors that were present at least once in the set of best models are represented. R^2m and R^2c represent fixed and fixed *plus* random factor, respectively.

Model	Model covariates				Model selection			
	Intercept	Predation	Origin	Diversity	K	Log L	$\Delta AICc$	R^2m (R^2c)
1	2.53			-0.52	1	-46.44	0.00	0.09 (0.46)
2	2.52	-0.52		-0.44	2	-45.18	0.04	0.12 (0.58)
3	2.51	-0.51			1	-46.79	0.70	0.07 (0.56)
4	2.53	-0.44	0.171	-0.43	3	-44.64	1.67	0.12 (0.65)
5	2.53		0.078	-0.52	2	-46.07	1.82	0.08 (0.53)
6	2.53	-0.53	0.357		2	-46.12	1.92	0.08 (0.62)

Table 3. Summary results after model averaging: effects of each parameter presents on the set of best models on herbivory rate. Bold parameter are significant. Relative importance is a measure of the prevalence of each parameter in each model used in model averaging.

Parameter	Estimate	Adjusted SE	Confidence interval	Relative importance
(Intercept)	2.53	0.31	(1.91, 3.14)	
Diversity	-0.48	0.21	(-0.91, -0.05)	0.72
Predation	-0.47	0.22	(-0.91, -0.003)	0.64
Origin	0.19	0.71	(-1.20, 1.60)	0.31

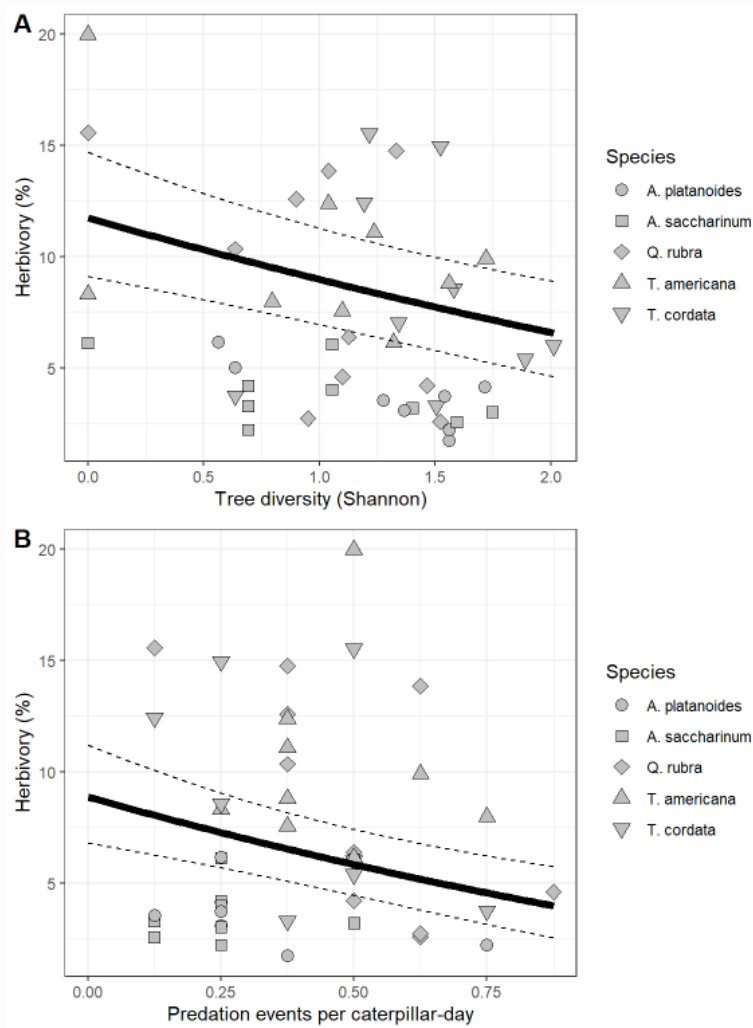


Figure 2. Effects of tree diversity (A) and predation rate (B) on insect herbivory. Solid and dashed lines represent prediction and adjusted standard error of the average model respectively (Table 3). Herbivory is the percentage of leaf area removed or impacted by herbivores in early summer. Tree diversity is represented by Shannon's diversity index. Predation events per caterpillar-day is the number of caterpillars attacked per day in late spring.

