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8 Identifying drivers of sewage-associated pollutants in pollinators across urban landscapes

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34 **ABSTRACT:** Human sewage can introduce pollutants into food webs and threaten ecosystem
35 integrity. Among the many sewage-associated pollutants, pharmaceuticals and personal care
36 products (PPCPs) are useful indicators of sewage in ecosystems and can also cause potent
37 ecological consequences even at minute concentrations (e.g., ng/L). Despite increased study over
38 the past three decades, PPCPs in terrestrial systems have been less studied than those in aquatic
39 ecosystems. To evaluate PPCP prevalence and drivers in a terrestrial ecosystem, we analyzed
40 managed and native bees collected from agroecosystems in Washington State (USA) for PPCPs.
41 Caffeine, paraxanthine, cotinine, and acetaminophen were detected in all three evaluated taxa
42 (*Bombus vosnesenskii*, *Agapostemon texanus*, and *Apis mellifera*), with *B. vosnesenskii* and *A.*
43 *texanus* having a higher probability of PPCP detection relative to *A. mellifera*. The probability
44 for PPCP presence in all three taxa increased in landscapes with more human development or
45 greater plant abundance, with significant but negative interactions among these factors. These
46 results suggest that human activity, availability of resources, and species-specific traits affect the
47 introduction and mobilization of PPCPs in terrestrial ecosystems. Consequently, monitoring
48 PPCPs and their ecological responses in terrestrial ecosystems creates opportunities to synthesize
49 consequences of sewage pollution across terrestrial and aquatic ecosystems and organism types.

50

51 **KEYWORDS:** Arthropoda, emerging contaminants, entomology, natural history, Persistent
52 Organic Pollutants

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54

55 INTRODUCTION

56 The introduction of wastewater and its byproducts into ecosystems can mobilize pollutants that
57 reshape communities and food webs (Edmondson 1970). Historically, research on wastewater
58 pollution has largely focused on changes in effluent nutrient concentrations as well as chemical
59 and biological oxygen demand (Edmondson 1970; Brydon and Frodsham 2001; Tong et al.
60 2020). Recently, research emphasis has broadened to include micropollutants that are often
61 found in sewage (Bernhardt et al. 2017). Pharmaceuticals and personal care products (PPCPs) in
62 particular have garnered increased attention as an emerging organic micropollutant because they
63 are consistently associated with human sewage and pose potent, yet often uncertain, ecological
64 consequences (Richmond et al. 2017; Meyer et al. 2019).

65 As PPCPs are consistently associated with sewage, their presence, even at minute
66 concentrations (e.g., ng/L), can indicate wastewater inputs into an ecosystem. Previous
67 continental and worldwide surveys show that PPCPs such as antibiotics, non-prescription and
68 prescription drugs, hormones, and fragrances are pervasive within subsurface and surface
69 systems (Kolpin et al. 2002; Focazio et al. 2008; Wilkinson et al. 2022). Once introduced into
70 ecosystems, PPCPs can propagate through food webs, where they can be metabolized (del Rey et
71 al. 2011), accumulate within organisms (Meador et al. 2016), and be transferred across trophic
72 levels (Richmond et al. 2018). However, patterns of PPCP prevalence across taxa can be related
73 to individual behavior and tissue allocation (Meador et al. 2016), and biological responses to
74 PPCPs are often uncertain and context-dependent. For example, activity and feeding rates of
75 individual fish (*Perca fluviatilis*; European perch) increased when they were exposed to the anti-
76 anxiolytic drug oxazepam, but sociality of fish populations decreased (Brodin et al. 2013).
77 Certain algal taxa have been shown to have reduced photosynthesis and increased 16-carbon

78 unsaturated fatty acid production when exposed to fluoxetine, even though essential fatty acid
79 synthesis was not affected (Feijão et al. 2020). Bumble bees fed nectar primed with caffeine have
80 shown increased foraging behavior on nectars with similar aromatic compounds (Wright et al.
81 2013; Arnold et al. 2021). Overall, the literature suggests that biological responses to PPCPs are
82 common and often deleterious, although specific responses may also be uncertain (Richmond et
83 al. 2017).

84 Although considerable evidence from aquatic systems suggests that PPCPs are ubiquitous
85 and disrupt ecological processes, comparatively few studies on PPCPs have been conducted in
86 terrestrial systems (Meyer et al. 2019). This imbalance in the literature creates opportunity to
87 assess how PPCPs may propagate through terrestrial food webs in comparison to aquatic food
88 webs. For example, bees are exclusively terrestrial insect taxa that provide pollination services in
89 natural and managed ecosystems (Kleijn et al. 2015). As has been observed in earthworms
90 (Carter et al. 2021), bee pollinators may be commonly exposed to PPCPs through soil contact,
91 interactions with plants, applications of biosolids for fertilizer, or through contamination of water
92 in terrestrial ecosystems. Due to the dramatic, recent declines in bee pollinators within terrestrial
93 food webs worldwide (Potts et al. 2010; Kleijn et al. 2015), bee taxa may be promising,
94 societally important model organisms for expanding PPCP research from aquatic to terrestrial
95 ecosystems.