225 Discussion

226 We confirmed that tree diversity can influence insect herbivory on urban trees. Specifically,
227 we found that insect herbivory decreased with increasing tree diversity providing support for
228 the associational resistance hypothesis (Castagneyrol et al., 2017). We also found a negative
229 correlation between predator attack rate and insect herbivory. Although further analyses are
230 needed to confirm this relationship, our findings provide support for the view that increasing
231 tree diversity can enhance regulation of insect herbivores by natural enemies in urban forests.

232 Our results are in line with several studies having reported reduced herbivory in trees
233 surrounded by heterospecific neighbours (reviewed by Castagneyrol et al., 2014; Jactel et al.,
234 2017). It also adds to the growing number of studies documenting diversity-resistance
235 relationships in urban environments (Clem & Held 2018; Doherty, Meagher, & Dale 2019;

236 Frank 2014). However, it conflicts with other results suggesting an increase in herbivore
237 abundance with increasing plant diversity in urban environments (Mata et al., 2017), although
238 the relationship between herbivore abundance and actual herbivory is not always positively
239 correlated (Barbosa et al., 2009; Schueller, Paul, Payer, Schultze, & Vikas, 2019). Tree
240 diversity may have influenced the probability of focal trees being found and colonized by
241 herbivores. Theory predicts that specialist herbivores have greater difficulties finding their host
242 trees when they are surrounded by heterospecific neighbours (Castagneyrol et al., 2014; H.
243 Jactel, Brockerhoff, & Duelli, 2009). It is possible that non-host neighbours disrupted the
244 physical and chemical cues used by insect herbivores to locate their hosts (Damien et al., 2016;
245 H. Jactel et al., 2011; Zhang & Schlyter, 2004). However, and contrary to our expectations, we
246 did not find any significant effect of conspecific tree density on insect herbivory, thus ruling
247 out the resource concentration hypothesis in this particular case. However, because our study
248 was observational, we could not separate the effect of conspecific neighbour density from
249 heterospecific neighbour density. In the absence of data on the identity of herbivores
250 responsible for herbivory, further speculation would be hazardous.

251 Insect herbivory varied across tree species but did not differ between native and non-native
252 species, thus not providing support for predictions of the enemy release hypothesis (Cincotta,
253 Adams, & Holzapfel, 2009; Meijer, Schilthuizen, Beukeboom, & Smit, 2016). One possible
254 explanation for this result could be that native herbivores spilled over exotic tree species from
255 neighbouring native tree species, as it was recorded in previous studies (Branco et al., 2015).
256 This would have been facilitated by the fact that exotic tree species (from Europe) had
257 congeneric species in Canada. Although we only surveyed a handful of native and exotic
258 species, making any generalization hazardous, we can speculate on the lack of difference
259 between native and non-native species. It is also important to note that a large part of the
260 variability in leaf insect damage was attributable to the species on which leaf samples were
261 collected. In particular, both *Acer platanoides* and *A. saccharinum* were far less damaged than
262 *Tilia cordata*, *T. americana* and *Quercus rubra*. In a recent study in Michigan, Schueller et al.,
263 (2019) also reported greater insect herbivory (and herbivore diversity) on *Quercus* species as
264 compared to *Acer* species, which is consistent with the view that plant species identity can
265 drive arthropods community and abundance on forest host trees (Burghardt, Tallamy, &
266 Gregory Shriver, 2009; Pearse & Hipp, 2009).

267 We found a significant negative correlation between predator attack rate and insect herbivory
268 measured later in the season. This finding suggests a potential relationship between herbivory
269 and predation in urban environments (Faeth, Warren, Shochat, & Marussich, 2005; Kozlov et
270 al., 2017 but see Long & Frank, 2020). However, we refrain from concluding that predation
271 was the main driver of insect herbivory for several reasons. First, the effect size of the
272 herbivory-predation relationship was small, as was model R^2 (Table 3). Second, concerns
273 remain about how well predation on artificial prey represents of actual predation (Lövei &
274 Ferrante, 2017; Rößler, Pröhl, & Lötters, 2018). In particular, artificial caterpillars used to
275 assess predation rate modelled lepidopteran-like leaf chewing caterpillars and thus, caution is
276 needed when it comes to extrapolate predator attack rates to other herbivore feeding guilds.
277 Third, we had no information on actual natural prey density in focal and neighbouring trees.