96 To assess PPCP prevalence and mobilization in pollinators within exclusively terrestrial
97 food webs, we evaluated the presence and drivers of PPCPs in three bee species: *Bombus*
98 *vosnesenskii*, *Agapostemon texanus*, and *Apis mellifera*. Two of these species are wild (*B.*
99 *vosnesenskii*, *A. texanus*), but *A. mellifera* (honey bee) is managed by humans. Our first goal was
100 to identify species-specific patterns of PPCP presence. We predicted that taxa interacting with

101 the soil (ground-nesting bees; i.e., *B. vosnesenskii*, *A. texanus*) would be more frequently
102 associated with PPCPs than managed, colony-forming species (i.e., *A. mellifera*), as ground-
103 nesting taxa would more likely encounter PPCPs within groundwater and biosolids, similar to
104 interactions observed in earthworms (Carter et al. 2021). Our second goal was to assess the
105 potential drivers of PPCP presence in each bee taxa. We predicted that PPCP presence would
106 increase at sites with greater human development and with a higher density of floral plant
107 resources at the sampling location, both of which can be potentially concentrated sources of
108 PPCPs (Carter et al. 2021; Meyer et al. 2022). Overall, our study provides some of the first
109 evidence for PPCP uptake in exclusively terrestrial bee species, while also demonstrating that
110 both landscape-context and organismal traits may affect exposure to and uptake of PPCPs.

111

112 MATERIALS AND METHODS

113 **Study System.** Our study of PPCPs in bees was nested within a study of bee ecology on 36
114 small (< 25 ha), diversified farms and community gardens (e.g., farms and gardens with more
115 than five unique flowering crops in simultaneous production) in western Washington (USA)
116 (Bloom et al. 2021). For the present study, bees were collected from a subset of 10 urban gardens
117 that were selected along an urbanization gradient (Fig. 1). Each garden was at least 1 km apart
118 for spatial independence (Bloom et al. 2021; Desaegher et al. 2022).

119

120 **Bee and plant surveys.** Bees were collected at each site using three blue vane traps
121 (SpringStar LLC, Woodinville, WA, USA) and 15 bee bowls placed along a 50 m transect. All
122 collected bees were preserved, pinned, and then identified to species (Bloom et al. 2021). Of the
123 6,539 specimens collected, we selected 101 specimens with approximately balanced specimens

124 from each of the three bee species (*B. vosnesenskii*, *A. texanus*, and *A. mellifera*); these taxa
125 reflect variation in social behavior (social vs. solitary taxa) and nesting strategies (i.e., cavity vs.
126 ground-nesting taxa). Because bee species were collected across three years (2014, 2015, 2016)
127 and three time points (spring, summer, and fall), we selected pinned specimens for this analysis
128 to have a balanced number of individuals across the sampling period. Specimen selection was
129 stratified by species, year, early/mid/late-summer timing, and sampling location, where each
130 species-timing-year-site combination included at least two individuals. As each species-timing-
131 year-site combination contained numerous individuals from the original sampling events,
132 personnel haphazardly selected individuals within a species-timing-site-year stratum.

133 To assess the role of plant communities in affecting PPCP presence in bees, plant richness
134 and abundance were measured at sites on the same date as bee sampling. Plants with flowers
135 serving as resources for bee visitation (i.e., having pollen and nectar) were recorded along the
136 same transect used for bee collection. A portable 1 × 1 m plot was placed over vegetation at 5 m
137 intervals and all plants with open flowers were identified to species. Each transect moved in a
138 serpentine fashion across each study site. Summarized field variables are detailed in Table S1.

139

140 **Landscape context.** As sites were located along an urbanization gradient, we determined
141 the amount of developed landscape within 1 km of each site using the United States Department
142 of Agriculture cropland data layer (CDL). Proportion development was defined as the count of
143 developed pixels within 1 km of each site divided by the total number of all pixels. Each pixel
144 within the CDL classifies a 30 × 30 m area as a landscape class, with 255 possible
145 classifications. To characterize all human development within a landscape, we summed across
146 low, moderate, and high intensity pixels (Table S1) (Han et al. 2014).

147

148 **PPCP extraction and quantification.** Each dried bee sample was massed and then ground
149 with mortar/pestle. Bee PPCPs were extracted using a three-phase sequential extraction, similar
150 to those described previously in Brodin et al. (2013) and Furlong et al. (2008). First, 1.5 mL of
151 methanol/water mix (7:3 ratio) with 0.1% formic acid was added into the mortar with the bee
152 parts and poured into a glass culture tube. An additional 1.5 mL methanol/water mix with 0.1%
153 formic acid was used to rinse the mortar and pestle, and then was poured into the same centrifuge
154 tube. Tubes with tissue and extract were centrifuged at 2,000 rpm for 5 minutes. Following
155 centrifugation, supernatant was then added to a clean centrifuge tube, and then immediately
156 wrapped in parafilm. A second 1.5 mL acetonitrile was next added to the first tube containing the
157 bee tissue. The tube containing bee tissue was then vortexed and centrifuged twice. After each of
158 the three extraction phases, the sample was placed under nitrogen flow in a 40°C bath. Once
159 samples were nearly evaporated completely, 1mL of formate buffer was added to each sample,
160 and the concentrated extract was transferred to a 1.5 mL amber glass autosampler vial. Samples
161 were preserved in the dark at -20°C until being analyzed with HPLC/MS.

162

163 **HPLC/MS Quantification.** PPCP identification and enumeration followed methods
164 similar to Furlong et al (2008) and Brodin et al. (2013). Standards for each PPCP are described
165 in table SM1. HPLC eluents included a 10-mM formate buffer and 100% acetonitrile solution
166 that varied in percent contribution over the quantification procedure (Table SM1). Minimal
167 detection limits were estimated to be 5 ng/g. A main difference between our methods and others
168 prior is that we split analyte signatures into separate channels, so as to avoid peak interference

169 (described in Table SM1). Mass spectrometer time-programmed operating conditions for
170 individual compounds are detailed in Table SM2-3.