278 Yet, prey availability may have influenced the functional response of bird insectivores (e.g.
279 optimal foraging) such that we cannot exclude that herbivory actually drove predation rate
280 instead of the other way around. Finally, the putative effect of predation on herbivory may be
281 weak in respect to other factors acting directly upon herbivores in urban environments such as
282 drought (Huberty & Denno, 2004; Mattson, 1980; Meineke & Frank, 2018), extreme heat (Dale
283 & Frank, 2014; Meineke, Dunn, Sexton, & Frank, 2013) and pollution leading to altered foliage
284 quality (Kozlov et al., 2017; Mattson, 1980; Moreira et al., 2019).

285 Contrary to the important effect of tree species identity on insect herbivory, tree species had no
286 clear influence on predation rate on dummy caterpillars, which contradicts the view that tree
287 species identity can modulate attack rates of caterpillars by birds (Mooney & Singer, 2012;
288 Nell et al., 2018). Variation in predator density between plants is often related to an indirect
289 effect of the plant on the density (Bailey et al. 2006) or quality (Brower et al. 1967, Clancy and
290 Price, 1987) of their preys (herbivores). However, such effect of plant identity is not relevant
291 when using dummy caterpillars, as neither their abundance nor their quality can be affected by
292 plant species identity, which could explain the contradiction between past results and our study.

293 Predation was greater during the first survey, in late spring, than during the second survey, in
294 early summer. This result could be explained either by a lower foliage density in trees in spring,
295 making it easier for predators to detect artificial caterpillars, or by greater predator activity
296 matching the phenology of wild caterpillars and feeding period of chicks (Coley, 1980; Raupp
297 & Denno, 1983). We cannot either exclude that birds learned to avoid artificial caterpillars,
298 thus resulting in much lower predation pressure during the second survey. However, unless
299 bird ability to avoid artificial caterpillar varied between tree species and neighbourhood, we do
300 not see this possibility as a major threat to our inferences.

301 Conclusion

302 Our study suggests several ecological factors drive leaf insect herbivory in the urban trees of
303 the Montreal city. In particular, we found that insect herbivory decreased with both increasing
304 tree diversity and predator activity. While biological invasions and global warming are
305 increasing risks to urban trees, more and more cities choose to ban or reduce the use of
306 pesticides in urban parks and green areas (Sustainable Use of Pesticides Directive 2009), such
307 as in Montreal. In this context, diversifying urban tree cover in urban parks might help to reduce
308 insect damage, which could result in a better provision of services provided by trees in cities
309 (Beyer et al., 2014; Bowler, Buyung-Ali, Knight, & Pullin, 2010; Nowak, Hirabayashi, Bodine,
310 & Greenfield, 2014).

311 Authors' contributions

312 BC and AP conceived the study. YK selected trees and coordinated field work. YK and BC
313 acquired data. AS, MLB and BC analysed the data. AS, MLB and BC wrote the first draft.
314 All authors contributed critically to the final manuscript.

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327 **Data accessibility**

328 Data and script available from the Data INRAE repository: <https://doi.org/10.15454/R4NESA>

329 **Conflict of interest disclosure**

330 The authors of this article declare that they have no financial conflict of interest with the content
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332 References