171 Samples were also analyzed in a manner to account for potential cross-sample
172 contamination and peak drift. Following standard samples, 2 blanks of 100% methanol were
173 processed and assessed for contamination. Following the two blank samples, bee samples were
174 processed in batches of 10 followed by 1 blank, 1 standard (100 ng/L), and 1 blank. This routine
175 allowed us to purge the column from potential cross-sample contamination, to assess if samples
176 were contaminating downstream samples within a batch, and to control for peak drift.

177 Following peak quantification, we noticed that concentrations tended to be bimodal,
178 where PPCPs were either detected in higher concentrations or not detected (Figure 2). We
179 assumed that the bimodality of these detections was likely a product of low-to-intermediate
180 PPCP concentrations degrading since the time of specimen collection or diffusing from bee
181 tissues as samples were originally collected in an ethanol solution before pinning (Bloom et al.,
182 2021). In order to provide conservative estimates of bee PPCP presence, we reduced
183 concentrations into categorical presence/absence informatics, such that subsequent models and
184 model interpretations should be conservative.

185

186 **Quality Assurance and Quality Control Procedures.** During PPCP extraction and
187 quantification, an internal standard was unintentionally not included within each sample. To
188 account for potential biases and contamination that might arise during the PPCP extraction and
189 quantification process, we developed quality assurance and quality control protocols to ensure
190 high confidence of model results.

191 Throughout the entirety of the sampling process, laboratory personnel refrained from
192 consuming caffeinated beverages, nicotine products, and non-prescription medications.
193 Personnel wore an N95 mask and nitrile gloves during extraction to reduce chance of
194 contamination during extraction.

195 Samples were double-blind throughout the entire specimen selection, PPCP extraction,
196 peak quantification, and modeling procedures. Personnel extracting PPCPs from bees were only
197 aware of “Sample ID” for each specimen, which was written on the autosampler vial.
198 Experimental autosampler vials were haphazardly placed in the auto-sampler to reduce
199 likelihood of biases due to order in the auto-sampler. A separate laboratory personnel analyzed
200 ion spectra. Throughout the entire spectra analysis, the technician only knew the “Sample ID”
201 and not any metadata associated with that “Sample ID”. Lastly, a third person modeled the data
202 from HPLC/MS quantification. Through each step, experimental personnel only knew samples
203 by a unique “Sample ID” and were not aware of any attributes associated with that sample.
204 Together, this scheme was intentionally designed to potentially eliminate personnel-specific
205 biases throughout the extraction and analysis process.

206

207 **Modeling PPCP presence.** We iterated through all combinations of logistic regression
208 models to predict PPCP presence based on six predictor variables: bee and plant abundance, bee
209 and plant diversity, percent development, bee taxon, and all two-way interactions between these
210 variables (Katz et al. 2015; Rheubert et al. 2020). To account for edge effects, all variables were
211 standardized by the perimeter:area ratio of their respective site. Once all models were generated,
212 we selected the best model based on AICc, R^2 , AUC, accuracy (percent true results), and p-
213 value. When multiple models had similar performance (AICc within two points of lowest AICc

214 value and R^2 , AUC, and accuracy above the median of all models), best performing models were
215 averaged to create a single statistical model. To validate our results, we repeated analyses 1,000
216 times with 80:20 train:test subsetted data, and compared the distribution of model parameters and
217 pseudo- R^2 values from subsetted data to those models generated with the entire dataset. This
218 subsetting routine allowed us to assess whether the distribution of possible model parameters
219 was multi-modal and to evaluate whether the model constructed with 100% of the training data
220 was overfit to those data (Rheubert et al. 2020).

221 All analyses were conducted within the R Statistical Environment (R Core Team 2022)
222 using the packages `glmulti` (Calcagno 2019), `tidyverse` (Wickham et al. 2019), `janitor` (Firke
223 2020), `MuMIn` (Barton 2020), `ggpubr` (Kassambara 2019), `ggeffects` (Lüdecke 2018), `lubridate`
224 (`Grolemund` and Wickham 2011), `plotrix` (Lemon 2006), `Hmisc` (Jr et al. 2020), `corrplot` (Wei
225 and Simko 2017), `ggrepel` (Slowikowski 2019), `ggspatial` (Dunnington 2021), `tigris` (Walker
226 2021), `cowplot` (Wilke 2019), `sf` (Pebesma 2018), and `readxl` (Wickham and Bryan 2019).

227 RESULTS

228 **Species-specific PPCP detections.** We detected four PPCPs across all three bee species:
229 caffeine, paraxanthine/1,7-dimethylxanthine, acetaminophen/paracetamol, and cotinine. We did
230 not detect evidence of seven other PPCPs: codeine, warfarin, trimethoprim, sulfamethoxazole,
231 diphenhydramine, thiabendazole, or albuterol. *Bombus vosnesenskii* (50% of samples; Table S2)
232 and *A. texanus* (44%; Table S2) tended to have a higher odds ratio of PPCP presence relative to
233 *A. mellifera* (26%; Table S2).