- 333 Adams, J. M., Fang, W., Callaway, R. M., Cipollini, D., Newell, E., & Transatlantic Acer
334 platanoides Invasion Network (TRAIN). (2009). A cross-continental test of the Enemy
335 Release Hypothesis: leaf herbivory on Acer platanoides (L.) is three times lower in North
336 America than in its native Europe. *Biological Invasions*, 11(4), 1005–1016. doi:
337 10.1007/s10530-008-9312-4
- 338 Akaike, H. 1973. Information theory as an extension of the maximum likelihood principle. *In*:
339 Second International Symposium on Information Theory (B.N. Petrov & F. Csaki, eds), pp.
340 267–281. *Akademiai Kiado, Budapest*.
- 341 Barbosa, P., Hines, J., Kaplan, I., Martinson, H., Szczepaniec, A., & Szendrei, Z. (2009).
342 Associational Resistance and Associational Susceptibility: Having Right or Wrong
343 Neighbors. *Annual Review of Ecology, Evolution, and Systematics*, 40(1), 1–20. doi:
344 10.1146/annurev.ecolsys.110308.120242
- 345 Bates, D., Mächler, M., Bolker, B., & Walker, S. (2015). Fitting Linear Mixed-Effects
346 Models Using lme4. *Journal of Statistical Software*, 67(1). doi: 10.18637/jss.v067.i01
- 347 Beyer, K. M. M., Kaltenbach, A., Szabo, A., Bogar, S., Nieto, F. J., & Malecki, K. M. (2014).
348 Exposure to Neighborhood Green Space and Mental Health: Evidence from the Survey of
349 the Health of Wisconsin. *International Journal of Environmental Research and Public
350 Health*, 11(3), 3453–3472. doi: 10.3390/ijerph110303453
- 351 Bowler, D. E., Buyung-Ali, L. M., Knight, T. M., & Pullin, A. S. (2010). A systematic
352 review of evidence for the added benefits to health of exposure to natural environments.
353 *BMC Public Health*, 10, 456. doi: 10.1186/1471-2458-10-456
- 354 Castagneyrol, B., Bonal, D., Damien, M., Jactel, H., Meredieu, C., Muiruri, E. W., &
355 Barbaro, L. (2017). Bottom-up and top-down effects of tree species diversity on leaf
356 insect herbivory. *Ecology and Evolution*, 7(10), 3520–3531. doi: 10.1002/ece3.2950
- 357 Castagneyrol, B., Jactel, H., Vacher, C., Brockerhoff, E. G., & Koricheva, J. (2014). Effects
358 of plant phylogenetic diversity on herbivory depend on herbivore specialization. *Journal
359 of Applied Ecology*, 51(1), 134–141. doi: 10.1111/1365-2664.12175
- 360 Cincotta, C. L., Adams, J. M., & Holzapfel, C. (2009). Testing the enemy release hypothesis:
361 a comparison of foliar insect herbivory of the exotic Norway maple (*Acer platanoides*
362 L.) and the native sugar maple (*A. saccharum* L.). *Biological Invasions*, 11(2), 379–388.
363 doi: 10.1007/s10530-008-9255-9
- 364 Clem, C. S., & Held, D. W. (2018). Associational Interactions Between Urban Trees: Are
365 Native Neighbors Better Than Non-Natives? *Environmental Entomology*, 47(4), 881–889.
366 doi: 10.1093/ee/nvy071
- 367 Coley, P. D. (1980). Effects of leaf age and plant life history patterns on herbivory. *Nature*,
368 284, 545–546. doi: 10.1038/284545a0
- 369 Cowett, F. D., & Bassuk, N. L. (2014). Statewide assessment of street trees in New York
370 State, USA. *Urban Forestry & Urban Greening*, 13(2), 213–220. doi:
371 10.1016/j.ufug.2014.02.001
- 372 Dale, A. G., & Frank, S. D. (2018). Urban plants and climate drive unique arthropod
373 interactions with unpredictable consequences. *Current Opinion in Insect Science*, 29,
374 27–33. doi: 10.1016/j.cois.2018.06.001
- 375 Dale, A. G., & Frank, S. D. (2014). Urban warming trumps natural enemy regulation of
376 herbivorous pests. *Ecological Applications: A Publication of the Ecological Society of
377 America*, 24(7), 1596–1607. doi: 10.1890/13-1961.1
- 378 Damien, M., Jactel, H., Meredieu, C., Régolini, M., van Halder, I., & Castagneyrol, B.
379 (2016) Pest damage in mixed forests: Disentangling the effects of neighbor identity, host