234

235 **Relating PPCP presence with human development and plant abundance.** Across all
236 bee taxa, the probability of PPCP presence increased in landscapes with greater human

237 development or higher plant abundance (Fig. 2). However, there was a significant negative
238 interaction between human development and plant abundance, suggesting the positive
239 association between development and PPCP presence decreased in landscapes with greater plant
240 abundance; and, the positive effects of plant abundance on PPCP presence decreased in
241 landscapes with greater development (Fig. 2, Fig. S1).

242 Our subsetting routine, which assessed the probability of observing the final model solely
243 by chance, suggested our final model coefficients were non-random (Fig. 3). Model parameter
244 estimates (Figure 3) as well as pseudo- R^2 values (Figure 4) were generally unimodal. Most
245 subsetted model runs included human development and plant abundance as positive predictors of
246 PPCP presence (Figure 3), and most coefficients for interactions between plant abundance and
247 development were negative (Figure 3). Coefficients from the subsetting routine also
248 corresponded with patterns of PPCP detections for each taxon, where *B. vosnesenskii* had higher
249 positive coefficients, and *A. mellifera* and *A. texanus* tended to have lower negative coefficients
250 (Figure 3).

251

252 **DISCUSSION**

253 Our study demonstrates the presence of PPCPs in a terrestrial ecosystem and highlights that
254 pinned insect specimens can be used along with environmental data to understand patterns of
255 PPCP prevalence. These results are notable, considering that the study of PPCPs in terrestrial
256 ecosystems is uncommon relative to aquatic environments (Meyer et al. 2019). Our results
257 corroborate those results from aquatic systems, showing that PPCP presence can be influenced
258 by human development and species-specific traits (Bendz et al. 2005; Meador et al. 2016) and

259 that PPCPs can be mobilized in terrestrial food webs (Lagesson et al. 2016; Richmond et al.
260 2018).

261 Our results suggest PPCPs are more prevalent in *B. vosnesenskii* and *A. texanus* than in *A.*
262 *mellifera* (Table S2), implying that differences in species' life history can affect PPCP
263 accumulation. Both *B. vosnesenskii* and *A. texanus* create nests in the soil, which may increase
264 exposure to human contaminants in the soil matrix and groundwater relative to *A. mellifera*,
265 which is managed in artificial nests by humans (Gradish et al. 2019). Additionally, *A. mellifera*
266 may prefer different resources within the same habitat (Thompson and Hunt 1999; Leonhardt
267 and Blüthgen 2012), generating variation in exposure. Niche partitioning of floral resources due
268 to differences in traits, such as tongue length, could also mediate pollutant exposure (e.g.,
269 pesticides) (Brittain and Potts 2011). In addition to pollinator traits, heightened PPCP exposure
270 for *B. vosnesenskii* and *A. texanus* may be due to variable production practices across sampling
271 locations, with farmer's application of wastewater byproducts to soil contaminating certain
272 pollinator nesting sites (Karnjanapiboonwong et al. 2010; Shahriar et al. 2021). While we are not
273 aware of wastewater byproduct applications at our sampling locations, prevalence of external
274 pollutants in conventional and organic farms is common (Humann-Guillemint et al. 2019).

275 Beyond species-specific patterns, our results suggest that human development and plant
276 abundance mediated PPCP presence across all three taxa. This result is consistent with previous
277 findings in aquatic systems, where the source concentration (e.g., amount of development) and
278 paths to trophic transfer (e.g., number of plants) are correlated with the presence of PPCPs at
279 higher trophic levels (Richmond et al. 2018). For example, PPCP concentrations in aquatic
280 systems have been shown to be directly proportional with human population size and inversely
281 proportional to distance from a human population center (Bendz et al. 2005; Meyer et al. 2022).

282 PPCPs also can transfer between trophic levels (Richmond et al. 2018), which may lead to non-
283 linear biological processes and consequences, such as our data showing a negative interaction
284 between plant abundance and human development. Consequences of PPCPs within food webs
285 are hard to predict, although they can cause biological responses at physiological (Feijão et al.
286 2020), behavioral (Brodin et al. 2013), population (Hoppe et al. 2012), community (Lee et al.
287 2016), and ecosystem (Richmond et al. 2019) levels. Aside from PPCPs eliciting direct
288 biological responses, PPCPs may co-occur with numerous human disturbances, such as nutrient
289 pollution, that may further obfuscate clear associations between biological consequences and
290 PPCP exposure.

291 To our knowledge, this is the first study to detect PPCPs in bee tissues that were preserved
292 and pinned in a manner similar to museums. While not originally intended for the present study,
293 detailed field notes and extensive covariate collection created the opportunity for us to evaluate
294 PPCP presence across taxa while testing potential mechanistic drivers of PPCP presence.
295 Considering the growing number of pharmaceuticals (Daughton and Ternes 1999) and changes
296 in pesticides used on national and global markets (Douglas et al. 2020), preserved collections,
297 such as in natural history museums, may offer ripe and previously untapped opportunities to
298 explore contaminant accumulation and mixtures within biota. For example, lake sediment cores
299 have been used to reconstruct interdecadal PPCP contamination loadings (Anger et al. 2013).
300 Similarly, museum collections may empower reconstructions of exposure, accumulation, and co-
301 contaminant histories, all of which may be useful for detailing how contaminant loading and
302 mixtures may change through time.

303

304 **Synthesizing PPCP patterns across ecosystem types.** Over the past three decades, the
305 study of PPCPs has expanded rapidly, and growing evidence suggests that PPCPs are pervasive
306 micropollutants across aquatic and terrestrial systems alike. Our results suggest that much like
307 aquatic systems, PPCPs tend to concentrate closer to their sources and have potential to enter
308 food webs when vectors for trophic transfer are present. Additionally, our results demonstrate
309 species-specific differences for PPCP uptake. Similar to patterns observed in PPCP accumulation
310 in aquatic and riparian systems (Meador et al. 2016; Richmond et al. 2018), *B. vosnesenskii* and
311 *A. texanus* had higher probabilities of PPCP detections relative to *A. mellifera*, implying that life
312 histories, behavior, or physiological differences between taxa may play a role in PPCP uptake.

313 Broadly, our results present novel opportunities for assessing PPCP presence throughout
314 food webs and suggest similarities to how terrestrial and aquatic systems accumulate PPCPs.
315 Where PPCP data are rarer, pairing human development, environmental, and ecological data may
316 aid managers in flagging systems that are more associated with PPCPs, and thus susceptible to
317 declines in ecosystem function and services (e.g., pollination). Regardless of the exact trajectory,
318 our study lays a foundation for future basic and applied PPCP research and creates a synoptic
319 view of how organic contaminants may mobilize within aquatic and terrestrial environments.

320

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323 presentation in a previous version of this manuscript.

324

325 **AUTHOR CONTRIBUTIONS**

326 MFM, MRB, BWL, EHB, and DWC designed the experiment. EHB collected field samples. MFM,
327 BWL, and MRB developed scripts analyses and data visualization. MFM and MLA extracted PPCPs and
328 analyzed HPLC/MS outputs. All authors contributed text, edited, and approved the final manuscript.

329

330 **FUNDING**

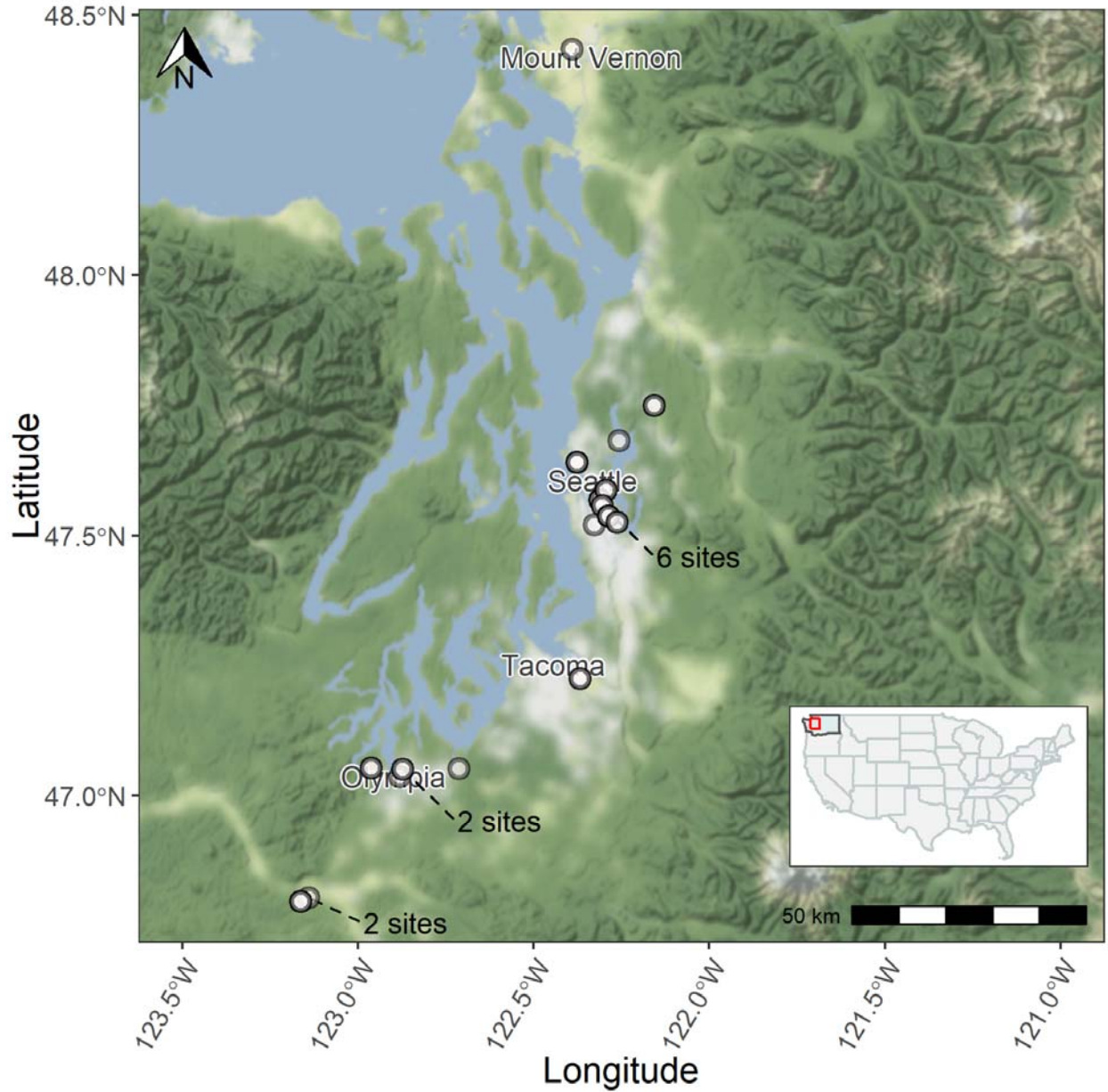
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334