- 380 density and host apparency at different spatial scales. *Forest Ecology and Management*,
381 378, 103–110. doi: 10.1016/j.foreco.2016.07.025
- 382 Doherty, E. M., Meagher, R. L., & Dale, A. G. (2019). Turfgrass Cultivar Diversity Provides
383 Associational Resistance in the Absence of Pest Resistant Cultivars. *Environmental*
384 *Entomology*, 48(3), 623–632. doi: 10.1093/ee/nvz026
- 385 Faeth, S. H., Warren, P. S., Shochat, E., & Marussich, W. A. (2005). Trophic Dynamics in
386 Urban Communities. *BioScience*, 55(5), 399–407. doi:
387 10.1641/0006-3568(2005)055[0399:TDIUC]2.0.CO;2
- 388 Ferrante, M., Lo Cacciato, A., & Lövei, G. (2014). Quantifying predation pressure along an
389 urbanisation gradient in Denmark using artificial caterpillars. *European Journal of*
390 *Entomology*, 111(5), 649–654.
- 391 Frank, Steven D. (2014). Bad neighbors: urban habitats increase cankerworm damage to
392 non-host understory plants. *Urban Ecosystems*, 17(4), 1135–1145. doi:
393 10.1007/s11252-014-0368-x
- 394 Gelman, A. (2008). Scaling regression inputs by dividing by two standard deviations.
395 *Statistics in Medicine*, 27(15), 2865–2873. doi: 10.1002/sim.3107
- 396 Grueber, C. E., Nakagawa, S., Laws, R. J., & Jamieson, I. G. (2011). Multimodel inference in
397 ecology and evolution: challenges and solutions. *Journal of Evolutionary Biology*,
398 24(4), 699–711. doi: 10.1111/j.1420-9101.2010.02210.x
- 399 Hambäck, P. A., & Englund, G. (2005). Patch area, population density and the scaling of
400 migration rates: the resource concentration hypothesis revisited. *Ecology Letters*, 8(10),
401 1057–1065. doi: 10.1111/j.1461-0248.2005.00811.x
- 402 Howe, A., Lövei, G. L., & Nachman, G. (2009). Dummy caterpillars as a simple method to
403 assess predation rates on invertebrates in a tropical agroecosystem. *Entomologia*
404 *Experimentalis et Applicata*, 131(3), 325–329. doi: 10.1111/j.1570-7458.2009.00860.x
- 405 Huberty, A. F., & Denno, R. F. (2004). Plant Water Stress and Its Consequences for
406 Herbivorous Insects: A New Synthesis. *Ecology*, 85(5), 1383–1398.
- 407 Jactel, H., Birgersson, G., Andersson, S., & Schlyter, F. (2011). Non-host volatiles mediate
408 associational resistance to the pine processionary moth. *Oecologia*, 166(3), 703–711. doi:
409 10.1007/s00442-011-1918-z
- 410 Jactel, H., Brockerhoff, E., & Duelli, P. (2009). A Test of the Biodiversity-Stability Theory:
411 Meta-analysis of Tree Species Diversity Effects on Insect Pest Infestations, and
412 Re-examination of Responsible Factors. In M. Scherer-Lorenzen, C. Körner, & E.-D.
413 Schulze (Eds.), *Forest Diversity and Function: Temperate and Boreal Systems* (pp. 235–
414 262). doi: 10.1007/3-540-26599-6_12
- 415 Staab, M., & Schuldt, A. (2020). The Influence of Tree Diversity on Natural Enemies—a
416 Review of the “Enemies” Hypothesis in Forests. *Current Forestry Reports*.
417 doi: 10.1007/s40725-020-00123-6
- 418 Jactel, H., Bauhus, J., Boberg, J., Bonal, D., Castagneyrol, B., Gardiner, B., ... Brockerhoff,
419 E. G. (2017). Tree Diversity Drives Forest Stand Resistance to Natural Disturbances.
420 *Current Forestry Reports*, 3(3), 223–243. doi: 10.1007/s40725-017-0064-1
- 421 Keane, R., & Crawley, M. J. (2002). Exotic plant invasions and the enemy release hypothesis.
422 *Trends in Ecology & Evolution*, 17(4), 164–170. doi: 10.1016/S0169-5347(02)02499-0
- 423 Kozlov, M. V., Lanta, V., Zverev, V., Rainio, K., Kunavin, M. A., & Zvereva, E. L. (2017).
424 Decreased losses of woody plant foliage to insects in large urban areas are explained by
425 bird predation. *Global Change Biology*, 23(10), 4354–4364. doi: 10.1111/gcb.13692
- 426 Kozlov, M. V., Zverev, V., & Zvereva, E. L. (2017). Combined effects of environmental
427 disturbance and climate warming on insect herbivory in mountain birch in subarctic
428 forests: Results of 26-year monitoring. *Science of The Total Environment*, 601–602, 802–
429 811. doi: 10.1016/j.scitotenv.2017.05.230

- 430 Lefcheck, J. S., Whalen, M. A., Davenport, T. M., Stone, J. P., & Duffy, J. E. (2013).
431 Physiological effects of diet mixing on consumer fitness: a meta-analysis. *Ecology*, 94(3),
432 565-572.
- 433 Long, L. C., & Frank, S. D. (2020). Risk of bird predation and defoliating insect abundance
434 are greater in urban forest fragments than street trees. *Urban Ecosystems*. doi:
435 10.1007/s11252-020-00939-x
- 436 Lövei, G. L., & Ferrante, M. (2017). A review of the sentinel prey method as a way of
437 quantifying invertebrate predation under field conditions. *Insect Science*, 24(4), 528–542.
438 doi: 10.1111/1744-7917.12405
- 439 Mata, L., Threlfall, C. G., Williams, N. S. G., Hahs, A. K., Malipatil, M., Stork, N. E., &
440 Livesley, S. J. (2017). Conserving herbivorous and predatory insects in urban green
441 spaces. *Scientific Reports*, 7. doi: 10.1038/srep40970
- 442 Mattson, W. J. (1980). Herbivory in Relation to Plant Nitrogen Content. *Annual Review of*
443 *Ecology and Systematics*, 11(1), 119–161. doi: 10.1146/annurev.es.11.110180.001003
- 444 Meijer, K., Schilthuizen, M., Beukeboom, L., & Smit, C. (2016). A review and meta-analysis
445 of the enemy release hypothesis in plant-herbivorous insect systems. *PeerJ*, 4, e2778. doi:
446 10.7717/peerj.2778
- 447 Meineke, E. K., Dunn, R. R., Sexton, J. O., & Frank, S. D. (2013). Urban warming drives
448 insect pest abundance on street trees. *PloS One*, 8(3), e59687. doi:
449 10.1371/journal.pone.0059687
- 450 Meineke, E. K., & Frank, S. D. (2018). Water availability drives urban tree growth responses
451 to herbivory and warming. *Journal of Applied Ecology*, 55(4), 1701–1713.
452 doi:10.1111/1365-2664.13130
- 453 Metcalfe, D. B., Asner, G. P., Martin, R. E., Espejo, J. E. S., Huasco, W. H., Amézquita, F. F.
454 F., ... Malhi, Y. (2014). Herbivory makes major contributions to ecosystem carbon and
455 nutrient cycling in tropical forests. *Ecology Letters*, 17(3), 324–332. doi:
456 10.1111/ele.12233
- 457 Mooney, K.A. & Singer, M.S. (2012). Plant effects on herbivore-enemy interactions in natural
458 systems. In *Trait-Mediated Indirect Interactions: Ecological and Evolutionary*
459 *Perspectives* (eds Ohgushi, T., Schmitz, O., Holt, R.D.). Cambridge University Press,
460 Cambridge, pp. 107–130.
- 461 Moreira, X., Abdala-Roberts, L., Teran, J. C. B. M. y, Covelo, F., Mata, R. de la, Francisco,
462 M., Tack, A. J. M. (2019). Impacts of urbanization on insect herbivory and plant
463 defences in oak trees. *Oikos*, 128(1), 113–123. doi: 10.1111/oik.05497
- 464 Moro, M. F., Westerkamp, C., & de Araújo, F. S. (2014). How much importance is given to
465 native plants in cities' treescape? A case study in Fortaleza, Brazil. *Urban Forestry &*
466 *Urban Greening*, 13(2), 365–374. doi: 10.1016/j.ufug.2014.01.005
- 467 Muiruri, E. W., Rainio, K., & Koricheva, J. (2016). Do birds see the forest for the trees?
468 Scale-dependent effects of tree diversity on avian predation of artificial larvae.
469 *Oecologia*, 180(3), 619–630. doi: 10.1007/s00442-015-3391-6
- 470 Nakagawa, S., & Schielzeth, H. (2013). A general and simple method for obtaining R² from
471 generalized linear mixed-effects models. *Methods in Ecology and Evolution*, 4(2), 133–
472 142. doi: 10.1111/j.2041-210x.2012.00261.x
- 473 Nell, C., Abdala-Roberts, L., Parra-Tabla, V. & Mooney, K. (2018). Tropical tree diversity
474 mediates foraging and predatory effects of insectivorous birds. *Proc. R Soc. B*, 285,
475 20181842
- 476 Nowak, D. J., Hirabayashi, S., Bodine, A., & Greenfield, E. (2014). Tree and forest effects on
477 air quality and human health in the United States. *Environmental Pollution (Barking,*
478 *Essex:1987)*, 193, 119–129. doi: 10.1016/j.envpol.2014.05.028
- 479 Nuckols, M. S., & Connor, E. F. (1995). Do trees in urban or ornamental plantings receive