335 **DATA AVAILABILITY STATEMENT**

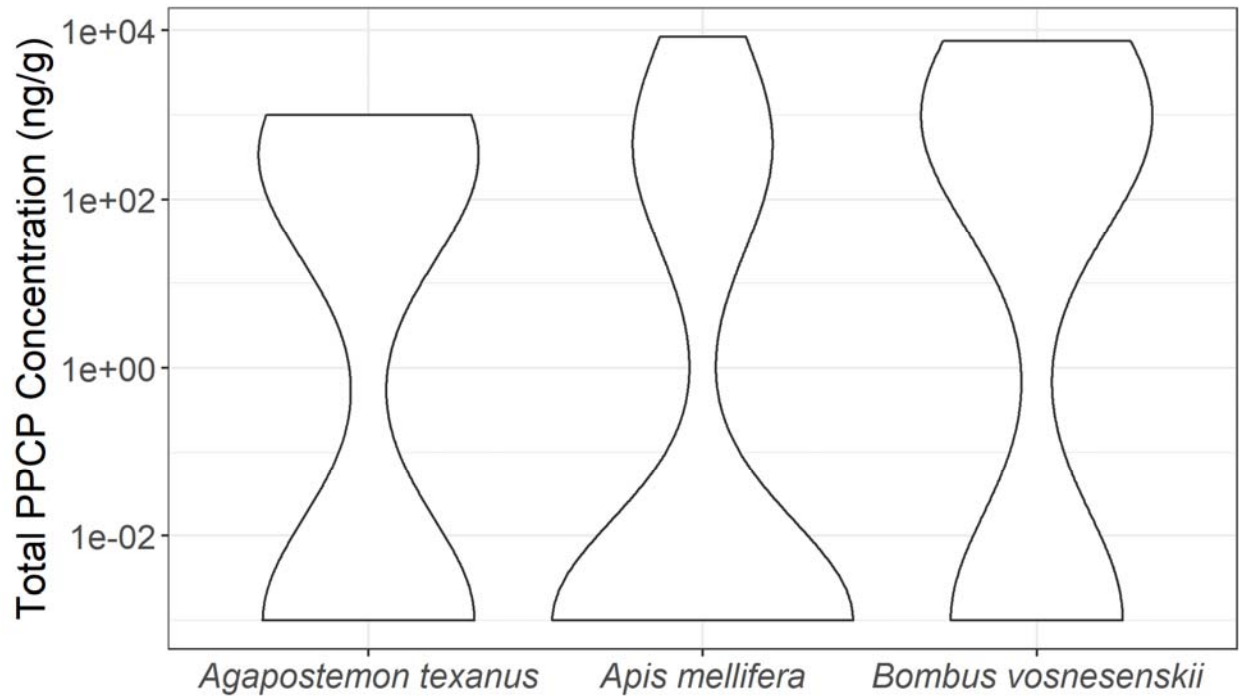
336 All data and scripts used to produce these analyses can be found in this project's companion
337 Open Science Framework repository (Brousil et al. 2022).

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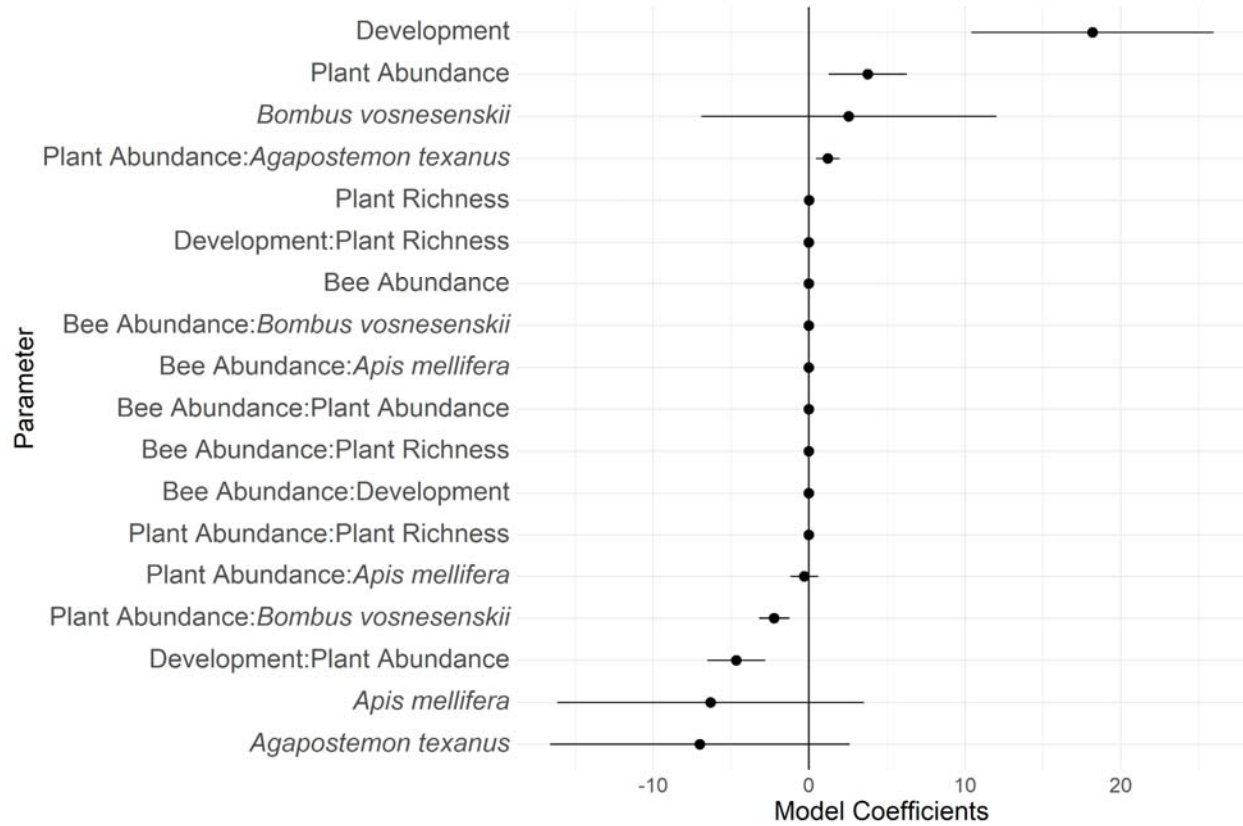
341 Figure 1: Map of bee sampling locations (white points). Map tiles by Stamen Design
342 (stamen.com), under CC BY 3.0 (creativecommons.org/licenses/by/3.0). Data by
343 OpenStreetMap (openstreetmap.org), under ODbL (openstreetmap.org/copyright). State data
344 from US Census Bureau 2019.



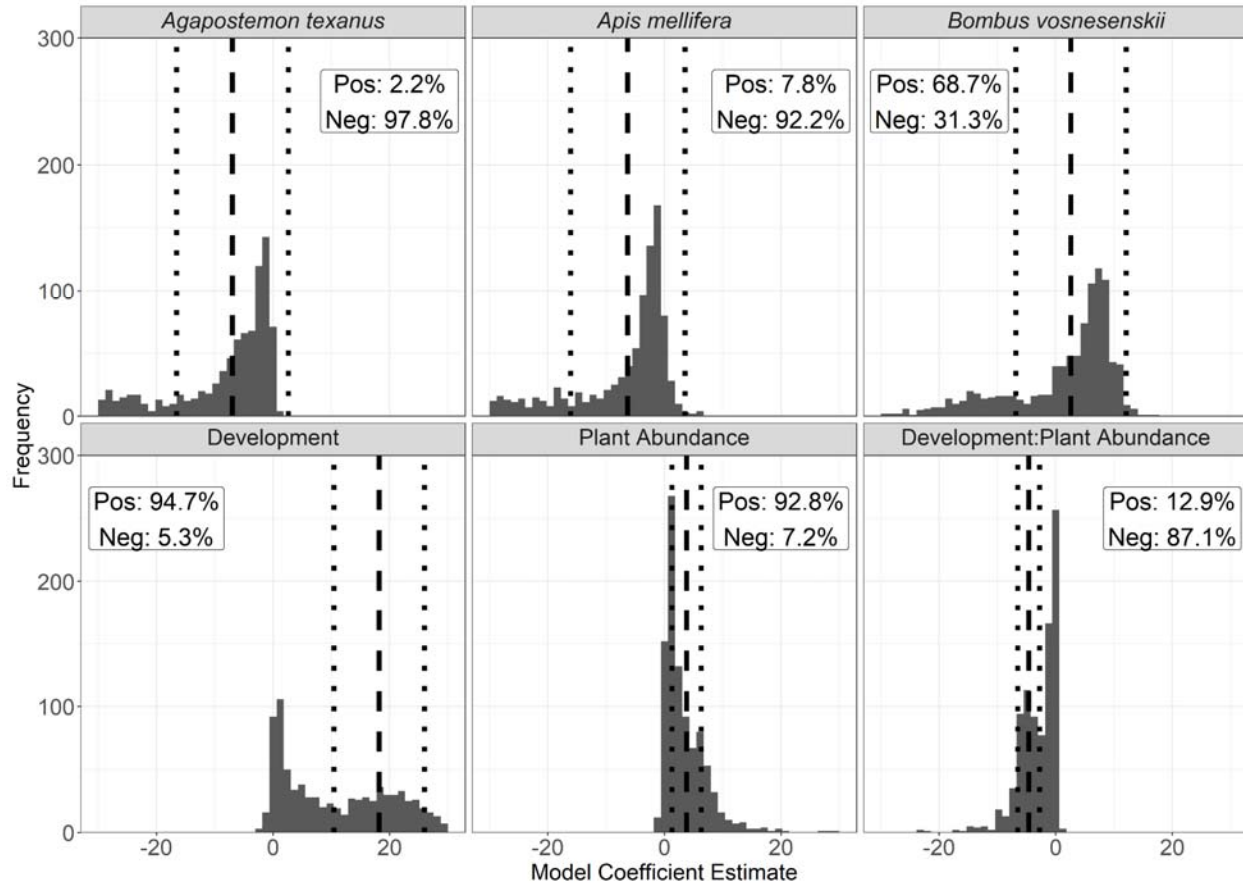
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347 Figure 2: Violin plot of total PPCP concentrations (on log10 scale) in each of the three examined
348 bee taxa. Although the figure shows continuous PPCP concentrations, few samples contained
349 “intermediate” PPCP concentrations, meaning that concentrations were either high (i.e., greater
350 than 500 ng/g) or not detectable. Given that this framework seemed unrealistic and could be a
351 product of how bees were preserved, our main analysis only focused on presence/absence of
352 PPCPs in samples. Therefore, our logistic regression approach should be more conservative than
353 continuous, linear analyses. Nevertheless, trends in PPCP concentrations mirror patterns
354 observed in PPCP presence/absence, where *B. vosnesenskii* and *A. texanus* have higher
355 concentrations and probabilities of PPCP presence relative to *A. mellifera*.