- 480 more damage by insects than trees in natural forests? *Ecological Entomology*, 20(3), 253–
481 260. doi: 10.1111/j.1365-2311.1995.tb00455.x
- 482 Ortega-Álvarez, R., Rodríguez-Correa, H. A., & MacGregor-Fors, I. (2011). Trees and the
483 City: Diversity and Composition along a Neotropical Gradient of Urbanization.
484 *International Journal of Ecology*, doi: <https://doi.org/10.1155/2011/704084>
- 485 Pearse, I. S., & Hipp, A. L. (2009). Phylogenetic and trait similarity to a native species
486 predict herbivory on non-native oaks. *Proceedings of the National Academy of Sciences*,
487 106(43), 18097–18102. doi: 10.1073/pnas.0904867106
- 488 QGIS Development Team (2019). QGIS Geographic Information System. Open Source
489 Geospatial Foundation Project.
- 490 R Core Team (2019). R: A language and environment for statistical computing. R Foundation
491 for Statistical Computing, Vienna, Austria. URL <https://www.R-project.org/>.
- 492 Raupp, & Denno. (1983). Leaf age as a predictor of herbivore distribution and abundance
493 [Herbivorous insects including the willow leaf beetle]. In *Academic Press. Variables*
494 *Plants and Herbivores in Natural and Managed Systems*.
- 495 Revision of World Urbanization Prospects | Multimedia Library - United Nations
496 Department of Economic and Social Affairs Retrieved May 30, 2019, from
497 [https://www.un.org/development/desa/publications/2018-revision-of-world-urbanization-](https://www.un.org/development/desa/publications/2018-revision-of-world-urbanization-prospects.html)
498 [prospects.html](https://www.un.org/development/desa/publications/2018-revision-of-world-urbanization-prospects.html)
- 499 Riihimäki, J., Kaitaniemi, P., Koricheva, J., & Vehviläinen, H. (2005). Testing the enemies
500 hypothesis in forest stands: the important role of tree species composition. *Oecologia*,
501 142(1), 90–97. doi: 10.1007/s00442-004-1696-y
- 502 Root, R. B. (1973). Organization of a Plant-Arthropod Association in Simple and Diverse
503 Habitats: The Fauna of Collards (*Brassica Oleracea*). *Ecological Monographs*, 43(1),
504 95–124. doi: 10.2307/1942161
- 505 Sjöman, H., Östberg, J., & Bühler, O. (2012). Diversity and distribution of the urban tree
506 population in ten major Nordic cities. *Urban Forestry & Urban Greening*, 11(1), 31–39.
507 doi: 10.1016/j.ufug.2011.09.004
- 508 Sustainable use of pesticides. Retrieved May 30, 2019, from
509 https://ec.europa.eu/food/plant/pesticides/sustainable_use_pesticides_en
- 510 Tubby, K. V., & Webber, J. F. (2010). Pests and diseases threatening urban trees under a
511 changing climate. *Forestry: An International Journal of Forest Research*, 83(4), 451–
512 459. doi: 10.1093/forestry/cpq027
- 513 Visakorpi, K., Gripenberg, S., Malhi, Y., Bolas, C., Oliveras, I., Harris, N., ... Riutta, T.
514 (2018). Small-scale indirect plant responses to insect herbivory could have major impacts
515 on canopy photosynthesis and isoprene emission. *The New Phytologist*, 220(3), 799–810.
516 doi: 10.1111/nph.15338
- 517 Zhang, Q.-H., & Schlyter, F. (2004). Olfactory recognition and behavioural avoidance of
518 angiosperm nonhost volatiles by conifer-inhabiting bark beetles. *Agricultural and Forest*
519 *Entomology*, 6(1), 1–20. doi: 10.1111/j.1461-9555.2004.00202.x
- 520 Zvereva, E. L., Zverev, V., & Kozlov, M. V. (2012). Little strokes fell great oaks: minor but
521 chronic herbivory substantially reduces birch growth. *Oikos*, 121(12), 2036–2043. doi:
522 10.1111/j.1600-0706.2012.20688.x