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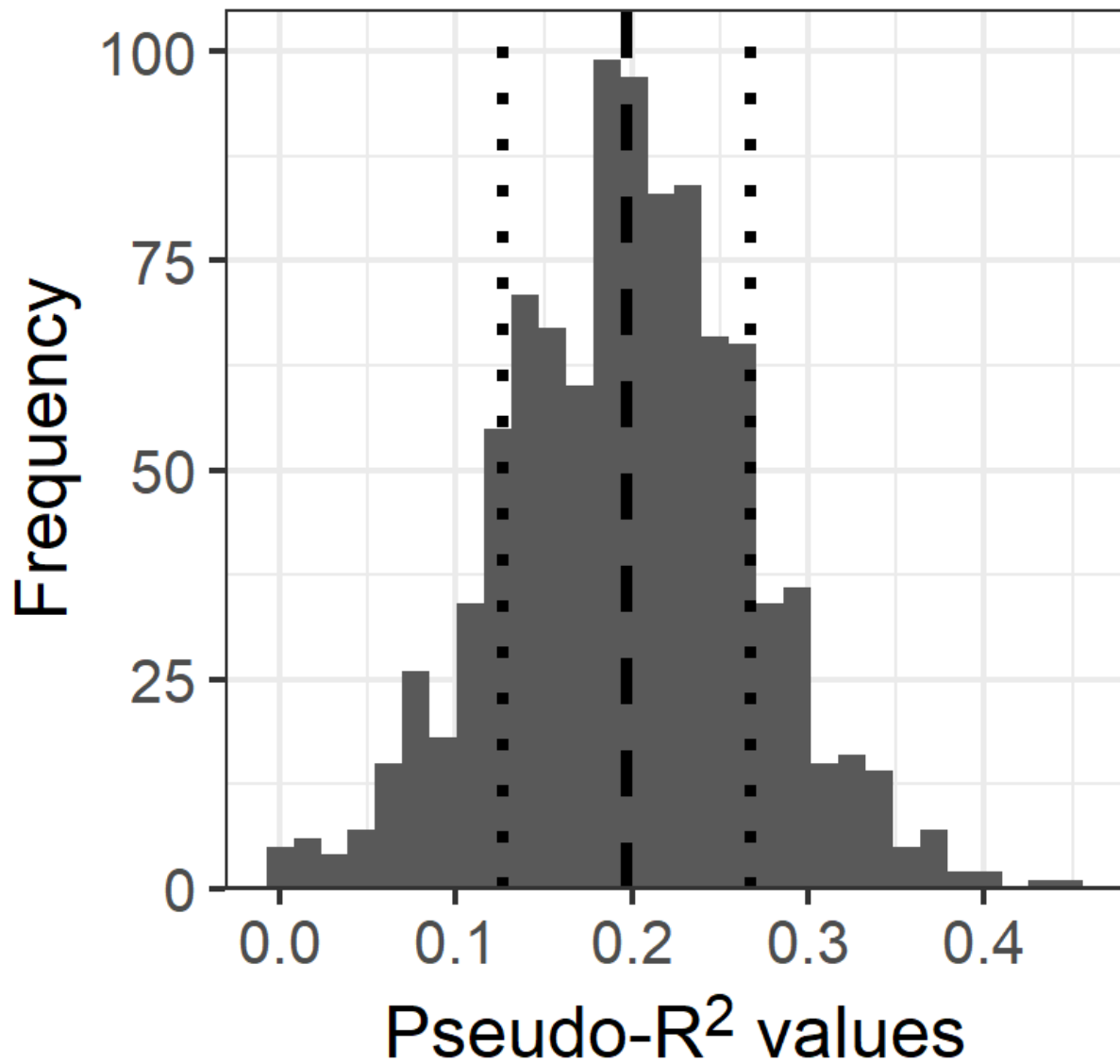


357
358 Figure 2: Coefficients for the final averaged model using all data for training. Points indicate the
359 estimated model coefficient, and bars reflect the adjusted standard error. A value of 0 indicates a
360 variable does not discriminate between PPCP presence or absence. Values greater or less than
361 zero correspond to predictors that are more or less likely to be associated with PPCP presence,
362 respectively. Parameters with a colon indicate two-way interaction terms. Parameters are
363 arranged by decreasing coefficient estimate value.
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Figure 3: Distributions of most influential model coefficients for the final averaged model using 80% of original data as training data. Model averaging protocols were repeated 1,000 times, with each iteration starting with a species-stratified, random subsample. Vertical, dashed lines indicate the coefficient of the parameter that appeared in the final averaged model, when 100% of the data were used. Vertical, dotted lines indicate the adjusted standard error of the coefficient of the parameter that appeared in the final averaged model, when 100% of the data were used. Labels within each facet reflect the percent of coefficients that were positive (i.e., greater than zero) and negative (i.e., less than zero).



374
375 Figure 4: Distribution of pseudo R² values from all averaged models in the permuted analysis.
376 The vertical dashed line represents the mean of the distribution (Pseudo R² = 0.197), and
377 vertical, dotted lines represent one standard deviation from the mean (sd = 0.07).
378

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